Statistics/Data analysis

```
      name:  <unnamed>
       log:  C:\Users\au206393\OneDrive - Aarhus universitet\Desktop\PSRM acceptance l
> og-file\trait_preferences_Ukraine.smcl
  log type:  smcl
 opened on:   6 May 2025, 14:11:31

1 .
2 .
3 . ********************************************************************************
  > ************************************************************
4 . ********************************************************************* RECODINGS **********
  > ************************************************************
5 . ********************************************************************************
  > ************************************************************
6 .
7 . ******************************************************************* Wave 1 ***********
  > ************************************************************
8 .
9 . ************************************************* Demographic background variables
  > ************************************************************
10. * Sex
11. recode w1_q3 (1=0 "Male") (2=1 "Female") (3=.), gen(sex)
  (1,081 differences between w1_q3 and sex)

12.
13. * Age
14. clonevar age = w1_q4

15.
16. * Education
17. recode w1_q6 (1 2 =1 "Primary or High school") (3=2 "Professional-technical (vocatio
  > nal)") (4=3 "Incomplete higher") (5=4 "Bachelor degree") (6 7=5 "Master degree & Doc
  > torate") (8=.), gen(education)
  (1,048 differences between w1_q6 and education)

18. tab education

    RECODE of w1_q6 (6. What is the
  highest level of education that you
                        have complet |      Freq.     Percent        Cum.
-------------------------------------+-----------------------------------
                Primary or High school |         95        8.86        8.86
  Professional-technical (vocational) |        185       17.26       26.12
                    Incomplete higher |         85        7.93       34.05
                      Bachelor degree |        188       17.54       51.59
             Master degree & Doctorate |        519       48.41      100.00
-------------------------------------+-----------------------------------
                               Total |      1,072      100.00

19.
20. * Region
21. clonevar region = w1_region_aggregate

22.
23.
24. ********************************************* Experimental treatment for Ideal Le
  > ader Experiment ********************************************
25. * Experimental treatment for leader trait evaluation questions
26. recode w1_leader_exp_condition (1=1 "Conflict, now") (2=2 "Peace, future"), generate
  > (Context)
  (0 differences between w1_leader_exp_condition and Context)
```

```
27.
28. clonevar Conflict_1 = Context

29.
30.
31. ************************************ Leadership trait preferences in IDEAL LEADER
  >  ************************************************************
32. * Competent
33. recode w1_q14_1 (8=.)
  (57 changes made to w1_q14_1)

34. generate Competence_1 = (w1_q14_1-1)/6
  (57 missing values generated)

35.
36. * Trustworthy
37. recode w1_q14_2 (8=.)
  (36 changes made to w1_q14_2)

38. generate Trustworthy_1 = (w1_q14_2-1)/6
  (36 missing values generated)

39.
40. * Dominant
41. recode w1_q14_3 (8=.)
  (44 changes made to w1_q14_3)

42. generate Dominant_1 = (w1_q14_3-1)/6
  (44 missing values generated)

43.
44. * Generous
45. recode w1_q14_4 (8=.)
  (36 changes made to w1_q14_4)

46. generate Generous_1 = (w1_q14_4-1)/6
  (36 missing values generated)

47.
48. * Strong
49. recode w1_q14_5 (8=.)
  (26 changes made to w1_q14_5)

50. generate Strong_1 = (w1_q14_5-1)/6
  (26 missing values generated)

51.
52. * Warm
53. recode w1_q14_6 (8=.)
  (36 changes made to w1_q14_6)

54. generate Warm_1 = (w1_q14_6-1)/6
  (36 missing values generated)

55.
56. * Tough-minded
57. recode w1_q14_7 (8=.)
  (39 changes made to w1_q14_7)

58. generate Toughminded_1 = (w1_q14_7-1)/6
  (39 missing values generated)
```

```
59.
60. summ Competence_1 Trustworthy_1 Dominant_1 Generous_1 Strong_1 Warm_1 Toughminded_1
```

Variable	Obs	Mean	Std. dev.	Min	Max
Competence_1	1,024	.8802083	.1955989	0	1
Trustworth~1	1,045	.9197767	.1490432	0	1
Dominant_1	1,037	.5940212	.2986719	0	1
Generous_1	1,045	.7279107	.2446149	0	1
Strong_1	1,055	.8840442	.1613094	0	1
Warm_1	1,045	.7090909	.2508124	0	1
Toughminde~1	1,042	.4328215	.2978518	0	1

```
61.
62.
63. *** Exploring dimensions in trait impressions of IDEAL LEADER based on Principal Com
  > ponent Analysis
64. factor Competence_1 Trustworthy_1 Dominant_1 Generous_1 Strong_1 Warm_1 Toughminded_
  > 1, pcf
  (obs=988)
```

Factor analysis/correlation — Number of obs = **988**
Method: principal-component factors — Retained factors = **3**
Rotation: (unrotated) — Number of params = **18**

Factor	Eigenvalue	Difference	Proportion	Cumulative
Factor1	**2.94691**	**1.63211**	**0.4210**	**0.4210**
Factor2	**1.31480**	**0.15262**	**0.1878**	**0.6088**
Factor3	**1.16218**	**0.64733**	**0.1660**	**0.7748**
Factor4	**0.51485**	**0.07283**	**0.0736**	**0.8484**
Factor5	**0.44202**	**0.11587**	**0.0631**	**0.9115**
Factor6	**0.32615**	**0.03306**	**0.0466**	**0.9581**
Factor7	**0.29309**	.	**0.0419**	**1.0000**

LR test: independent vs. saturated: chi2(**21**) = **2287.98** Prob>chi2 = **0.0000**

Factor loadings (pattern matrix) and unique variances

Variable	Factor1	Factor2	Factor3	Uniqueness
Competence_1	**0.7244**	**-0.1156**	**-0.4688**	**0.2422**
Trustworth~1	**0.7713**	**-0.0784**	**-0.4435**	**0.2023**
Dominant_1	**0.5129**	**0.6225**	**0.3071**	**0.2550**
Generous_1	**0.6790**	**-0.3166**	**0.5285**	**0.1593**
Strong_1	**0.7863**	**0.0207**	**-0.2619**	**0.3127**
Warm_1	**0.6192**	**-0.3805**	**0.5378**	**0.1825**
Toughminde~1	**0.3185**	**0.8138**	**0.1195**	**0.2220**

```
65. rotate, oblique oblimin
```

Factor analysis/correlation — Number of obs = **988**
Method: principal-component factors — Retained factors = **3**
Rotation: oblique oblimin (Kaiser off) — Number of params = **18**

Factor	Variance	Proportion	Rotated factors are correlated
Factor1	**2.52111**	**0.3602**	
Factor2	**2.07562**	**0.2965**	
Factor3	**1.65819**	**0.2369**	

LR test: independent vs. saturated: chi2(**21**) = **2287.98** Prob>chi2 = **0.0000**

Rotated factor loadings (pattern matrix) and unique variances

Variable	Factor1	Factor2	Factor3	Uniqueness
Competence_1	**0.8948**	**-0.0361**	**-0.0660**	**0.2422**
Trustworth~1	**0.8997**	**-0.0118**	**-0.0110**	**0.2023**
Dominant_1	**0.0012**	**0.1889**	**0.8147**	**0.2550**
Generous_1	**0.0321**	**0.8991**	**0.0337**	**0.1593**
Strong_1	**0.7419**	**0.1038**	**0.1418**	**0.3127**
Warm_1	**-0.0077**	**0.9118**	**-0.0395**	**0.1825**
Toughminde~1	**0.0112**	**-0.1479**	**0.8895**	**0.2220**

Factor rotation matrix

	Factor1	Factor2	Factor3
Factor1	**0.8723**	**0.7067**	**0.4645**
Factor2	**-0.0817**	**-0.3812**	**0.8526**
Factor3	**-0.4822**	**0.5961**	**0.2395**

```
66.
67. * Creates factor score variables for robustness tests of main results
68. predict Comp_PCA_1 Warm_PCA_1 Domi_PCA_1
  (option regression assumed; regression scoring)
```

Scoring coefficients (method = regression; based on oblimin(0) rotated factors)

Variable	Factor1	Factor2	Factor3
Competence_1	**0.41606**	**-0.03323**	**-0.05736**
Trustworth~1	**0.41715**	**-0.01981**	**-0.02064**
Dominant_1	**-0.01427**	**0.10008**	**0.54781**
Generous_1	**0.00139**	**0.52569**	**0.01066**
Strong_1	**0.34010**	**0.04823**	**0.08340**
Warm_1	**-0.01619**	**0.53466**	**-0.03833**
Toughminde~1	**-0.00586**	**-0.09827**	**0.60254**

```
69. corr Comp_PCA_1 Warm_PCA_1 Domi_PCA_1
  (obs=988)
```

	Comp_P~1	Warm_P~1	Domi_P~1
Comp_PCA_1	**1.0000**		
Warm_PCA_1	**0.3601**	**1.0000**	
Domi_PCA_1	**0.2201**	**0.1460**	**1.0000**

```
70.
71. *** Main outcome variables: Composite scales for dominance, warmth and competence (o
  > n 0-1 scales)
72. egen Domi_scale_1 = rowmean(Dominant_1 Toughminded_1)
  (30 missing values generated)

73.
74. egen Comp_scale_1 = rowmean(Competence_1 Trustworthy_1 Strong_1)
  (24 missing values generated)
```

```
75.
76. egen Warm_scale_1 = rowmean(Warm_1 Generous_1)
  (26 missing values generated)

77.
78. summ Domi_scale_1 Comp_scale_1 Warm_scale_1
```

Variable	Obs	Mean	Std. dev.	Min	Max
Domi_scale_1	**1,051**	**.512052**	**.2588131**	**0**	**1**
Comp_scale_1	**1,057**	**.8936718**	**.145519**	**0**	**1**
Warm_scale_1	**1,055**	**.7185624**	**.2263167**	**0**	**1**

```
79. corr Domi_scale_1 Comp_scale_1 Warm_scale_1
  (obs=1,044)
```

	Domi_s~1	Comp_s~1	Warm_s~1
Domi_scale_1	**1.0000**		
Comp_scale_1	**0.2295**	**1.0000**	
Warm_scale_1	**0.1570**	**0.3740**	**1.0000**

```
80.
81.
82. ************************************** Leadership trait perceptions of CURRENT L
  > EADER, Zelenskyy ********************************************
83. * Competent
84. recode w1_q15_1 (8=.)
  (23 changes made to w1_q15_1)

85. generate Zel_Comp_1 = (w1_q15_1-1)/6
  (23 missing values generated)

86.
87. * Trustworthy
88. recode w1_q15_2 (8=.)
  (23 changes made to w1_q15_2)

89. generate Zel_Trust_1 = (w1_q15_2-1)/6
  (23 missing values generated)

90.
91. * Dominant
92. recode w1_q15_3 (8=.)
  (42 changes made to w1_q15_3)

93. generate Zel_Domi_1 = (w1_q15_3-1)/6
  (42 missing values generated)

94.
95. * Generous
96. recode w1_q15_4 (8=.)
  (38 changes made to w1_q15_4)

97. generate Zel_Generous_1 = (w1_q15_4-1)/6
  (38 missing values generated)

98.
99. * Strong
```

```
100 recode w1_q15_5 (8=.)
  (22 changes made to w1_q15_5)

101 generate Zel_Strong_1 = (w1_q15_5-1)/6
  (22 missing values generated)

102
103 * Warm
104 recode w1_q15_6 (8=.)
  (34 changes made to w1_q15_6)

105 generate Zel_Warm_1 = (w1_q15_6-1)/6
  (34 missing values generated)

106
107 * Tough-minded
108 recode w1_q15_7 (8=.)
  (43 changes made to w1_q15_7)

109 generate Zel_Tough_1 = (w1_q15_7-1)/6
  (43 missing values generated)

110
111
112 summ Zel_Comp_1 Zel_Trust_1 Zel_Domi_1 Zel_Generous_1 Zel_Strong_1 Zel_Warm_1 Zel_To
  > ugh_1
```

Variable	Obs	Mean	Std. dev.	Min	Max
Zel_Comp_1	1,058	.7463768	.3055257	0	1
Zel_Trust_1	1,058	.7840265	.2991983	0	1
Zel_Domi_1	1,039	.563683	.3329842	0	1
Zel_Genero~1	1,043	.7102908	.3076971	0	1
Zel_Strong_1	1,059	.7610954	.3130257	0	1
Zel_Warm_1	1,047	.7184018	.3060069	0	1
Zel_Tough_1	1,038	.3908157	.3077853	0	1

```
113
114
115 ** Creates composite scales for perceptions of Zelenskyy on the same three trait dim
  > ensions as for "ideal leader ratings": dominance, warmth and competence.
116 egen Comp_scale_Zel1 = rowmean(Zel_Comp_1 Zel_Trust_1 Zel_Strong_1)
  (18 missing values generated)

117 alpha Zel_Comp_1 Zel_Trust_1 Zel_Strong_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .0798795
  Number of items in the scale:            3
  Scale reliability coefficient:      0.9458

118
119 egen Warm_scale_Zel1 = rowmean(Zel_Warm_1 Zel_Generous_1)
  (27 missing values generated)

120 alpha Zel_Warm_1 Zel_Generous_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .0786946
  Number of items in the scale:            2
  Scale reliability coefficient:      0.9105
```

```
121
122 egen Domi_scale_Zel1= rowmean(Zel_Domi_1 Zel_Tough_1)
  (29 missing values generated)

123 alpha Zel_Domi_1 Zel_Tough_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:      .050889
  Number of items in the scale:            2
  Scale reliability coefficient:      0.6622

124
125 summ Comp_scale_Zel1 Warm_scale_Zel1 Domi_scale_Zel1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
 Comp_scal~l1 |      1,063    .7637713    .2908337          0          1
 Warm_scal~l1 |      1,054    .7145003    .2931831          0          1
 Domi_scal~l1 |      1,052    .4771863    .2808222          0          1

126
127
128
129
130
131 ************************************************** Self-reported emotional react
  > ions over last week ****************************************
132 * Afraid
133 recode w1_q11_1 (8=.), generate(afraid_1)
  (18 differences between w1_q11_1 and afraid_1)

134 * Frightened
135 recode w1_q11_2 (8=.), generate(frightened_1)
  (14 differences between w1_q11_2 and frightened_1)

136 * Scared
137 recode w1_q11_3 (8=.), generate(scared_1)
  (20 differences between w1_q11_3 and scared_1)

138
139 ** Composite scale for anxiety
140 corr afraid_1 frightened_1 scared_1
  (obs=1,056)

             | afraid_1 fright~1 scared_1
-------------+---------------------------
    afraid_1 |   1.0000
frightened_1 |   0.8053   1.0000
    scared_1 |   0.7059   0.7202   1.0000


141 alpha afraid_1 frightened_1 scared_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:      2.42818
  Number of items in the scale:            3
  Scale reliability coefficient:      0.8956
```

```
142 egen fearfull_scale_W1_7 = rowmean(afraid_1 frightened_1 scared_1)
  (11 missing values generated)

143 generate fearfull_scale_1 = (fearfull_scale_W1_7-1)/6
  (11 missing values generated)

144
145 * Angry
146 recode w1_q11_4 (8=.), generate(angry_1)
  (17 differences between w1_q11_4 and angry_1)

147 * Hostile
148 recode w1_q11_5 (8=.), generate(hostile_1)
  (37 differences between w1_q11_5 and hostile_1)

149 * Disgusted
150 recode w1_q11_6 (8=.), generate(disgusted_1)
  (37 differences between w1_q11_6 and disgusted_1)

151
152 ** Composite scale for agressive emotions
153 corr angry_1 hostile_1 disgusted_1
  (obs=1,021)

             |  angry_1 hostil~1 disgus~1
-------------+---------------------------
     angry_1 |   1.0000
   hostile_1 |   0.5608   1.0000
 disgusted_1 |   0.5519   0.4918   1.0000


154 alpha angry_1 hostile_1 disgusted_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.540405
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7685

155 egen aggressive_scale_W1_7 = rowmean(angry_1 hostile_1 disgusted_1)
  (11 missing values generated)

156 generate aggressive_scale_1 = (aggressive_scale_W1_7-1)/6
  (11 missing values generated)

157
158 * Sad
159 recode w1_q11_7 (8=.), generate(sad_1)
  (11 differences between w1_q11_7 and sad_1)

160 * Lonely
161 recode w1_q11_8 (8=.), generate(lonely_1)
  (23 differences between w1_q11_8 and lonely_1)

162 * Downhearted
163 recode w1_q11_9 (8=.), generate(downhearted_1)
  (16 differences between w1_q11_9 and downhearted_1)

164
165 ** Composite scale for sadness
```

```
166 corr sad_1 lonely_1 downhearted_1
  (obs=1,054)

             |    sad_1 lonely_1 downhe~1
-------------+---------------------------
       sad_1 |   1.0000
    lonely_1 |   0.3668   1.0000
  downhearte~1 |   0.6018   0.4300   1.0000


167 alpha sad_1 lonely_1 downhearted_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.410318
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7230

168 egen sadness_scale_W1_7 = rowmean(sad_1 lonely_1 downhearted_1)
  (7 missing values generated)

169 generate sadness_scale_1 = (sadness_scale_W1_7-1)/6
  (7 missing values generated)

170
171 * Proud
172 recode w1_q11_10 (8=.), generate(proud_1)
  (26 differences between w1_q11_10 and proud_1)

173 * Strong
174 recode w1_q11_11 (8=.), generate(strong_1)
  (23 differences between w1_q11_11 and strong_1)

175 * Confident
176 recode w1_q11_12 (8=.), generate(confident_1)
  (18 differences between w1_q11_12 and confident_1)

177
178 ** Composite scale for self-confident emotions
179 corr proud_1 strong_1 confident_1
  (obs=1,041)

             |  proud_1 strong_1 confid~1
-------------+---------------------------
     proud_1 |   1.0000
    strong_1 |   0.5553   1.0000
  confident_1 |   0.4869   0.6624   1.0000


180 alpha proud_1 strong_1 confident_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.484637
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7922

181 egen selfconf_scale_W1_7 = rowmean(proud_1 strong_1 confident_1)
  (11 missing values generated)
```

```
182 generate selfconf_scale_1 = (selfconf_scale_W1_7-1)/6
  (11 missing values generated)

183
184 summ fearfull_scale_1 aggressive_scale_1 sadness_scale_1 selfconf_scale_1
```

Variable	Obs	Mean	Std. dev.	Min	Max
fearfull_s~1	1,070	.4757529	.2744772	0	1
aggressive~1	1,070	.6362669	.2370329	0	1
sadness_sc~1	1,074	.5127767	.2336689	0	1
selfconf_s~1	1,070	.6069574	.2299484	0	1

```
185
186
187
188
189
190 ****************************************** Self-reported Victimization of Russian
  > Attacks ***************************************************
191 **** Inspects all three items
192 codebook w1_q9_1 w1_q9_2 w1_q9_3
```

---

w1_q9_1
9.1 How often: The invading Russian or pro-Russian forces have directly attacked

---

Type: Numeric (**byte**)
Label: **labels11_wave1**

Range: [**1,6**] Units: **1**
Unique values: **6** Missing .: **0**/**1,081**

Tabulation: Freq.	Numeric	Label
803	**1**	Never
81	**2**	Once
57	**3**	2 to 4 times
28	**4**	5 to 10 times
44	**5**	More than 10 times
68	**6**	Prefer not to say

---

w1_q9_2
9.2 How often: The invading Russian or pro-Russian forces have directly attacked

---

Type: Numeric (**byte**)
Label: **labels11_wave1**

Range: [**1,6**] Units: **1**
Unique values: **6** Missing .: **0**/**1,081**

Tabulation: Freq.	Numeric	Label
646	**1**	Never
100	**2**	Once
96	**3**	2 to 4 times
75	**4**	5 to 10 times
105	**5**	More than 10 times
59	**6**	Prefer not to say

---

w1_q9_3
9.3 How often: The invading Russian or pro-Russian forces have directly attacked

---

Type: Numeric (**byte**)
Label: **labels11_wave1**

Range: [**1,6**] Units: **1**
Unique values: **6** Missing .: **0**/**1,081**

Tabulation: Freq.	Numeric	Label
417	**1**	Never
136	**2**	Once
169	**3**	2 to 4 times
107	**4**	5 to 10 times
197	**5**	More than 10 times
55	**6**	Prefer not to say

193 tab1 w1_q9_1 w1_q9_2 w1_q9_3

-> tabulation of w1_q9_1

9.1 How often: The invading Russian or pro-Russian forces have directly attacked	Freq.	Percent	Cum.
Never	**803**	**74.28**	**74.28**
Once	**81**	**7.49**	**81.78**
2 to 4 times	**57**	**5.27**	**87.05**
5 to 10 times	**28**	**2.59**	**89.64**
More than 10 times	**44**	**4.07**	**93.71**
Prefer not to say	**68**	**6.29**	**100.00**
Total	**1,081**	**100.00**	

-> tabulation of w1_q9_2

9.2 How often: The invading Russian or pro-Russian forces have directly attacked	Freq.	Percent	Cum.
Never	**646**	**59.76**	**59.76**
Once	**100**	**9.25**	**69.01**
2 to 4 times	**96**	**8.88**	**77.89**
5 to 10 times	**75**	**6.94**	**84.83**
More than 10 times	**105**	**9.71**	**94.54**
Prefer not to say	**59**	**5.46**	**100.00**
Total	**1,081**	**100.00**	

-> tabulation of w1_q9_3

9.3 How often: The invading Russian or pro-Russian forces have directly attacked	Freq.	Percent	Cum.
Never	**417**	**38.58**	**38.58**
Once	**136**	**12.58**	**51.16**
2 to 4 times	**169**	**15.63**	**66.79**
5 to 10 times	**107**	**9.90**	**76.69**
More than 10 times	**197**	**18.22**	**94.91**
Prefer not to say	**55**	**5.09**	**100.00**
Total	**1,081**	**100.00**	

```
194 recode w1_q9_1 w1_q9_2 w1_q9_3 (6=.)
  (68 changes made to w1_q9_1)
  (59 changes made to w1_q9_2)
  (55 changes made to w1_q9_3)

195 rename w1_q9_1 w1_victim_self

196 rename w1_q9_2 w1_victim_family

197 rename w1_q9_3 w1_victim_other

198 corr w1_victim_self w1_victim_family w1_victim_other
  (obs=993)

             | w1_vic~f w1_vic~y w1_vic~r
-------------+---------------------------
  w1_victim_~f |   1.0000
  w1_victim_~y |   0.4590   1.0000
  w1_victim_~r |   0.3194   0.6462   1.0000


199 alpha w1_victim_self w1_victim_family w1_victim_other

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .8512443
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7277

200
201 ** Generates victimization scale
202 egen Victimization_W1_5 = rowmean(w1_victim_self w1_victim_family w1_victim_other)
  (38 missing values generated)

203 generate Victimization_1 = (Victimization_W1_5-1)/4
  (38 missing values generated)

204 summ Victimization_1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
  Victimizat~1 |      1,043    .2472435    .2787995          0          1

205
206
207
208 ****************************************** Identification with Ukraine, Russia and
  >  Europe ******************************************************
209 ** Ukraine
210 recode w1_q23_1 (8=.), generate(Ukraine_ID_W1_7)
  (29 differences between w1_q23_1 and Ukraine_ID_W1_7)

211 recode w1_q24_1 (8=.), generate(Ukraine_close_W1_7)
  (24 differences between w1_q24_1 and Ukraine_close_W1_7)

212 corr Ukraine_ID_W1_7 Ukraine_close_W1_7
  (obs=1,045)

             | U~D_W1_7 U~e_W1_7
-------------+------------------
  Ukraine_ID~7 |   1.0000
  Ukraine_cl~7 |   0.7665   1.0000
```

```
213 egen ID_Ukraine_W1_7 = rowmean(Ukraine_ID_W1_7 Ukraine_close_W1_7)
  (17 missing values generated)

214 generate ID_Ukraine_1 = (ID_Ukraine_W1_7-1)/6
  (17 missing values generated)

215
216 ** Russia
217 recode w1_q23_2 (8=.), generate(Russia_ID_W1_7)
  (48 differences between w1_q23_2 and Russia_ID_W1_7)

218 recode w1_q24_2 (8=.), generate(Russia_close_W1_7)
  (37 differences between w1_q24_2 and Russia_close_W1_7)

219 corr Russia_ID_W1_7 Russia_close_W1_7
  (obs=1,025)

             | R~D_W1_7 R~e_W1_7
-------------+------------------
Russia_ID_~7 |   1.0000
Russia_clo~7 |   0.7635   1.0000


220 egen ID_Russia_W1_7 = rowmean(Russia_ID_W1_7 Russia_close_W1_7)
  (29 missing values generated)

221 generate ID_Russia_1 = (ID_Russia_W1_7-1)/6
  (29 missing values generated)

222
223 ** Europe
224 recode w1_q23_3 (8=.), generate(Europe_ID_W1_7)
  (40 differences between w1_q23_3 and Europe_ID_W1_7)

225 recode w1_q24_3 (8=.), generate(Europe_close_W1_7)
  (27 differences between w1_q24_3 and Europe_close_W1_7)

226 corr Europe_ID_W1_7 Europe_close_W1_7
  (obs=1,037)

             | E~D_W1_7 E~e_W1_7
-------------+------------------
Europe_ID_~7 |   1.0000
Europe_clo~7 |   0.7978   1.0000


227 egen ID_Europe_W1_7 = rowmean(Europe_ID_W1_7 Europe_close_W1_7)
  (23 missing values generated)

228 generate ID_Europe_1 = (ID_Europe_W1_7-1)/6
  (23 missing values generated)

229
230 summ ID_Ukraine_1 ID_Russia_1 ID_Europe_1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
ID_Ukraine_1 |      1,064    .9348371    .1526613          0          1
 ID_Russia_1 |      1,052    .1351394    .2279704          0          1
 ID_Europe_1 |      1,058    .7063642    .2693415          0          1
```

```
231
232
233
234
235 ********************************************************************************
  > ************************************************************
236 ************************************************************ WAVE 2 *************
  > ************************************************************
237 ********************************************************************************
  > ************************************************************
238
239 ***************************************** Leadership trait preferences in IDEAL LEA
  > DER ************************************************************
240 * Competent
241 recode w2_q12_1 (8=.)
  (28 changes made to w2_q12_1)

242 generate Competence_2 = (w2_q12_1-1)/6
  (298 missing values generated)

243
244 * Trustworthy
245 recode w2_q12_2 (8=.)
  (31 changes made to w2_q12_2)

246 generate Trustworthy_2 = (w2_q12_2-1)/6
  (301 missing values generated)

247
248 * Dominant
249 recode w2_q12_3 (8=.)
  (50 changes made to w2_q12_3)

250 generate Dominant_2 = (w2_q12_3-1)/6
  (320 missing values generated)

251
252 * Generous
253 recode w2_q12_4 (8=.)
  (38 changes made to w2_q12_4)

254 generate Generous_2 = (w2_q12_4-1)/6
  (308 missing values generated)

255
256 * Strong
257 recode w2_q12_5 (8=.)
  (27 changes made to w2_q12_5)

258 generate Strong_2 = (w2_q12_5-1)/6
  (297 missing values generated)

259
260 * Warm
261 recode w2_q12_6 (8=.)
  (45 changes made to w2_q12_6)

262 generate Warm_2 = (w2_q12_6-1)/6
  (315 missing values generated)
```

```
263
264 * Tough-minded
265 recode w2_q12_7 (8=.)
  (45 changes made to w2_q12_7)

266 generate Toughminded_2 = (w2_q12_7-1)/6
  (315 missing values generated)

267
268 summ Competence_2 Trustworthy_2 Dominant_2 Generous_2 Strong_2 Warm_2 Toughminded_2
```

Variable	Obs	Mean	Std. dev.	Min	Max
Competence_2	783	.8997446	.1773713	0	1
Trustworth~2	780	.9213675	.1557599	0	1
Dominant_2	761	.5516864	.2980696	0	1
Generous_2	773	.6836999	.2757289	0	1
Strong_2	784	.8988095	.1647398	0	1
Warm_2	766	.6668842	.2672514	0	1
Toughminde~2	766	.5047868	.2929434	0	1

```
269
270 *** Exploring dimensions in trait impressions of IDEAL LEADER based on Principal Com
  > ponent Analysis
271 factor Competence_2 Trustworthy_2 Dominant_2 Generous_2 Strong_2 Warm_2 Toughminded_
  > 2, pcf
  (obs=741)

  Factor analysis/correlation                    Number of obs    =        741
      Method: principal-component factors        Retained factors =          3
      Rotation: (unrotated)                      Number of params =         18
```

Factor	Eigenvalue	Difference	Proportion	Cumulative
Factor1	2.58584	1.11377	0.3694	0.3694
Factor2	1.47207	0.29144	0.2103	0.5797
Factor3	1.18063	0.62712	0.1687	0.7484
Factor4	0.55350	0.07261	0.0791	0.8274
Factor5	0.48090	0.10175	0.0687	0.8961
Factor6	0.37914	0.03122	0.0542	0.9503
Factor7	0.34793	.	0.0497	1.0000

```
    LR test: independent vs. saturated:  chi2(21) = 1362.36 Prob>chi2 = 0.0000

  Factor loadings (pattern matrix) and unique variances
```

Variable	Factor1	Factor2	Factor3	Uniqueness
Competence_2	0.6970	0.0107	-0.3653	0.3806
Trustworth~2	0.7532	-0.0689	-0.4252	0.2472
Dominant_2	0.4018	0.6760	0.3715	0.2436
Generous_2	0.6212	-0.3821	0.5270	0.1903
Strong_2	0.7491	0.1092	-0.3606	0.2969
Warm_2	0.6091	-0.4277	0.5186	0.1772
Toughminde~2	0.2308	0.8182	0.2274	0.2256

```
272 rotate, oblique oblimin
```

Factor analysis/correlation — Number of obs = **741**
Method: principal-component factors — Retained factors = **3**
Rotation: oblique oblimin (Kaiser off) — Number of params = **18**

Factor	Variance	Proportion	Rotated factors are correlated
Factor1	**2.27768**	**0.3254**	
Factor2	**1.88209**	**0.2689**	
Factor3	**1.57754**	**0.2254**	

LR test: independent vs. saturated: chi2(**21**) = **1362.36** Prob>chi2 = **0.0000**

Rotated factor loadings (pattern matrix) and unique variances

Variable	Factor1	Factor2	Factor3	Uniqueness
Competence_2	**0.7857**	**0.0036**	**0.0012**	**0.3806**
Trustworth~2	**0.8711**	**0.0193**	**-0.0836**	**0.2472**
Dominant_2	**0.0055**	**0.1415**	**0.8504**	**0.2436**
Generous_2	**0.0074**	**0.8958**	**0.0259**	**0.1903**
Strong_2	**0.8215**	**-0.0191**	**0.1021**	**0.2969**
Warm_2	**0.0044**	**0.9064**	**-0.0210**	**0.1772**
Toughminde~2	**0.0037**	**-0.1228**	**0.8766**	**0.2256**

Factor rotation matrix

	Factor1	Factor2	Factor3
Factor1	**0.8836**	**0.6759**	**0.3542**
Factor2	**0.0144**	**-0.4518**	**0.8700**
Factor3	**-0.4681**	**0.5823**	**0.3429**

```
273
274 * Creates factor score variables for robustness tests of main results
275 predict Comp_PCA_2 Warm_PCA_2 Domi_PCA_2
```
(option **regression** assumed; regression scoring)

Scoring coefficients (method = regression; based on oblimin(0) rotated factors)

Variable	Factor1	Factor2	Factor3
Competence_2	**0.38310**	**-0.00123**	**-0.00427**
Trustworth~2	**0.42526**	**0.00830**	**-0.06105**
Dominant_2	**-0.00338**	**0.08076**	**0.56248**
Generous_2	**-0.00042**	**0.53956**	**0.01238**
Strong_2	**0.40001**	**-0.01559**	**0.06241**
Warm_2	**-0.00167**	**0.54622**	**-0.01872**
Toughminde~2	**-0.00330**	**-0.07862**	**0.58121**

```
276 corr Comp_PCA_2 Warm_PCA_2 Domi_PCA_2
```
(obs=741)

	Comp_P~2	Warm_P~2	Domi_P~2
Comp_PCA_2	**1.0000**		
Warm_PCA_2	**0.3181**	**1.0000**	
Domi_PCA_2	**0.1650**	**0.0460**	**1.0000**

```
277
278 *** Main outcome variables: Composite scales for dominance, warmth and competence (o
  > n 0-1 scales)
279 egen Domi_scale_2 = rowmean(Dominant_2 Toughminded_2)
  (306 missing values generated)

280
281 egen Comp_scale_2 = rowmean(Competence_2 Trustworthy_2 Strong_2)
  (291 missing values generated)

282
283 egen Warm_scale_2 = rowmean(Warm_2 Generous_2)
  (305 missing values generated)

284
285 summ Domi_scale_2 Comp_scale_2 Warm_scale_2
```

Variable	Obs	Mean	Std. dev.	Min	Max
Domi_scale_2	**775**	**.5288172**	**.2583886**	**0**	**1**
Comp_scale_2	**790**	**.906083**	**.1372004**	**0**	**1**
Warm_scale_2	**776**	**.6743986**	**.2480989**	**0**	**1**

```
286 corr Domi_scale_2 Comp_scale_2 Warm_scale_2
  (obs=767)
```

	Domi_s~2	Comp_s~2	Warm_s~2
Domi_scale_2	**1.0000**		
Comp_scale_2	**0.1922**	**1.0000**	
Warm_scale_2	**0.0637**	**0.3291**	**1.0000**

```
287
288
289 ********************************************* Self-reported emotional reactions over l
  > ast week ***************************************************
290 * Afraid
291 recode w2_q11_1 (8=.), generate(afraid_2)
  (9 differences between w2_q11_1 and afraid_2)

292 * Frightened
293 recode w2_q11_2 (8=.), generate(frightened_2)
  (8 differences between w2_q11_2 and frightened_2)

294 * Scared
295 recode w2_q11_3 (8=.), generate(scared_2)
  (7 differences between w2_q11_3 and scared_2)

296
297 ** Composite scale for anxiety
298 corr afraid_2 frightened_2 scared_2
  (obs=799)
```

	afraid_2	fright~2	scared_2
afraid_2	**1.0000**		
frightened_2	**0.7676**	**1.0000**	
scared_2	**0.7061**	**0.7152**	**1.0000**

```
299 alpha afraid_2 frightened_2 scared_2

  Test scale = mean(unstandardized items)

  Average interitem covariance:     2.187711
  Number of items in the scale:            3
  Scale reliability coefficient:      0.8899

300 egen fearfull_scale_W2_7 = rowmean(afraid_2 frightened_2 scared_2)
  (275 missing values generated)

301 generate fearfull_scale_2 = (fearfull_scale_W2_7-1)/6
  (275 missing values generated)

302
303
304 * Angry
305 recode w2_q11_4 (8=.), generate(angry_2)
  (9 differences between w2_q11_4 and angry_2)

306 * Hostile
307 recode w2_q11_5 (8=.), generate(hostile_2)
  (10 differences between w2_q11_5 and hostile_2)

308 * Disgusted
309 recode w2_q11_6 (8=.), generate(disgusted_2)
  (12 differences between w2_q11_6 and disgusted_2)

310
311 ** Composite scale for agressive emotions
312 corr angry_2 hostile_2 disgusted_2
  (obs=795)

             |  angry_2 hostil~2 disgus~2
-------------+---------------------------
     angry_2 |   1.0000
   hostile_2 |   0.6010   1.0000
 disgusted_2 |   0.5372   0.5186   1.0000


313 alpha angry_2 hostile_2 disgusted_2

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.436327
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7834

314 egen aggressive_scale_W2_7 = rowmean(angry_2 hostile_2 disgusted_2)
  (277 missing values generated)

315 generate aggressive_scale_2 = (aggressive_scale_W2_7-1)/6
  (277 missing values generated)

316
317
318 * Sad
319 recode w2_q11_7 (8=.), generate(sad_2)
  (7 differences between w2_q11_7 and sad_2)
```

```
320 * Lonely
321 recode w2_q11_8 (8=.), generate(lonely_2)
  (10 differences between w2_q11_8 and lonely_2)

322 * Downhearted
323 recode w2_q11_9 (8=.), generate(downhearted_2)
  (11 differences between w2_q11_9 and downhearted_2)

324
325 ** Composite scale for sadness
326 corr sad_2 lonely_2 downhearted_2
  (obs=797)

             |    sad_2 lonely_2 downhe~2
-------------+---------------------------
       sad_2 |  1.0000
    lonely_2 |  0.3580   1.0000
downhearte~2 |  0.6237   0.4347   1.0000


327 alpha sad_2 lonely_2 downhearted_2

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.350082
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7259

328 egen sadness_scale_W2_7 = rowmean(sad_2 lonely_2 downhearted_2)
  (276 missing values generated)

329 generate sadness_scale_2 = (sadness_scale_W2_7-1)/6
  (276 missing values generated)

330
331
332 * Proud
333 recode w2_q11_10 (8=.), generate(proud_2)
  (13 differences between w2_q11_10 and proud_2)

334 * Strong
335 recode w2_q11_11 (8=.), generate(strong_2)
  (11 differences between w2_q11_11 and strong_2)

336 * Confident
337 recode w2_q11_12 (8=.), generate(confident_2)
  (8 differences between w2_q11_12 and confident_2)

338
339 ** Composite scale for self-confident emotions
340 corr proud_2 strong_2 confident_2
  (obs=793)

             |  proud_2 strong_2 confid~2
-------------+---------------------------
     proud_2 |  1.0000
    strong_2 |  0.5883   1.0000
 confident_2 |  0.5508   0.7228   1.0000
```

```
341 alpha proud_2 strong_2 confident_2

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.534326
  Number of items in the scale:            3
  Scale reliability coefficient:      0.8307

342 egen selfconf_scale_W2_7 = rowmean(proud_2 strong_2 confident_2)
  (276 missing values generated)

343 generate selfconf_scale_2 = (selfconf_scale_W2_7-1)/6
  (276 missing values generated)

344
345 summ fearfull_scale_2 aggressive_scale_2 sadness_scale_2 selfconf_scale_2

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
  fearfull_s~2 |        806    .4276261    .2618236          0          1
  aggressive~2 |        804    .6732311    .2259084          0          1
  sadness_sc~2 |        805    .5142857    .2271874          0          1
  selfconf_s~2 |        805    .6017253    .2261692          0          1

346
347
348
349 ****************************************** Self-reported Victimization of Russian A
  > ttacks ***************************************************
350 **** Recodes all three items
351 codebook w2_q8_1 w2_q8_2 w2_q8_3

-------------------------------------------------------------------------------
w2_q8_1
      8. Please tell us how often the events described below have happened over the la
-------------------------------------------------------------------------------

                  Type: Numeric (byte)
                 Label: labels12_wave2

                 Range: [1,6]                         Units: 1
         Unique values: 6                         Missing .: 270/1,081

            Tabulation: Freq.   Numeric  Label
                         653         1  Never
                          43         2  Once
                          37         3  2 to 4 times
                          23         4  5 to 10 times
                          16         5  More than 10 times
                          39         6  Prefer not to say
                         270         .

-------------------------------------------------------------------------------
w2_q8_2
      8. Please tell us how often the events described below have happened over the la
-------------------------------------------------------------------------------

                  Type: Numeric (byte)
                 Label: labels12_wave2

                 Range: [1,6]                         Units: 1
         Unique values: 6                         Missing .: 270/1,081

            Tabulation: Freq.   Numeric  Label
                         498         1  Never
                          94         2  Once
                          92         3  2 to 4 times
                          36         4  5 to 10 times
                          53         5  More than 10 times
                          38         6  Prefer not to say
                         270         .
```

w2_q8_3
8. Please tell us how often the events described below have happened over the la

Type: Numeric (**byte**)
Label: **labels12_wave2**

Range: [**1,6**] Units: **1**
Unique values: **6** Missing .: **270/1,081**

Tabulation: Freq.	Numeric	Label
307	**1**	Never
114	**2**	Once
146	**3**	2 to 4 times
78	**4**	5 to 10 times
123	**5**	More than 10 times
43	**6**	Prefer not to say
270	**.**	

352 tab1 w2_q8_1 w2_q8_2 w2_q8_3

-> tabulation of w2_q8_1

8. Please tell us how often the events described below have happened over the la	Freq.	Percent	Cum.
Never	**653**	**80.52**	**80.52**
Once	**43**	**5.30**	**85.82**
2 to 4 times	**37**	**4.56**	**90.38**
5 to 10 times	**23**	**2.84**	**93.22**
More than 10 times	**16**	**1.97**	**95.19**
Prefer not to say	**39**	**4.81**	**100.00**
Total	**811**	**100.00**	

-> tabulation of w2_q8_2

8. Please tell us how often the events described below have happened over the la	Freq.	Percent	Cum.
Never	**498**	**61.41**	**61.41**
Once	**94**	**11.59**	**73.00**
2 to 4 times	**92**	**11.34**	**84.34**
5 to 10 times	**36**	**4.44**	**88.78**
More than 10 times	**53**	**6.54**	**95.31**
Prefer not to say	**38**	**4.69**	**100.00**
Total	**811**	**100.00**	

-> tabulation of w2_q8_3

8. Please tell us how often the events described below have happened over the la	Freq.	Percent	Cum.
Never	**307**	**37.85**	**37.85**
Once	**114**	**14.06**	**51.91**
2 to 4 times	**146**	**18.00**	**69.91**
5 to 10 times	**78**	**9.62**	**79.53**
More than 10 times	**123**	**15.17**	**94.70**

Prefer not to say	**43**	**5.30**	**100.00**
Total	**811**	**100.00**	

```
353 recode w2_q8_1 w2_q8_2 w2_q8_3 (6=.)
  (39 changes made to w2_q8_1)
  (38 changes made to w2_q8_2)
  (43 changes made to w2_q8_3)

354 rename w2_q8_1 w2_victim_self

355 rename w2_q8_2 w2_victim_family

356 rename w2_q8_3 w2_victim_other

357 corr w2_victim_self w2_victim_family w2_victim_other
  (obs=752)
```

	w2_vic~f	w2_vic~y	w2_vic~r
w2_victim_~f	**1.0000**		
w2_victim_~y	**0.4004**	**1.0000**	
w2_victim_~r	**0.2456**	**0.6305**	**1.0000**

```
358 alpha w2_victim_self w2_victim_family w2_victim_other

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .6259643
  Number of items in the scale:            3
  Scale reliability coefficient:      0.6847

359
360
361 ** Generates victimization scale
362 egen Victimization_W2_5 = rowmean(w2_victim_self w2_victim_family w2_victim_other)
  (296 missing values generated)

363 generate Victimization_2 = (Victimization_W2_5-1)/4
  (296 missing values generated)

364 summ Victimization_2
```

Variable	Obs	Mean	Std. dev.	Min	Max
Victimizat~2	**785**	**.2165074**	**.2450247**	**0**	**1**

```
365
366
367 ********************************************* Identification with Ukraine, Russia an
  > d Europe ***************************************************
368 ** Ukraine
369 recode w2_q13_1 (8=.), generate(Ukraine_ID_W2_7)
  (12 differences between w2_q13_1 and Ukraine_ID_W2_7)

370 recode w2_q14_1 (8=.), generate(Ukraine_close_W2_7)
  (10 differences between w2_q14_1 and Ukraine_close_W2_7)

371 corr Ukraine_ID_W2_7 Ukraine_close_W2_7
  (obs=797)
```

	U~D_W2_7	U~e_W2_7
Ukrai~D_W2_7	**1.0000**	
Ukrai~e_W2_7	**0.7919**	**1.0000**

```
372 egen ID_Ukraine_W2_7 = rowmean(Ukraine_ID_W2_7 Ukraine_close_W2_7)
  (278 missing values generated)

373 generate ID_Ukraine_2 = (ID_Ukraine_W2_7-1)/6
  (278 missing values generated)

374
375 ** Russia
376 recode w2_q13_2 (8=.), generate(Russia_ID_W2_7)
  (15 differences between w2_q13_2 and Russia_ID_W2_7)

377 recode w2_q14_2 (8=.), generate(Russia_close_W2_7)
  (14 differences between w2_q14_2 and Russia_close_W2_7)

378 corr Russia_ID_W2_7 Russia_close_W2_7
  (obs=793)

             | R~D_W2_7 R~e_W2_7
-------------+------------------
Russia_I~2_7 |   1.0000
Russia_c~2_7 |   0.8163   1.0000


379 egen ID_Russia_W2_7 = rowmean(Russia_ID_W2_7 Russia_close_W2_7)
  (281 missing values generated)

380 generate ID_Russia_2 = (ID_Russia_W2_7-1)/6
  (281 missing values generated)

381
382 ** Europe
383 recode w2_q13_3 (8=.), generate(Europe_ID_W2_7)
  (21 differences between w2_q13_3 and Europe_ID_W2_7)

384 recode w2_q14_3 (8=.), generate(Europe_close_W2_7)
  (19 differences between w2_q14_3 and Europe_close_W2_7)

385 corr Europe_ID_W2_7 Europe_close_W2_7
  (obs=786)

             | E~D_W2_7 E~e_W2_7
-------------+------------------
Europe_I~2_7 |   1.0000
Europe_c~2_7 |   0.8779   1.0000


386 egen ID_Europe_W2_7 = rowmean(Europe_ID_W2_7 Europe_close_W2_7)
  (285 missing values generated)

387 generate ID_Europe_2 = (ID_Europe_W2_7-1)/6
  (285 missing values generated)

388
389 summ ID_Ukraine_2 ID_Russia_2 ID_Europe_2

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
ID_Ukraine_2 |        803    .9432337    .1384083          0          1
 ID_Russia_2 |        800    .0653125    .1658194          0          1
 ID_Europe_2 |        796    .6871859    .2807245          0          1
```

```
390
391
392
393 ********************************** Creates variable for whether data for all trait r
  > ating variables is present **********************************
394 summ Domi_scale_2 Comp_scale_2 Warm_scale_2
```

Variable	Obs	Mean	Std. dev.	Min	Max
Domi_scale_2	**775**	**.5288172**	**.2583886**	**0**	**1**
Comp_scale_2	**790**	**.906083**	**.1372004**	**0**	**1**
Warm_scale_2	**776**	**.6743986**	**.2480989**	**0**	**1**

```
395 generate include = .
  (1,081 missing values generated)

396 replace include = 1 if Domi_scale_1 !=. & Warm_scale_1 !=. & Comp_scale_1 !=. & Domi
  > _scale_2 !=. & Warm_scale_2 !=. & Comp_scale_2 !=.
  (753 real changes made)

397 tab include
```

include	Freq.	Percent	Cum.
1	**753**	**100.00**	**100.00**
Total	**753**	**100.00**	

```
398
399 ** Creates similar variable based on PCA
400 generate include_PCA = .
  (1,081 missing values generated)

401 replace include_PCA = 1 if Domi_PCA_1 !=. & Warm_PCA_1 !=. & Comp_PCA_1 !=. & Domi_P
  > CA_2 !=. & Warm_PCA_2 !=. & Comp_PCA_2 !=.
  (704 real changes made)

402 tab include_PCA
```

include_PCA	Freq.	Percent	Cum.
1	**704**	**100.00**	**100.00**
Total	**704**	**100.00**	

```
403
404
405 ****************************************** Experimental treatment for Ideal Leader E
  > xperiment in Wave 2 ******************************************
406 * Experimental treatment for leader trait evaluation questions; codes all respondent
  > s to be assigned to think of "Conflict, now"
407 generate Conflict_2=1

408
409
410
411 ********************************************************************************
  > ************************************************************
412 ***************************************************** Difference scores **********
  > ************************************************************
```

```
413 ********************************************************************************
  > ************************************************************
414
415 *** Leader trait preferences in ideal leaders
416 * Composite scales
417 generate Domi_scale_diff = Domi_scale_2 - Domi_scale_1
  (317 missing values generated)

418 generate Comp_scale_diff = Comp_scale_2 - Comp_scale_1
  (302 missing values generated)

419 generate Warm_scale_diff = Warm_scale_2 - Warm_scale_1
  (316 missing values generated)

420
421 summ Domi_scale_diff Comp_scale_diff Warm_scale_diff
```

Variable	Obs	Mean	Std. dev.	Min	Max
Domi_scale~f	764	.0224695	.2304394	-1	.8333333
Comp_scale~f	779	.0018186	.1368881	-.8333333	.8333333
Warm_scale~f	765	-.046732	.2221718	-.9166666	.8333333

```
422
423 * Single-item traits
424 generate Dominance_diff = Dominant_2 - Dominant_1
  (339 missing values generated)

425 generate Toughminded_diff = Toughminded_2 - Toughminded_1
  (331 missing values generated)

426
427 generate Competence_diff = Competence_2 - Competence_1
  (328 missing values generated)

428 generate Trustworthy_diff = Trustworthy_2 - Trustworthy_1
  (317 missing values generated)

429 generate Strong_diff = Strong_2 - Strong_1
  (308 missing values generated)

430
431 generate Warm_diff = Warm_2 - Warm_1
  (331 missing values generated)

432 generate Generous_diff = Generous_2 - Generous_1
  (324 missing values generated)

433
434 summ Dominance_diff Toughminded_diff Competence_diff Trustworthy_diff Strong_diff Wa
  > rm_diff Generous_diff
```

Variable	Obs	Mean	Std. dev.	Min	Max
Dominance_~f	742	-.0404313	.2808646	-1	1
Toughminde~f	750	.08	.2888164	-1	1
Competence~f	753	.0055334	.197705	-1	1
Trustworth~f	764	-.0119983	.1558341	-1	.8333333
Strong_diff	773	.0097025	.1809887	-1	.8333333
Warm_diff	750	-.0431111	.2538692	-1	1
Generous_d~f	757	-.0464553	.2611553	-1	.8333333

```
435
436
437 * Variables based on PCA results
438 generate Comp_PCA_diff = Comp_PCA_2 - Comp_PCA_1
  (377 missing values generated)

439 generate Warm_PCA_diff = Warm_PCA_2 - Warm_PCA_1
  (377 missing values generated)

440 generate Domi_PCA_diff = Domi_PCA_2 - Domi_PCA_1
  (377 missing values generated)

441
442
443 *** Emotional reactions
444 generate fearfull_diff = fearfull_scale_2 - fearfull_scale_1
  (281 missing values generated)

445 generate aggressive_diff = aggressive_scale_2 - aggressive_scale_1
  (281 missing values generated)

446
447 generate sadness_diff = sadness_scale_2 - sadness_scale_1
  (280 missing values generated)

448 generate selfconf_diff = selfconf_scale_2 - selfconf_scale_1
  (280 missing values generated)

449
450 summ fearfull_diff aggressive_diff sadness_diff selfconf_diff
```

Variable	Obs	Mean	Std. dev.	Min	Max
fearfull_d~f	800	-.0549306	.2021551	-.8333333	.5555556
aggressive~f	800	.0198611	.1971963	-.8333333	.7777777
sadness_diff	801	-.0074906	.1955764	-.6666666	.7777778
selfconf_d~f	801	-.0044736	.1874851	-.6944444	1

```
451
452 *** Identification with Ukraine, Russia and Europe
453 generate ID_Ukraine_diff = ID_Ukraine_2  - ID_Ukraine_1
  (284 missing values generated)

454 generate ID_Europe_diff = ID_Europe_2 - ID_Europe_1
  (292 missing values generated)

455 generate ID_Russia_diff = ID_Russia_2 - ID_Russia_1
  (295 missing values generated)

456
457 summ ID_Ukraine_diff ID_Europe_diff ID_Russia_diff
```

Variable	Obs	Mean	Std. dev.	Min	Max
ID_Ukraine~f	797	-.0003137	.110713	-.8333333	1
ID_Europe_~f	789	-.0362273	.2276356	-1	1
ID_Russia_~f	786	-.0483461	.1585455	-.75	.6666667

```
458
459 *** Victimization of Russian attacks
```

```
460 generate Victimization_diff = Victimization_2- Victimization_1
  (310 missing values generated)

461
462 summ Victimization_diff
```

Variable	Obs	Mean	Std. dev.	Min	Max
Victimizat~f	**771**	**-.0368029**	**.2661178**	**-1**	**.9166666**

```
463
464
465
466
467
468
469 ********************************************************************************
  > ************************************************************
470 ************************************ SOM.1: Descriptive statistics for key variable
  > s across waves ***********************************************
471 ********************************************************************************
  > ************************************************************
472
473 **** Demographics across waves
474 * Wave 1
475 tab sex
```

RECODE of w1_q3 (3. Sex)	Freq.	Percent	Cum.
Male	**470**	**43.48**	**43.48**
Female	**611**	**56.52**	**100.00**
Total	**1,081**	**100.00**	

```
476 summ age
```

Variable	Obs	Mean	Std. dev.	Min	Max
age	**1,081**	**35.61332**	**8.730129**	**18**	**55**

```
477 tab education
```

RECODE of w1_q6 (6. What is the highest level of education that you have complet	Freq.	Percent	Cum.
Primary or High school	**95**	**8.86**	**8.86**
Professional-technical (vocational)	**185**	**17.26**	**26.12**
Incomplete higher	**85**	**7.93**	**34.05**
Bachelor degree	**188**	**17.54**	**51.59**
Master degree & Doctorate	**519**	**48.41**	**100.00**
Total	**1,072**	**100.00**	

```
478 tab region
```

Region	Freq.	Percent	Cum.
East	**137**	**12.67**	**12.67**
West	**171**	**15.82**	**28.49**
Kyiv	**195**	**18.04**	**46.53**
North	**112**	**10.36**	**56.89**
Centre	**273**	**25.25**	**82.15**
South	**193**	**17.85**	**100.00**
Total	**1,081**	**100.00**	

```
479
480 * wave 2
481 tab sex if include == 1
```

RECODE of w1_q3 (3. Sex)	Freq.	Percent	Cum.
Male	308	40.90	40.90
Female	445	59.10	100.00
Total	753	100.00	

```
482 summ age if include == 1
```

Variable	Obs	Mean	Std. dev.	Min	Max
age	753	36.32537	8.547372	18	54

```
483 tab education if include == 1
```

RECODE of w1_q6 (6. What is the highest level of education that you have complet	Freq.	Percent	Cum.
Primary or High school	57	7.60	7.60
Professional-technical (vocational)	121	16.13	23.73
Incomplete higher	57	7.60	31.33
Bachelor degree	134	17.87	49.20
Master degree & Doctorate	381	50.80	100.00
Total	750	100.00	

```
484 tab region if include == 1
```

Region	Freq.	Percent	Cum.
East	83	11.02	11.02
West	123	16.33	27.36
Kyiv	143	18.99	46.35
North	78	10.36	56.71
Centre	194	25.76	82.47
South	132	17.53	100.00
Total	753	100.00	

```
485
486
487 **** Trait rating scales
488 * Wave 1
489 summ Comp_scale_1 Warm_scale_1 Domi_scale_1
```

Variable	Obs	Mean	Std. dev.	Min	Max
Comp_scale_1	1,057	.8936718	.145519	0	1
Warm_scale_1	1,055	.7185624	.2263167	0	1
Domi_scale_1	1,051	.512052	.2588131	0	1

```
490 alpha Competence_1 Trustworthy_1 Strong_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .0169582
  Number of items in the scale:            3
  Scale reliability coefficient:      0.8117
```

```
491 alpha Warm_1 Generous_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .0414576
  Number of items in the scale:            2
  Scale reliability coefficient:      0.8063

492 alpha Dominant_1 Toughminded_1

  Test scale = mean(unstandardized items)

  Average interitem covariance:      .042429
  Number of items in the scale:            2
  Scale reliability coefficient:      0.6459

493
494
495 * Wave 2
496 summ Comp_scale_2 Warm_scale_2 Domi_scale_2 if include == 1
```

Variable	Obs	Mean	Std. dev.	Min	Max
Comp_scale_2	753	.908219	.132078	0	1
Warm_scale_2	753	.6737494	.2455379	0	1
Domi_scale_2	753	.5281098	.2580758	0	1

```
497 alpha Competence_2 Trustworthy_2 Strong_2 if include==1

  Test scale = mean(unstandardized items)

  Average interitem covariance:      .012883
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7408

498 alpha Warm_2 Generous_2 if include==1

  Test scale = mean(unstandardized items)

  Average interitem covariance:       .04601
  Number of items in the scale:            2
  Scale reliability coefficient:      0.7738

499 alpha Dominant_2 Toughminded_2 if include==1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .0438885
  Number of items in the scale:            2
  Scale reliability coefficient:      0.6701

500
501 **** Emotional reactions
502 * Wave 1
503 summ fearfull_scale_1 aggressive_scale_1 sadness_scale_1 selfconf_scale_1
```

Variable	Obs	Mean	Std. dev.	Min	Max
fearfull_s~1	1,070	.4757529	.2744772	0	1
aggressive~1	1,070	.6362669	.2370329	0	1
sadness_sc~1	1,074	.5127767	.2336689	0	1
selfconf_s~1	1,070	.6069574	.2299484	0	1

```
504 alpha afraid_1 frightened_1 scared_1 if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:      2.41545
  Number of items in the scale:            3
  Scale reliability coefficient:      0.8953

505 alpha sad_1 lonely_1 downhearted_1 if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.404803
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7254

506 alpha proud_1 strong_1 confident_1 if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.457104
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7866

507 alpha angry_1 hostile_1 disgusted_1 if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.456529
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7630

508
509 * wave 2
510 summ fearfull_scale_2 aggressive_scale_2 sadness_scale_2 selfconf_scale_2 if include
  > ==1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
 fearfull_s~2 |        752    .4295582    .2609825          0          1
 aggressive~2 |        750    .6693333    .2272125          0          1
 sadness_sc~2 |        751    .5145362    .2254564          0          1
 selfconf_s~2 |        751    .6013094     .226263          0          1

511 alpha afraid_2 frightened_2 scared_2 if include==1

  Test scale = mean(unstandardized items)

  Average interitem covariance:      2.18168
  Number of items in the scale:            3
  Scale reliability coefficient:      0.8898

512 alpha sad_2 lonely_2 downhearted_2 if include==1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.319197
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7195
```

```
513 alpha proud_2 strong_2 confident_2 if include==1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.528404
  Number of items in the scale:            3
  Scale reliability coefficient:      0.8291

514 alpha angry_2 hostile_2 disgusted_2 if include==1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.449787
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7818

515
516
517 **** Identities
518 * Wave 1
519 sum ID_Ukraine_1 ID_Europe_1 ID_Russia_1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
ID_Ukraine_1 |      1,064    .9348371    .1526613          0          1
 ID_Europe_1 |      1,058    .7063642    .2693415          0          1
 ID_Russia_1 |      1,052    .1351394    .2279704          0          1

520 alpha Ukraine_ID_W1_7 Ukraine_close_W1_7

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .7195476
  Number of items in the scale:            2
  Scale reliability coefficient:      0.8640

521 alpha Russia_ID_W1_7 Russia_close_W1_7

  Test scale = mean(unstandardized items)

  Average interitem covariance:     1.607426
  Number of items in the scale:            2
  Scale reliability coefficient:      0.8649

522 alpha Europe_ID_W1_7 Europe_close_W1_7

  Test scale = mean(unstandardized items)

  Average interitem covariance:     2.259844
  Number of items in the scale:            2
  Scale reliability coefficient:      0.8826

523
524 * Wave 2
525 sum ID_Ukraine_2 ID_Europe_2 ID_Russia_2 if include == 1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
ID_Ukraine_2 |        748     .943516    .1398345          0          1
 ID_Europe_2 |        745    .6887025    .2810555          0          1
 ID_Russia_2 |        746    .0654602    .1666578          0          1
```

```
526 alpha Ukraine_ID_W2_7 Ukraine_close_W2_7 if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .6294436
  Number of items in the scale:            2
  Scale reliability coefficient:      0.8920

527 alpha Russia_ID_W2_7 Russia_close_W2_7 if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .8761365
  Number of items in the scale:            2
  Scale reliability coefficient:      0.9005

528 alpha Europe_ID_W2_7 Europe_close_W2_7 if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     2.632705
  Number of items in the scale:            2
  Scale reliability coefficient:      0.9325

529
530 ** Victimization
531 summ Victimization_1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
Victimizat~1 |      1,043    .2472435    .2787995          0          1

532 alpha w1_victim_self w1_victim_family w1_victim_other

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .8512443
  Number of items in the scale:            3
  Scale reliability coefficient:      0.7277

533 summ Victimization_2 if include == 1

    Variable |        Obs        Mean    Std. dev.       Min        Max
-------------+---------------------------------------------------------
Victimizat~2 |        738    .2140357    .2394265          0          1

534 alpha w2_victim_self w2_victim_family w2_victim_other if include == 1

  Test scale = mean(unstandardized items)

  Average interitem covariance:     .5939174
  Number of items in the scale:            3
  Scale reliability coefficient:      0.6716

535
536
537 ********************************************************************************
  > ************************************************************
538 ************************************************************ MAIN ANALYSES *****
  > ************************************************************
```

```
539 ********************************************************************************
  > ************************************************************
540
541 ****************************************** MAPPING WARTIME LEADER TRAIT PREFEREN
  > CES ******************************************************
542 *** Key results reported in main text and full models in SOM.2
543 reg Comp_scale_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 374
Residual	7.13159372	373	.019119554	F(0, 373)	= 0.00
				Prob > F	= .
Total	7.13159372	373	.019119554	R-squared	= 0.0000
				Adj R-squared	= 0.0000
				Root MSE	= .13827

Comp_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.8989899	.00715	125.73	0.000	.8849306	.9130492

```
544 reg Comp_scale_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 374
Residual	6.52602185	373	.017496037	F(0, 373)	= 0.00
				Prob > F	= .
Total	6.52602185	373	.017496037	R-squared	= 0.0000
				Adj R-squared	= 0.0000
				Root MSE	= .13227

Comp_scale_2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.9091652	.0068397	132.93	0.000	.8957161	.9226143

```
545
546 reg Warm_scale_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 374
Residual	19.7236508	373	.05287842	F(0, 373)	= 0.00
				Prob > F	= .
Total	19.7236508	373	.05287842	R-squared	= 0.0000
				Adj R-squared	= 0.0000
				Root MSE	= .22995

Warm_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.7034314	.0118906	59.16	0.000	.6800504	.7268124

```
547 reg Warm_scale_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 374
Residual	24.7685853	373	.066403714	F(0, 373)	= 0.00
				Prob > F	= .
Total	24.7685853	373	.066403714	R-squared	= 0.0000
				Adj R-squared	= 0.0000
				Root MSE	= .25769

Warm_scale_2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.6691176	.0133248	50.22	0.000	.6429165	.6953188

```
548
549 reg Domi_scale_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	**0**	**0**	.	Number of obs	= **374**
Residual	**25.3232503**	**373**	**.067890751**	F(0, 373)	= **0.00**
				Prob > F	= .
Total	**25.3232503**	**373**	**.067890751**	R-squared	= **0.0000**
				Adj R-squared	= **0.0000**
				Root MSE	= **.26056**

Domi_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	**.5274064**	**.0134732**	**39.14**	**0.000**	**.5009135**	**.5538993**

```
550 reg Domi_scale_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	**0**	**0**	.	Number of obs	= **374**
Residual	**27.3532561**	**373**	**.073333126**	F(0, 373)	= **0.00**
				Prob > F	= .
Total	**27.3532561**	**373**	**.073333126**	R-squared	= **0.0000**
				Adj R-squared	= **0.0000**
				Root MSE	= **.2708**

Domi_scale_2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	**.5211676**	**.0140028**	**37.22**	**0.000**	**.4936333**	**.5487019**

```
551
552 *** Tests differences between traits in wave 1
553 ttest Comp_scale_1==Warm_scale_1 if Conflict_1 == 1 & include == 1
```

Paired t test

Variable	Obs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
Com~le_1	**374**	**.8989899**	**.00715**	**.1382735**	**.8849306**	**.9130492**
Warm~e_1	**374**	**.7034314**	**.0118906**	**.2299531**	**.6800504**	**.7268124**
diff	**374**	**.1955585**	**.0117395**	**.2270313**	**.1724746**	**.2186424**

mean(diff) = mean(**Comp_scale_1 - Warm_scale_1**) t = **16.6582**
H0: mean(diff) = 0 Degrees of freedom = **373**

Ha: mean(diff) < 0 Ha: mean(diff) != 0 Ha: mean(diff) > 0
Pr(T < t) = **1.0000** Pr(|T| > |t|) = **0.0000** Pr(T > t) = **0.0000**

```
554 ttest Comp_scale_1==Domi_scale_1 if Conflict_1 == 1 & include == 1
```

Paired t test

Variable	Obs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
Com~le_1	**374**	**.8989899**	**.00715**	**.1382735**	**.8849306**	**.9130492**
Domi~e_1	**374**	**.5274064**	**.0134732**	**.2605585**	**.5009135**	**.5538993**
diff	**374**	**.3715835**	**.0133583**	**.2583376**	**.3453164**	**.3978505**

mean(diff) = mean(**Comp_scale_1 - Domi_scale_1**) t = **27.8166**
H0: mean(diff) = 0 Degrees of freedom = **373**

Ha: mean(diff) < 0 Ha: mean(diff) != 0 Ha: mean(diff) > 0
Pr(T < t) = **1.0000** Pr(|T| > |t|) = **0.0000** Pr(T > t) = **0.0000**

```
555 ttest Warm_scale_1==Domi_scale_1 if Conflict_1 == 1 & include == 1
```

Paired t test

Variable	Obs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
Warm~e_1	**374**	**.7034314**	**.0118906**	**.2299531**	**.6800504**	**.7268124**
Domi~e_1	**374**	**.5274064**	**.0134732**	**.2605585**	**.5009135**	**.5538993**
diff	**374**	**.176025**	**.0165154**	**.3193936**	**.1435499**	**.2085**

mean(diff) = mean(**Warm_scale_1 - Domi_scale_1**) t = **10.6582**
H0: mean(diff) = 0 Degrees of freedom = **373**

Ha: mean(diff) < 0 Ha: mean(diff) != 0 Ha: mean(diff) > 0
Pr(T < t) = **1.0000** Pr(|T| > |t|) = **0.0000** Pr(T > t) = **0.0000**

```
556
557 *** Tests differences between traits in wave 2
558 ttest Comp_scale_1==Warm_scale_2 if Conflict_2 == 1 & include == 1
```

Paired t test

Variable	Obs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
Com~le_1	**753**	**.907887**	**.0046071**	**.1264234**	**.8988426**	**.9169313**
Warm_s~2	**753**	**.6737494**	**.0089479**	**.2455379**	**.6561836**	**.6913153**
diff	**753**	**.2341375**	**.0094643**	**.2597079**	**.215558**	**.2527171**

mean(diff) = mean(**Comp_scale_1 - Warm_scale_2**) t = **24.7391**
H0: mean(diff) = 0 Degrees of freedom = **752**

Ha: mean(diff) < 0 Ha: mean(diff) != 0 Ha: mean(diff) > 0
Pr(T < t) = **1.0000** Pr(|T| > |t|) = **0.0000** Pr(T > t) = **0.0000**

```
559 ttest Comp_scale_1==Domi_scale_2 if Conflict_2 == 1 & include == 1
```

Paired t test

Variable	Obs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
Com~le_1	**753**	**.907887**	**.0046071**	**.1264234**	**.8988426**	**.9169313**
Domi_s~2	**753**	**.5281098**	**.0094048**	**.2580758**	**.509647**	**.5465726**
diff	**753**	**.3797772**	**.0097278**	**.2669391**	**.3606803**	**.398874**

mean(diff) = mean(**Comp_scale_1 - Domi_scale_2**) t = **39.0404**
H0: mean(diff) = 0 Degrees of freedom = **752**

Ha: mean(diff) < 0 Ha: mean(diff) != 0 Ha: mean(diff) > 0
Pr(T < t) = **1.0000** Pr(|T| > |t|) = **0.0000** Pr(T > t) = **0.0000**

```
560 ttest Warm_scale_1==Domi_scale_2 if Conflict_2 == 1 & include == 1
```

Paired t test

Variable	Obs	Mean	Std. err.	Std. dev.	[95% conf.	interval]
Warm~e_1	**753**	**.7201195**	**.008101**	**.2222977**	**.7042163**	**.7360227**
Domi_s~2	**753**	**.5281098**	**.0094048**	**.2580758**	**.509647**	**.5465726**
diff	**753**	**.1920097**	**.0118247**	**.3244801**	**.1687964**	**.2152231**

mean(diff) = mean(**Warm_scale_1 - Domi_scale_2**) t = **16.2380**
H0: mean(diff) = 0 Degrees of freedom = **752**

Ha: mean(diff) < 0 Ha: mean(diff) != 0 Ha: mean(diff) > 0
Pr(T < t) = **1.0000** Pr(|T| > |t|) = **0.0000** Pr(T > t) = **0.0000**

```
561
562 *** Correlations between preferences for same traits across waves
563 pwcorr Comp_scale_1 Comp_scale_2 if Conflict_1 == 1, sig
```

	Com~le_1	Comp_s~2
Comp_scale_1	**1.0000**	
Comp_scale_2	**0.4252** **0.0000**	**1.0000**

```
564 pwcorr Competence_1 Trustworthy_1 Strong_1 Competence_2 Trustworthy_2 Strong_2 if Co
  > nflict_1 == 1, sig
```

	Compet~1	Trustw~1	Strong_1	Compet~2	Trustw~2	Strong_2
Competence_1	**1.0000**					
Trustworth~1	**0.6473** **0.0000**	**1.0000**				
Strong_1	**0.5173** **0.0000**	**0.5964** **0.0000**	**1.0000**			
Competence_2	**0.3561** **0.0000**	**0.2109** **0.0000**	**0.2188** **0.0000**	**1.0000**		
Trustworth~2	**0.2927** **0.0000**	**0.3245** **0.0000**	**0.2561** **0.0000**	**0.4250** **0.0000**	**1.0000**	
Strong_2	**0.2627** **0.0000**	**0.2959** **0.0000**	**0.3425** **0.0000**	**0.4040** **0.0000**	**0.4687** **0.0000**	**1.0000**

```
565 reg Comp_scale_1 Comp_scale_2 if Conflict_1 == 1
```

Source	SS	df	MS		
Model	**1.32159262**	**1**	**1.32159262**	Number of obs	= **389**
Residual	**5.98867591**	**387**	**.015474615**	F(1, 387)	= **85.40**
				Prob > F	= **0.0000**
				R-squared	= **0.1808**
				Adj R-squared	= **0.1787**
Total	**7.31026853**	**388**	**.018840898**	Root MSE	= **.1244**

Comp_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Comp_scale_2	**.4366544**	**.0472497**	**9.24**	**0.000**	**.3437562**	**.5295526**
_cons	**.5007107**	**.0434152**	**11.53**	**0.000**	**.4153517**	**.5860698**

```
566
567 pwcorr Warm_scale_1 Warm_scale_2 if Conflict_1 == 1, sig
```

	Warm~e_1	Warm_s~2
Warm_scale_1	**1.0000**	
Warm_scale_2	**0.5508** **0.0000**	**1.0000**

```
568 pwcorr Generous_1 Warm_1  Generous_2 Warm_2 if Conflict_1 == 1, sig
```

	Genero~1	Warm_1	Genero~2	Warm_2
Generous_1	**1.0000**			
Warm_1	**0.6699** **0.0000**	**1.0000**		
Generous_2	**0.4936** **0.0000**	**0.4403** **0.0000**	**1.0000**	
Warm_2	**0.4482** **0.0000**	**0.4801** **0.0000**	**0.6290** **0.0000**	**1.0000**

```
569 reg Warm_scale_1 Warm_scale_2 if Conflict_1 == 1
```

Source	SS	df	MS		
Model	**6.0520385**	**1**	**6.0520385**	Number of obs	= **382**
Residual	**13.8962587**	**380**	**.036569102**	F(1, 380)	= **165.50**
				Prob > F	= **0.0000**
				R-squared	= **0.3034**
				Adj R-squared	= **0.3016**
Total	**19.9482972**	**381**	**.052357735**	Root MSE	= **.19123**

Warm_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Warm_scale_2	**.4847426**	**.0376806**	**12.86**	**0.000**	**.4106541**	**.5588311**
_cons	**.3802865**	**.0270122**	**14.08**	**0.000**	**.3271744**	**.4333986**

```
570
571 pwcorr Domi_scale_1 Domi_scale_2 if Conflict_1 == 1, sig
```

	Domi~e_1	Domi_s~2
Domi_scale_1	**1.0000**	
Domi_scale_2	**0.6405** **0.0000**	**1.0000**

```
572 pwcorr Dominant_1 Toughminded_1 Dominant_2 Toughminded_2 if Conflict_1 == 1, sig
```

	Domina~1	Toughm~1	Domina~2	Toughm~2
Dominant_1	**1.0000**			
Toughminde~1	**0.4859** **0.0000**	**1.0000**		
Dominant_2	**0.5467** **0.0000**	**0.4457** **0.0000**	**1.0000**	
Toughminde~2	**0.4058** **0.0000**	**0.5493** **0.0000**	**0.5506** **0.0000**	**1.0000**

```
573 reg Domi_scale_1 Domi_scale_2 if Conflict_1 == 1
```

Source	SS	df	MS		
Model	**10.5973542**	**1**	**10.5973542**	Number of obs	= **379**
Residual	**15.2308481**	**377**	**.040400127**	F(1, 377)	= **262.31**
				Prob > F	= **0.0000**
				R-squared	= **0.4103**
				Adj R-squared	= **0.4087**
Total	**25.8282023**	**378**	**.068328577**	Root MSE	= **.201**

Domi_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Domi_scale_2	**.6183063**	**.0381765**	**16.20**	**0.000**	**.5432407**	**.6933719**
_cons	**.2019736**	**.0225106**	**8.97**	**0.000**	**.1577117**	**.2462356**

```
574
575
576 ***** Figure 1 is produced below after restructuring the data-file from wide to long
577
578
579
580 ************************************************* TESTING THE CONFLICT-SENSITIVITY
  > HYPOTHESIS ************************************************
581 **** Tests if thinking about a peaceful future affect trait preferences in a leader
  > in wave 1
582 *** Key results reported in main text and full models in SOM.3a
583 reg Comp_scale_1 i.Conflict_1
```

Source	SS	df	MS		
Model	**.133615968**	**1**	**.133615968**	Number of obs	= **1,057**
Residual	**22.2280058**	**1,055**	**.0210692**	F(1, 1055)	= **6.34**
				Prob > F	= **0.0119**
				R-squared	= **0.0060**
				Adj R-squared	= **0.0050**
Total	**22.3616218**	**1,056**	**.021175778**	Root MSE	= **.14515**

Comp_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**.0224865**	**.0089293**	**2.52**	**0.012**	**.0049653**	**.0400077**
_cons	**.8824179**	**.0063169**	**139.69**	**0.000**	**.8700227**	**.8948131**

```
584 margins, dydx(Conflict_1) level(95)
```

Conditional marginal effects — Number of obs = **1,057**
Model VCE: **OLS**

Expression: **Linear prediction, predict()**
dy/dx wrt: **2.Conflict_1**

	dy/dx	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**.0224865**	**.0089293**	**2.52**	**0.012**	**.0049653**	**.0400077**

Note: dy/dx for factor levels is the discrete change from the base level.

```
585 marginsplot,  recastci(rcap) yscale(range(-.1(.05).1)) ylabel(-.1(.05).1) recast(sca
  > tter) yline(0) ///
  > xtitle("Peace") ytitle("Marg. Effect of Peace on Competence Importance") title("Comp
  > etence") scheme(s1mono) legend(off)  name(Competence, replace)

  Variables that uniquely identify margins:

586
587 reg Warm_scale_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,055
				F(1, 1053)	=	7.82
Model	.397981906	1	.397981906	Prob > F	=	0.0053
Residual	53.5871122	1,053	.050889945	R-squared	=	0.0074
				Adj R-squared	=	0.0064
Total	53.9850941	1,054	.051219254	Root MSE	=	.22559

Warm_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	.0388455	.0138907	2.80	0.005	.0115888	.0661021
_cons	.6990476	.0098455	71.00	0.000	.6797287	.7183666

```
588 margins, dydx(Conflict_1) level(95)

  Conditional marginal effects                                    Number of obs = 1,055
  Model VCE: OLS

  Expression: Linear prediction, predict()
  dy/dx wrt:  2.Conflict_1
```

	dy/dx	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	.0388455	.0138907	2.80	0.005	.0115888	.0661021

```
  Note: dy/dx for factor levels is the discrete change from the base level.

589 marginsplot,  recastci(rcap) yscale(range(-.1(.05).1)) ylabel(-.1(.05).1) recast(sca
  > tter) yline(0) ///
  > xtitle("Peace")  ytitle("Marg. Effect of Peace on Warmth Importance") title("Warmth"
  > ) scheme(s1mono) legend(off)  name(Warmth, replace)

  Variables that uniquely identify margins:

590
591 reg Domi_scale_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,051
				F(1, 1049)	=	4.41
Model	.294218428	1	.294218428	Prob > F	=	0.0360
Residual	70.039232	1,049	.066767619	R-squared	=	0.0042
				Adj R-squared	=	0.0032
Total	70.3334504	1,050	.066984238	Root MSE	=	.25839

Domi_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	-.0334636	.0159412	-2.10	0.036	-.0647439	-.0021834
_cons	.5288953	.0113096	46.77	0.000	.5067032	.5510873

```
592 margins, dydx(Conflict_1) level(95)

  Conditional marginal effects                              Number of obs = 1,051
  Model VCE: OLS

  Expression: Linear prediction, predict()
  dy/dx wrt:  2.Conflict_1
```

	dy/dx	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1 Peace, future	-.0334636	.0159412	-2.10	0.036	-.0647439	-.0021834

```
  Note: dy/dx for factor levels is the discrete change from the base level.

593 marginsplot,  recastci(rcap) yscale(range(-.1(.05).1)) ylabel(-.1(.05).1) recast(sca
  > tter) yline(0) ///
  > xtitle("Peace")  ytitle("Marg. Effect of Peace on Dominance Importance") title("Domi
  > nance") scheme(s1mono) legend(off)  name(Dominance, replace)

  Variables that uniquely identify margins:

594
595 *** Creates Figure 2
596 graph combine Competence Warmth Dominance, scheme(s1mono) col(3) ysize(3) xsize(6)

597 graph export fig2.pdf, replace
  file fig2.pdf saved as PDF format

598
599
600 ***** Within-subjects test of conflict-sensitivity hypothesis is produced below afte
  > r restructuring the data-file from wide to long
601
602
603
604
605 ***************************** TESTING THE EFFECTS OF EMOTIONAL REACTIONS TO THE WAR
  > ON LEADER TRAIT PREFERENCES **********************************
606 *** Key results reported in main text and full models in SOM.4
607 reg Comp_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS		
Model	.500624175	7	.071517739	Number of obs	726
Residual	12.4404021	718	.017326465	F(7, 718)	4.13
				Prob > F	0.0002
				R-squared	0.0387
				Adj R-squared	0.0293
Total	12.9410263	725	.017849691	Root MSE	.13163

Comp_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0581295	.0267534	-2.17	0.030	-.1106536	-.0056054
aggressive_diff	.0723606	.0264822	2.73	0.006	.0203688	.1243524
sadness_diff	.0503068	.0274303	1.83	0.067	-.0035465	.1041601
selfconf_diff	.0396835	.027631	1.44	0.151	-.0145638	.0939307
ID_Ukraine_diff	.1065099	.0443735	2.40	0.017	.0193925	.1936273
ID_Europe_diff	.014314	.022299	0.64	0.521	-.029465	.0580929
ID_Russia_diff	-.0254169	.0316162	-0.80	0.422	-.0874881	.0366542
_cons	-.0031561	.005406	-0.58	0.560	-.0137696	.0074574

```
608
609 reg Warm_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	726
				F(7, 718)	=	4.46
Model	1.47300002	7	.210428574	Prob > F	=	0.0001
Residual	33.8399859	718	.0471309	R-squared	=	0.0417
				Adj R-squared	=	0.0324
Total	35.3129859	725	.048707567	Root MSE	=	.2171

Warm_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0771967	.0441241	-1.75	0.081	-.1638244	.009431
aggressive_diff	.0914454	.0436769	2.09	0.037	.0056957	.177195
sadness_diff	.1047498	.0452407	2.32	0.021	.01593	.1935697
selfconf_diff	-.0477427	.0455716	-1.05	0.295	-.1372123	.0417269
ID_Ukraine_diff	.0450279	.073185	0.62	0.539	-.0986543	.18871
ID_Europe_diff	.1459127	.0367775	3.97	0.000	.0737083	.218117
ID_Russia_diff	-.0462364	.0521443	-0.89	0.376	-.14861	.0561372
_cons	-.0490779	.0089161	-5.50	0.000	-.0665827	-.0315731

```
610
611 reg Domi_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	726
				F(7, 718)	=	1.25
Model	.443999179	7	.063428454	Prob > F	=	0.2719
Residual	36.3831826	718	.050672956	R-squared	=	0.0121
				Adj R-squared	=	0.0024
Total	36.8271818	725	.050796113	Root MSE	=	.22511

Domi_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	.0067655	.0457521	0.15	0.882	-.0830584	.0965894
aggressive_diff	.099491	.0452884	2.20	0.028	.0105775	.1884044
sadness_diff	.0315586	.0469099	0.67	0.501	-.0605383	.1236556
selfconf_diff	-.0005131	.0472531	-0.01	0.991	-.0932838	.0922576
ID_Ukraine_diff	.0718635	.0758852	0.95	0.344	-.0771199	.220847
ID_Europe_diff	-.0243516	.0381345	-0.64	0.523	-.09922	.0505168
ID_Russia_diff	.0678296	.0540683	1.25	0.210	-.0383212	.1739804
_cons	.0250402	.0092451	2.71	0.007	.0068895	.0431908

```
612
613
614 *** Creates Figure 3
615 * Calculates observed ranges of changes (difference across waves) in fearfull and ag
  > gressive emotions in the sample
616 sum fearfull_diff if include==1
```

Variable	Obs	Mean	Std. dev.	Min	Max
fearfull_d~f	748	-.0581922	.2019064	-.8333333	.5555556

```
617 sum aggressive_diff if include==1
```

Variable	Obs	Mean	Std. dev.	Min	Max
aggressive~f	749	.0140558	.1950568	-.8333333	.7777777

```
618
619 *Plot marginal effects for the ranges of changes in our sample
620 *Requires instalation of coefplot
621 *ssc install coefplot
622 *Comp
623 reg Comp_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS		
Model	.500624175	7	.071517739	Number of obs	= 726
Residual	12.4404021	718	.017326465	F(7, 718)	= 4.13
				Prob > F	= 0.0002
Total	12.9410263	725	.017849691	R-squared	= 0.0387
				Adj R-squared	= 0.0293
				Root MSE	= .13163

Comp_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0581295	.0267534	-2.17	0.030	-.1106536	-.0056054
aggressive_diff	.0723606	.0264822	2.73	0.006	.0203688	.1243524
sadness_diff	.0503068	.0274303	1.83	0.067	-.0035465	.1041601
selfconf_diff	.0396835	.027631	1.44	0.151	-.0145638	.0939307
ID_Ukraine_diff	.1065099	.0443735	2.40	0.017	.0193925	.1936273
ID_Europe_diff	.014314	.022299	0.64	0.521	-.029465	.0580929
ID_Russia_diff	-.0254169	.0316162	-0.80	0.422	-.0874881	.0366542
_cons	-.0031561	.005406	-0.58	0.560	-.0137696	.0074574

```
624 margins, at(c.fearfull_diff=(-.8333333(0.2).5555556)) post

Predictive margins                                      Number of obs = 726
Model VCE: OLS

Expression: Linear prediction, predict()
1._at: fearfull_diff = -.8333333
2._at: fearfull_diff = -.6333333
3._at: fearfull_diff = -.4333333
4._at: fearfull_diff = -.2333333
5._at: fearfull_diff = -.0333333
6._at: fearfull_diff =  .1666667
7._at: fearfull_diff =  .3666667
```

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_at						
1	.0465808	.0212673	2.19	0.029	.0048273	.0883344
2	.0349549	.0161067	2.17	0.030	.0033331	.0665768
3	.023329	.0111271	2.10	0.036	.0014836	.0451745
4	.0117031	.0067422	1.74	0.083	-.0015336	.0249398
5	.0000772	.0049357	0.02	0.988	-.0096129	.0097674
6	-.0115487	.0077798	-1.48	0.138	-.0268225	.0037252
7	-.0231746	.0124076	-1.87	0.062	-.0475341	.0011849

```
625 est store comp_fearful

626
627 reg Comp_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS		
Model	.500624175	7	.071517739	Number of obs	= 726
Residual	12.4404021	718	.017326465	F(7, 718)	= 4.13
				Prob > F	= 0.0002
Total	12.9410263	725	.017849691	R-squared	= 0.0387
				Adj R-squared	= 0.0293
				Root MSE	= .13163

Comp_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0581295	.0267534	-2.17	0.030	-.1106536	-.0056054
aggressive_diff	.0723606	.0264822	2.73	0.006	.0203688	.1243524
sadness_diff	.0503068	.0274303	1.83	0.067	-.0035465	.1041601
selfconf_diff	.0396835	.027631	1.44	0.151	-.0145638	.0939307
ID_Ukraine_diff	.1065099	.0443735	2.40	0.017	.0193925	.1936273
ID_Europe_diff	.014314	.022299	0.64	0.521	-.029465	.0580929
ID_Russia_diff	-.0254169	.0316162	-0.80	0.422	-.0874881	.0366542
_cons	-.0031561	.005406	-0.58	0.560	-.0137696	.0074574

```
628 margins, at(c.aggressive_diff=(-.8333333(0.2).7777777)) post

Predictive margins                                              Number of obs = 726
Model VCE: OLS

Expression: Linear prediction, predict()
1._at: aggressive_diff = -.8333333
2._at: aggressive_diff = -.6333333
3._at: aggressive_diff = -.4333333
4._at: aggressive_diff = -.2333333
5._at: aggressive_diff = -.0333333
6._at: aggressive_diff =  .1666667
7._at: aggressive_diff =  .3666667
8._at: aggressive_diff =  .5666667
9._at: aggressive_diff =  .7666667
```

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_at						
1	-.0600058	.0230719	-2.60	0.009	-.1053023	-.0147094
2	-.0455337	.0179307	-2.54	0.011	-.0807365	-.0103309
3	-.0310616	.0129155	-2.40	0.016	-.0564182	-.005705
4	-.0165895	.0082592	-2.01	0.045	-.0328045	-.0003745
5	-.0021174	.0050718	-0.42	0.676	-.0120748	.00784
6	.0123548	.006272	1.97	0.049	.0000412	.0246683
7	.0268269	.010443	2.57	0.010	.0063244	.0473293
8	.041299	.0153258	2.69	0.007	.0112104	.0713877
9	.0557711	.0204158	2.73	0.006	.0156893	.095853

```
629 est store comp_aggresive

630
631 coefplot (comp_fearful, recast(line) lcolor(green) lwidth(*3) ciopts(recast(rarea) c
  > olor(green%50)) ylab(-0.15(0.05)0.15) ///
  >             label("{&Delta} Fearfull emotions") yline(0, lpattern(dash)))  ///
  >                  (comp_aggresive, recast(line) lcolor(blue) lwidth(*3) ciopts(recast
  > (rarea) color(blue%50)) ylab(-0.15(0.05)0.15) ///
  >                    label("{&Delta} Aggressive emotions")), ytitle("Predicted change
  >  in {bf:competence} preference" " ") scheme(s1mono) at

632         graph save comp, replace
  (file comp.gph not found)
  file comp.gph saved
```

```
633
634 *Warm
635 reg Warm_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	726
				F(7, 718)	=	4.46
Model	1.47300002	7	.210428574	Prob > F	=	0.0001
Residual	33.8399859	718	.0471309	R-squared	=	0.0417
				Adj R-squared	=	0.0324
Total	35.3129859	725	.048707567	Root MSE	=	.2171

Warm_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0771967	.0441241	-1.75	0.081	-.1638244	.009431
aggressive_diff	.0914454	.0436769	2.09	0.037	.0056957	.177195
sadness_diff	.1047498	.0452407	2.32	0.021	.01593	.1935697
selfconf_diff	-.0477427	.0455716	-1.05	0.295	-.1372123	.0417269
ID_Ukraine_diff	.0450279	.073185	0.62	0.539	-.0986543	.18871
ID_Europe_diff	.1459127	.0367775	3.97	0.000	.0737083	.218117
ID_Russia_diff	-.0462364	.0521443	-0.89	0.376	-.14861	.0561372
_cons	-.0490779	.0089161	-5.50	0.000	-.0665827	-.0315731

```
636 margins, at(c.fearfull_diff=(-.8333333(0.2).5555556)) post
```

```
Predictive margins                                      Number of obs = 726
Model VCE: OLS

Expression: Linear prediction, predict()
1._at: fearfull_diff = -.8333333
2._at: fearfull_diff = -.6333333
3._at: fearfull_diff = -.4333333
4._at: fearfull_diff = -.2333333
5._at: fearfull_diff = -.0333333
6._at: fearfull_diff =  .1666667
7._at: fearfull_diff =  .3666667
```

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_at						
1	.0132382	.0350761	0.38	0.706	-.0556257	.0821021
2	-.0022011	.0265647	-0.08	0.934	-.0543548	.0499526
3	-.0176405	.0183518	-0.96	0.337	-.05367	.0183891
4	-.0330798	.0111198	-2.97	0.003	-.054911	-.0112486
5	-.0485191	.0081405	-5.96	0.000	-.0645011	-.0325372
6	-.0639585	.0128312	-4.98	0.000	-.0891496	-.0387674
7	-.0793978	.0204638	-3.88	0.000	-.1195738	-.0392218

```
637 est store warm_fearfull

638
639 reg Warm_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	726
				F(7, 718)	=	4.46
Model	1.47300002	7	.210428574	Prob > F	=	0.0001
Residual	33.8399859	718	.0471309	R-squared	=	0.0417
				Adj R-squared	=	0.0324
Total	35.3129859	725	.048707567	Root MSE	=	.2171

Warm_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0771967	.0441241	-1.75	0.081	-.1638244	.009431
aggressive_diff	.0914454	.0436769	2.09	0.037	.0056957	.177195
sadness_diff	.1047498	.0452407	2.32	0.021	.01593	.1935697
selfconf_diff	-.0477427	.0455716	-1.05	0.295	-.1372123	.0417269
ID_Ukraine_diff	.0450279	.073185	0.62	0.539	-.0986543	.18871
ID_Europe_diff	.1459127	.0367775	3.97	0.000	.0737083	.218117
ID_Russia_diff	-.0462364	.0521443	-0.89	0.376	-.14861	.0561372
_cons	-.0490779	.0089161	-5.50	0.000	-.0665827	-.0315731

```
640 margins, at(c.aggressive_diff=(-.8333333(0.2).7777777)) post

Predictive margins                                      Number of obs = 726
Model VCE: OLS

Expression: Linear prediction, predict()
1._at: aggressive_diff = -.8333333
2._at: aggressive_diff = -.6333333
3._at: aggressive_diff = -.4333333
4._at: aggressive_diff = -.2333333
5._at: aggressive_diff = -.0333333
6._at: aggressive_diff =  .1666667
7._at: aggressive_diff =  .3666667
8._at: aggressive_diff =  .5666667
9._at: aggressive_diff =  .7666667
```

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_at						
1	-.1243505	.0380523	-3.27	0.001	-.1990576	-.0496434
2	-.1060614	.0295729	-3.59	0.000	-.1641211	-.0480017
3	-.0877724	.0213014	-4.12	0.000	-.1295928	-.045952
4	-.0694833	.0136218	-5.10	0.000	-.0962266	-.04274
5	-.0511942	.0083649	-6.12	0.000	-.0676169	-.0347716
6	-.0329051	.0103443	-3.18	0.002	-.0532138	-.0125965
7	-.0146161	.0172235	-0.85	0.396	-.0484306	.0191984
8	.003673	.0252767	0.15	0.885	-.045952	.053298
9	.0219621	.0336717	0.65	0.514	-.0441447	.0880688

```
641 est store warm_aggressive

642
643 coefplot (warm_fearfull, recast(line) lcolor(green) lwidth(*3) ciopts(recast(rarea)
  > color(green%50)) ylab(-0.15(0.05)0.15) ///
  >             label("{&Delta} Fearfull emotions") yline(0, lpattern(dash)))  ///
  >                  (warm_aggressive, recast(line) lcolor(blue) lwidth(*3) ciopts(recas
  > t(rarea) color(blue%50)) ylab(-0.15(0.05)0.15) ///
  >                    label("{&Delta} Aggressive emotions")), ytitle("Predicted change
  >  in {bf:warmth} preference" " ") scheme(s1mono) at

644         graph save warm, replace
  (file warm.gph not found)
  file warm.gph saved
```

```
645
646 *Domi
647 reg Domi_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	726
				F(7, 718)	=	1.25
Model	.443999179	7	.063428454	Prob > F	=	0.2719
Residual	36.3831826	718	.050672956	R-squared	=	0.0121
				Adj R-squared	=	0.0024
Total	36.8271818	725	.050796113	Root MSE	=	.22511

Domi_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	.0067655	.0457521	0.15	0.882	-.0830584	.0965894
aggressive_diff	.099491	.0452884	2.20	0.028	.0105775	.1884044
sadness_diff	.0315586	.0469099	0.67	0.501	-.0605383	.1236556
selfconf_diff	-.0005131	.0472531	-0.01	0.991	-.0932838	.0922576
ID_Ukraine_diff	.0718635	.0758852	0.95	0.344	-.0771199	.220847
ID_Europe_diff	-.0243516	.0381345	-0.64	0.523	-.09922	.0505168
ID_Russia_diff	.0678296	.0540683	1.25	0.210	-.0383212	.1739804
_cons	.0250402	.0092451	2.71	0.007	.0068895	.0431908

```
648 margins, at(c.fearfull_diff=(-.8333333(0.2).5555556)) post

Predictive margins                                              Number of obs = 726
Model VCE: OLS

Expression: Linear prediction, predict()
1._at: fearfull_diff = -.8333333
2._at: fearfull_diff = -.6333333
3._at: fearfull_diff = -.4333333
4._at: fearfull_diff = -.2333333
5._at: fearfull_diff = -.0333333
6._at: fearfull_diff =  .1666667
7._at: fearfull_diff =  .3666667
```

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_at						
1	.0182964	.0363702	0.50	0.615	-.0531083	.0897011
2	.0196495	.0275448	0.71	0.476	-.0344285	.0737275
3	.0210026	.0190289	1.10	0.270	-.0163563	.0583615
4	.0223557	.0115301	1.94	0.053	-.000281	.0449924
5	.0237088	.0084408	2.81	0.005	.0071372	.0402804
6	.0250619	.0133046	1.88	0.060	-.0010586	.0511825
7	.026415	.0212188	1.24	0.214	-.0152433	.0680733

```
649 est store domi_fearfull

650
651 reg Domi_scale_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  >  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	726
				F(7, 718)	=	1.25
Model	.443999179	7	.063428454	Prob > F	=	0.2719
Residual	36.3831826	718	.050672956	R-squared	=	0.0121
				Adj R-squared	=	0.0024
Total	36.8271818	725	.050796113	Root MSE	=	.22511

Domi_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	.0067655	.0457521	0.15	0.882	-.0830584	.0965894
aggressive_diff	.099491	.0452884	2.20	0.028	.0105775	.1884044
sadness_diff	.0315586	.0469099	0.67	0.501	-.0605383	.1236556
selfconf_diff	-.0005131	.0472531	-0.01	0.991	-.0932838	.0922576
ID_Ukraine_diff	.0718635	.0758852	0.95	0.344	-.0771199	.220847
ID_Europe_diff	-.0243516	.0381345	-0.64	0.523	-.09922	.0505168
ID_Russia_diff	.0678296	.0540683	1.25	0.210	-.0383212	.1739804
_cons	.0250402	.0092451	2.71	0.007	.0068895	.0431908

```
652 margins, at(c.aggressive_diff=(-.8333333(0.2).7777777)) post

Predictive margins                                              Number of obs = 726
Model VCE: OLS

Expression: Linear prediction, predict()
1._at: aggressive_diff = -.8333333
2._at: aggressive_diff = -.6333333
3._at: aggressive_diff = -.4333333
4._at: aggressive_diff = -.2333333
5._at: aggressive_diff = -.0333333
6._at: aggressive_diff =  .1666667
7._at: aggressive_diff =  .3666667
8._at: aggressive_diff =  .5666667
9._at: aggressive_diff =  .7666667
```

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_at						
1	-.0611827	.0394563	-1.55	0.121	-.1386462	.0162808
2	-.0412845	.030664	-1.35	0.179	-.1014864	.0189174
3	-.0213863	.0220873	-0.97	0.333	-.0647498	.0219771
4	-.0014881	.0141244	-0.11	0.916	-.0292182	.0262419
5	.01841	.0086736	2.12	0.034	.0013815	.0354386
6	.0383082	.010726	3.57	0.000	.0172502	.0593662
7	.0582064	.017859	3.26	0.001	.0231443	.0932686
8	.0781046	.0262093	2.98	0.003	.0266486	.1295606
9	.0980028	.034914	2.81	0.005	.029457	.1665486

```
653 est store domi_aggressive

654
655 coefplot (domi_fearfull, recast(line) lcolor(green) lwidth(*3) ciopts(recast(rarea)
  > color(green%50)) ylab(-0.15(0.05)0.15) ///
  >             label("{&Delta} Fearfull emotions") yline(0, lpattern(dash)))  ///
  >                  (domi_aggressive, recast(line) lcolor(blue) lwidth(*3) ciopts(recas
  > t(rarea) color(blue%50)) ylab(-0.15(0.05)0.15) ///
  >                    label("{&Delta} Aggressive emotions")), ytitle("Predicted change
  >  in {bf:dominance} preference" " ")scheme(s1mono) at

656         graph save domi, replace
  (file domi.gph not found)
  file domi.gph saved
```

```
657
658 graph combine comp.gph warm.gph domi.gph, ycommon xsize(12) ysize(3) col(3) scale(1.
  > 45) scheme(s1mono)

659 graph export fig3.pdf, replace
  file fig3.pdf saved as PDF format

660
661
662 ********************************************************************************
  > ************************************************************
663 ************************************************************ SUPPLEMENTARY ANALYSES **
  > ************************************************************
664 ********************************************************************************
  > ************************************************************
665
666 ********************************************** SOM.6. ANALYSES USING SINGLE-ITEM TRA
  > IT VARIABLES ************************************************
667 *** SOM 6.a: Average trait importance across survey waves based on single-item trait
  >  measures
668 * Wave 1
669 reg Competence_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	**0**	**0**	**.**	Number of obs	= **363**
Residual	**14.3027241**	**362**	**.039510287**	F(0, 362)	= **0.00**
Total	**14.3027241**	**362**	**.039510287**	Prob > F	= **.**
				R-squared	= **0.0000**
				Adj R-squared	= **0.0000**
				Root MSE	= **.19877**

Competence_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	**.8797061**	**.0104328**	**84.32**	**0.000**	**.8591896**	**.9002227**

```
670 reg Trustworthy_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	**0**	**0**	**.**	Number of obs	= **369**
Residual	**5.52318604**	**368**	**.015008658**	F(0, 368)	= **0.00**
Total	**5.52318604**	**368**	**.015008658**	Prob > F	= **.**
				R-squared	= **0.0000**
				Adj R-squared	= **0.0000**
				Root MSE	= **.12251**

Trustworth~1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	**.932701**	**.0063776**	**146.25**	**0.000**	**.9201599**	**.9452421**

```
671 reg Strong_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	**0**	**0**	**.**	Number of obs	= **374**
Residual	**9.90411453**	**373**	**.026552586**	F(0, 373)	= **0.00**
Total	**9.90411453**	**373**	**.026552586**	Prob > F	= **.**
				R-squared	= **0.0000**
				Adj R-squared	= **0.0000**
				Root MSE	= **.16295**

Strong_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	**.8881462**	**.0084259**	**105.41**	**0.000**	**.8715779**	**.9047144**

```
672
673 reg Warm_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 371
Residual	22.7845154	370	.061579771	F(0, 370)	= 0.00
				Prob > F	= .
				R-squared	= 0.0000
				Adj R-squared	= 0.0000
Total	22.7845154	370	.061579771	Root MSE	= .24815

Warm_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.6922731	.0128834	53.73	0.000	.6669392	.7176071

```
674 reg Generous_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 371
Residual	23.1495948	370	.062566472	F(0, 370)	= 0.00
				Prob > F	= .
				R-squared	= 0.0000
				Adj R-squared	= 0.0000
Total	23.1495948	370	.062566472	Root MSE	= .25013

Generous_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.7160827	.0129863	55.14	0.000	.6905465	.7416188

```
675
676 reg Dominant_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 369
Residual	33.3515499	368	.090629212	F(0, 368)	= 0.00
				Prob > F	= .
				R-squared	= 0.0000
				Adj R-squared	= 0.0000
Total	33.3515499	368	.090629212	Root MSE	= .30105

Dominant_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.6102078	.0156719	38.94	0.000	.5793901	.6410254

```
677 reg Toughminded_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 374
Residual	34.1527031	373	.091562207	F(0, 373)	= 0.00
				Prob > F	= .
				R-squared	= 0.0000
				Adj R-squared	= 0.0000
Total	34.1527031	373	.091562207	Root MSE	= .30259

Toughminde~1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.4465241	.0156467	28.54	0.000	.4157573	.4772908

```
678
679 * Wave 2
680 reg Competence_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS	Number of obs	=	373
				F(0, 372)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	12.3297588	372	.033144513	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	12.3297588	372	.033144513	Root MSE	=	.18206

Competence_2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.9008043	.0094265	95.56	0.000	.8822683	.9193402

```
681 reg Trustworthy_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS	Number of obs	=	371
				F(0, 370)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	9.78870927	370	.026455971	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	9.78870927	370	.026455971	Root MSE	=	.16265

Trustworth~2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.9218329	.0084445	109.16	0.000	.9052276	.9384382

```
682 reg Strong_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS	Number of obs	=	374
				F(0, 373)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	9.33571006	373	.025028713	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	9.33571006	373	.025028713	Root MSE	=	.1582

Strong_2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.9064171	.0081806	110.80	0.000	.8903313	.9225029

```
683
684 reg Warm_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS	Number of obs	=	370
				F(0, 369)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	29.1673414	369	.079044286	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	29.1673414	369	.079044286	Root MSE	=	.28115

Warm_2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.6581081	.0146162	45.03	0.000	.6293666	.6868496

```
685 reg Generous_1 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 371
				F(0, 370)	= 0.00
Residual	23.1495948	370	.062566472	Prob > F	= .
				R-squared	= 0.0000
Total	23.1495948	370	.062566472	Adj R-squared	= 0.0000
				Root MSE	= .25013

Generous_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.7160827	.0129863	55.14	0.000	.6905465	.7416188

```
686
687 reg Dominant_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 369
				F(0, 368)	= 0.00
Residual	34.2083703	368	.092957528	Prob > F	= .
				R-squared	= 0.0000
Total	34.2083703	368	.092957528	Adj R-squared	= 0.0000
				Root MSE	= .30489

Dominant_2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.5438121	.0158719	34.26	0.000	.5126011	.5750231

```
688 reg Toughminded_2 if Conflict_1 == 1 & include == 1
```

Source	SS	df	MS		
Model	0	0	.	Number of obs	= 372
				F(0, 371)	= 0.00
Residual	35.1925771	371	.094858698	Prob > F	= .
				R-squared	= 0.0000
Total	35.1925771	371	.094858698	Adj R-squared	= 0.0000
				Root MSE	= .30799

Toughminde~2	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.4977599	.0159686	31.17	0.000	.4663595	.5291602

```
689
690
691
692 *** SOM 6.b: Testing the conflict-sensitivity hypothesis (with single-item trait var
  > iables)
693 reg Competence_1 i.Conflict_1
```

Source	SS	df	MS		
Model	.458968535	1	.458968535	Number of obs	= 1,024
				F(1, 1022)	= 12.13
Residual	38.6799199	1,022	.03784728	Prob > F	= 0.0005
				R-squared	= 0.0117
Total	39.1388884	1,023	.038258933	Adj R-squared	= 0.0108
				Root MSE	= .19454

Competence_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	.042346	.0121601	3.48	0.001	.0184843	.0662076
_cons	.8587459	.0086571	99.20	0.000	.8417582	.8757336

```
694 eststo SOM3b_model1

695
696 reg Trustworthy_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,045
				F(1, 1043)	=	1.03
Model	.022871706	1	.022871706	Prob > F	=	0.3105
Residual	23.1684098	1,043	.02221324	R-squared	=	0.0010
				Adj R-squared	=	0.0000
Total	23.1912815	1,044	.022213871	Root MSE	=	.14904

Trustworthy_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	.009357	.0092213	1.01	0.310	-.0087375	.0274515
_cons	.9150579	.0065485	139.74	0.000	.9022082	.9279076

```
697 eststo SOM3b_model2

698
699 reg Strong_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,055
				F(1, 1053)	=	1.85
Model	.047997023	1	.047997023	Prob > F	=	0.1745
Residual	27.3778584	1,053	.025999866	R-squared	=	0.0018
				Adj R-squared	=	0.0008
Total	27.4258554	1,054	.026020736	Root MSE	=	.16124

Strong_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	.01349	.0099286	1.36	0.175	-.0059922	.0329721
_cons	.8772928	.0070239	124.90	0.000	.8635104	.8910753

```
700 eststo SOM3b_model3

701
702 reg Warm_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,045
				F(1, 1043)	=	10.75
Model	.669721507	1	.669721507	Prob > F	=	0.0011
Residual	65.0050234	1,043	.062325046	R-squared	=	0.0102
				Adj R-squared	=	0.0092
Total	65.6747449	1,044	.062906844	Root MSE	=	.24965

Warm_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	.0506319	.0154457	3.28	0.001	.0203236	.0809401
_cons	.6836538	.0109479	62.45	0.000	.6621715	.7051362

```
703 eststo SOM3b_model4

704
705 reg Generous_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,045
				F(1, 1043)	=	2.86
Model	.170818797	1	.170818797	Prob > F	=	0.0911
Residual	62.2984504	1,043	.059730058	R-squared	=	0.0027
				Adj R-squared	=	0.0018
Total	62.4692692	1,044	.059836465	Root MSE	=	.2444

Generous_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	.0255708	.0151207	1.69	0.091	-.0040997	.0552414
_cons	.7150641	.0107175	66.72	0.000	.6940337	.7360945

```
706 eststo SOM3b_model5

707
708 reg Dominant_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,037
				F(1, 1035)	=	4.57
Model	.406655908	1	.406655908	Prob > F	=	0.0327
Residual	92.0096065	1,035	.088898171	R-squared	=	0.0044
				Adj R-squared	=	0.0034
Total	92.4162624	1,036	.089204886	Root MSE	=	.29816

Dominant_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	-.0396076	.0185187	-2.14	0.033	-.0759462	-.003269
_cons	.6140351	.013164	46.65	0.000	.5882039	.6398663

```
709 eststo SOM3b_model6

710
711 reg Toughminded_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	1,042
				F(1, 1040)	=	2.35
Model	.208266783	1	.208266783	Prob > F	=	0.1255
Residual	92.1447924	1,040	.088600762	R-squared	=	0.0023
				Adj R-squared	=	0.0013
Total	92.3530592	1,041	.088715715	Root MSE	=	.29766

Toughminded_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	-.0282754	.0184424	-1.53	0.126	-.064464	.0079132
_cons	.4470135	.0130658	34.21	0.000	.4213752	.4726518

```
712 eststo SOM3b_model7

713
714 etable, estimates(SOM3b_model1 SOM3b_model2 SOM3b_model3 SOM3b_model4 SOM3b_model5 S
  > OM3b_model6 SOM3b_model7) mstat(N) mstat(r2_a) showstars showstarsnote export(SOM3b.
  > docx)

 ──────────────────────────────────────────────────────────────────────────────────────
   ────────────────────────────────────────
                                           Competence_1 Trustworthy_1  Strong_1    Warm
 > _1   Generous_1 Dominant_1 Toughminded_1
 ──────────────────────────────────────────────────────────────────────────────────────
   ────────────────────────────────────────
 RECODE of w1_leader_exp_condition (Split)
 >
   Peace, future                              0.042 **
 >
                                            (0.012)
 >
 Intercept                                    0.859 **
 >
                                            (0.009)
 >
 RECODE of w1_leader_exp_condition (Split)
 >
   Peace, future                                            0.009
 >
                                                          (0.009)
 >
 Intercept                                                  0.915 **
 >
                                                          (0.007)
 >
 RECODE of w1_leader_exp_condition (Split)
 >
   Peace, future                                                         0.013
 >
                                                                       (0.010)
 >
 Intercept                                                               0.877 **
 >
                                                                       (0.007)
 >
 RECODE of w1_leader_exp_condition (Split)
 >
   Peace, future                                                                   0.05
 > 1 **
                                                                                 (0.015
 > )
 Intercept                                                                         0.68
 > 4 **
                                                                                 (0.011
 > )
 RECODE of w1_leader_exp_condition (Split)
 >
   Peace, future
 >        0.026

 >      (0.015)
 Intercept
 >        0.715 **

 >      (0.011)
 RECODE of w1_leader_exp_condition (Split)
 >
   Peace, future
 >                  -0.040 *

 >                 (0.019)
 Intercept
 >                   0.614 **
```

```
>                 (0.013)
RECODE of w1_leader_exp_condition (Split)
>
  Peace, future
>                              -0.028

>                             (0.018)
Intercept
>                               0.447 **

>                             (0.013)
Number of observations                        1024          1045        1055       104
> 5       1045       1037         1042
Adjusted R-squared                            0.01          0.00        0.00       0.0
> 1       0.00       0.00         0.00
_____________________________________________________________________________________
  ____________________________________
** p<.01, * p<.05
(collection ETable exported to file SOM3b.docx)

715
716
717
718 *** SOM 6.c: Testing the role of emotional reactions (with single-item trait variabl
> es)
719 reg Competence_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
>  c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS
Model	.511978013	7	.073139716
Residual	27.226699	701	.038839799
Total	27.738677	708	.039178922

Number of obs	=	709
F(7, 701)	=	1.88
Prob > F	=	0.0696
R-squared	=	0.0185
Adj R-squared	=	0.0087
Root MSE	=	.19708

Competence_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0820309	.0405701	-2.02	0.044	-.1616844	-.0023775
aggressive_diff	.0246926	.0397871	0.62	0.535	-.0534236	.1028088
sadness_diff	.073109	.0418207	1.75	0.081	-.0089998	.1552177
selfconf_diff	.0326776	.0419613	0.78	0.436	-.0497073	.1150625
ID_Ukraine_diff	.1071831	.0667399	1.61	0.109	-.023851	.2382173
ID_Europe_diff	.0255527	.0335915	0.76	0.447	-.0403992	.0915047
ID_Russia_diff	-.0486822	.0480928	-1.01	0.312	-.1431054	.0457411
_cons	-.002027	.0082118	-0.25	0.805	-.0181497	.0140958

```
720
721 reg Trustworthy_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_dif
> f c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS
Model	.538074039	7	.07686772
Residual	16.1267334	709	.022745745
Total	16.6648074	716	.023274871

Number of obs	=	717
F(7, 709)	=	3.38
Prob > F	=	0.0015
R-squared	=	0.0323
Adj R-squared	=	0.0227
Root MSE	=	.15082

Trustworthy_d~f	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0525905	.0308852	-1.70	0.089	-.1132279	.0080469
aggressive_diff	.0782978	.0305555	2.56	0.011	.0183076	.1382879
sadness_diff	.051061	.0317049	1.61	0.108	-.0111858	.1133078
selfconf_diff	-.0036653	.0319384	-0.11	0.909	-.0663705	.0590399
ID_Ukraine_diff	.147509	.0517147	2.85	0.004	.0459767	.2490412
ID_Europe_diff	.0268788	.0257113	1.05	0.296	-.0236007	.0773583
ID_Russia_diff	.0378519	.0364933	1.04	0.300	-.0337959	.1094997
_cons	-.0135873	.0062387	-2.18	0.030	-.025836	-.0013387

```
722
723 reg Strong_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff c.I
  > D_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	725
				F(7, 717)	=	3.98
Model	.84250916	7	.120358451	Prob > F	=	0.0003
Residual	21.6568776	717	.03020485	R-squared	=	0.0374
				Adj R-squared	=	0.0280
Total	22.4993868	724	.031076501	Root MSE	=	.1738

Strong_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0497466	.0353671	-1.41	0.160	-.119182	.0196887
aggressive_diff	.0978924	.0352371	2.78	0.006	.0287121	.1670728
sadness_diff	.0480876	.0362421	1.33	0.185	-.0230657	.1192408
selfconf_diff	.0913268	.0369648	2.47	0.014	.0187546	.163899
ID_Ukraine_diff	.0855933	.0586684	1.46	0.145	-.0295892	.2007758
ID_Europe_diff	-.0029368	.0294678	-0.10	0.921	-.0607903	.0549166
ID_Russia_diff	-.0591288	.0420704	-1.41	0.160	-.1417246	.0234671
_cons	.0057704	.0071378	0.81	0.419	-.0082431	.019784

```
724
725 reg Warm_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff c.ID_
  > Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	716
				F(7, 708)	=	3.76
Model	1.61252579	7	.230360828	Prob > F	=	0.0005
Residual	43.4207976	708	.06132881	R-squared	=	0.0358
				Adj R-squared	=	0.0263
Total	45.0333234	715	.062983669	Root MSE	=	.24765

Warm_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.1045381	.0505709	-2.07	0.039	-.2038249	-.0052513
aggressive_diff	.0844083	.0501873	1.68	0.093	-.0141255	.1829421
sadness_diff	.1436926	.0525139	2.74	0.006	.040591	.2467942
selfconf_diff	-.0209114	.0524423	-0.40	0.690	-.1238724	.0820496
ID_Ukraine_diff	.0618101	.084831	0.73	0.466	-.1047402	.2283605
ID_Europe_diff	.1390536	.0434035	3.20	0.001	.0538386	.2242686
ID_Russia_diff	.0162804	.0603714	0.27	0.787	-.102248	.1348087
_cons	-.0437774	.0102851	-4.26	0.000	-.0639704	-.0235845

```
726
727 reg Generous_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff c
  > .ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	720
				F(7, 712)	=	2.56
Model	1.20632126	7	.172331609	Prob > F	=	0.0131
Residual	47.9538226	712	.067350874	R-squared	=	0.0245
				Adj R-squared	=	0.0149
Total	49.1601439	719	.06837294	Root MSE	=	.25952

Generous_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0482396	.0533254	-0.90	0.366	-.1529335	.0564543
aggressive_diff	.0826283	.0523894	1.58	0.115	-.0202279	.1854846
sadness_diff	.0848324	.0543814	1.56	0.119	-.0219346	.1915994
selfconf_diff	-.0578568	.0549396	-1.05	0.293	-.1657198	.0500063
ID_Ukraine_diff	.018716	.0875917	0.21	0.831	-.1532528	.1906849
ID_Europe_diff	.1319676	.0441594	2.99	0.003	.0452694	.2186657
ID_Russia_diff	-.0928652	.062403	-1.49	0.137	-.215381	.0296506
_cons	-.0509659	.0107097	-4.76	0.000	-.0719922	-.0299395

```
728
729 reg Dominance_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff
  > c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS		
Model	.41743423	7	.059633461	Number of obs	709
				F(7, 701)	0.75
Residual	55.5947869	701	.079307827	Prob > F	0.6280
				R-squared	0.0075
				Adj R-squared	-0.0025
Total	56.0122212	708	.079113307	Root MSE	.28162

Dominance_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.0079294	.0576976	-0.14	0.891	-.1212102	.1053515
aggressive_diff	.0016055	.0576353	0.03	0.978	-.1115529	.114764
sadness_diff	.1055763	.0596606	1.77	0.077	-.0115587	.2227112
selfconf_diff	-.0196567	.0606294	-0.32	0.746	-.1386936	.0993802
ID_Ukraine_diff	.1119102	.0961817	1.16	0.245	-.0769286	.300749
ID_Europe_diff	-.0079579	.0487477	-0.16	0.870	-.1036668	.087751
ID_Russia_diff	.0270551	.0686863	0.39	0.694	-.1078004	.1619106
_cons	-.036551	.0116822	-3.13	0.002	-.0594874	-.0136146

```
730
731 reg Toughminded_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_dif
  > f c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include==1
```

Source	SS	df	MS		
Model	1.20497034	7	.17213862	Number of obs	721
				F(7, 713)	2.11
Residual	58.1900809	713	.081613017	Prob > F	0.0405
				R-squared	0.0203
				Adj R-squared	0.0107
Total	59.3950513	720	.082493127	Root MSE	.28568

Toughminded_d~f	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	.0002145	.0582912	0.00	0.997	-.1142284	.1146575
aggressive_diff	.1822095	.057533	3.17	0.002	.0692553	.2951638
sadness_diff	-.0252569	.0596021	-0.42	0.672	-.1422735	.0917598
selfconf_diff	.041708	.0601011	0.69	0.488	-.0762883	.1597043
ID_Ukraine_diff	.0638153	.0972792	0.66	0.512	-.1271726	.2548032
ID_Europe_diff	-.0183312	.0486229	-0.38	0.706	-.1137924	.0771299
ID_Russia_diff	.1002133	.0701895	1.43	0.154	-.0375895	.2380161
_cons	.0833837	.01177	7.08	0.000	.0602758	.1064917

```
732
733
734
735 ********************************************** SOM.7. ANALYSES USING PCA FACTOR SCOR
  > ES AS TRAIT VARIABLES ************************************************
736
737 *** SOM 7.a: Testing the Conflict-Sensitivity Hypothesis and effect of emotional rea
  > ctions using factor score variables for trait dimensions
738 ** Between-respondent analyses
739 reg Comp_PCA_1 i.Conflict_1
```

Source	SS	df	MS		
Model	3.5308171	1	3.5308171	Number of obs	988
				F(1, 986)	3.54
Residual	983.469185	986	.99743325	Prob > F	0.0602
				R-squared	0.0036
				Adj R-squared	0.0026
Total	987.000002	987	1	Root MSE	.99872

Comp_PCA_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**.1195807**	**.0635573**	**1.88**	**0.060**	**-.0051424**	**.2443039**
_cons	**-.0608797**	**.0453493**	**-1.34**	**0.180**	**-.149872**	**.0281127**

740 margins, dydx(Conflict_1) level(95)

Conditional marginal effects — Number of obs = **988**
Model VCE: **OLS**

Expression: **Linear prediction, predict()**
dy/dx wrt: **2.Conflict_1**

	dy/dx	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**.1195807**	**.0635573**	**1.88**	**0.060**	**-.0051424**	**.2443039**

Note: dy/dx for factor levels is the discrete change from the base level.

741 marginsplot, recastci(rcap) yscale(range(-.3(.1).3)) ylabel(-.3(.1).3) recast(scatt
> er) yline(0) ///
> xtitle("Peace") ytitle("Marg. Effect of Peace on Competence (PCA factor score)") tit
> le("Competence (PCA factor score)") scheme(s1mono) legend(off) name(Competence_PCA,
> replace)

Variables that uniquely identify margins:

742
743 reg Warm_PCA_1 i.Conflict_1

Source	SS	df	MS		
Model	**5.39956244**	**1**	**5.39956244**	Number of obs	= **988**
Residual	**981.600435**	**986**	**.995537967**	F(1, 986)	= **5.42**
				Prob > F	= **0.0201**
Total	**986.999998**	**987**	**.999999998**	R-squared	= **0.0055**
				Adj R-squared	= **0.0045**
				Root MSE	= **.99777**

Warm_PCA_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**.1478778**	**.0634969**	**2.33**	**0.020**	**.0232732**	**.2724824**
_cons	**-.075286**	**.0453062**	**-1.66**	**0.097**	**-.1641937**	**.0136218**

744 margins, dydx(Conflict_1) level(95)

Conditional marginal effects — Number of obs = **988**
Model VCE: **OLS**

Expression: **Linear prediction, predict()**
dy/dx wrt: **2.Conflict_1**

	dy/dx	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**.1478778**	**.0634969**	**2.33**	**0.020**	**.0232732**	**.2724824**

Note: dy/dx for factor levels is the discrete change from the base level.

```
745 marginsplot,  recastci(rcap) yscale(range(-.3(.1).3)) ylabel(-.3(.1).3) recast(scatt
  > er) yline(0) ///
  > xtitle("Peace")  ytitle("Marg. Effect of Peace on Warmth (PCA factor score)") title(
  > "Warmth (PCA factor score)") scheme(s1mono) legend(off)  name(Warmth_PCA, replace)

  Variables that uniquely identify margins:

746
747 reg Domi_PCA_1 i.Conflict_1
```

Source	SS	df	MS	Number of obs	=	**988**
				F(1, 986)	=	**5.89**
Model	**5.86408685**	**1**	**5.86408685**	Prob > F	=	**0.0154**
Residual	**981.135917**	**986**	**.995066853**	R-squared	=	**0.0059**
				Adj R-squared	=	**0.0049**
Total	**987.000004**	**987**	**1**	Root MSE	=	**.99753**

Domi_PCA_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**-.1541075**	**.0634819**	**-2.43**	**0.015**	**-.2786826**	**-.0295324**
_cons	**.0784576**	**.0452955**	**1.73**	**0.084**	**-.0104291**	**.1673443**

```
748 margins, dydx(Conflict_1) level(95)

  Conditional marginal effects                                    Number of obs = 988
  Model VCE: OLS

  Expression: Linear prediction, predict()
  dy/dx wrt:  2.Conflict_1
```

	dy/dx	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	**-.1541075**	**.0634819**	**-2.43**	**0.015**	**-.2786826**	**-.0295324**

```
  Note: dy/dx for factor levels is the discrete change from the base level.

749 marginsplot,  recastci(rcap) yscale(range(-.3(.1).3)) ylabel(-.3(.1).3) recast(scatt
  > er) yline(0) ///
  > xtitle("Peace")  ytitle("Marg. Effect of Peace on Dominance (PCA factor score)") tit
  > le("Dominance (PCA factor score)") scheme(s1mono) legend(off)  name(Dominance_PCA, r
  > eplace)

  Variables that uniquely identify margins:

750
751 * Creates common figure (not displayed in SOM.7)
752 graph combine Competence_PCA Warmth_PCA Dominance_PCA, scheme(s1mono) col(3) ysize(3
  > ) xsize(6)

753 graph export fig2_PCA.pdf, replace
  file fig2_PCA.pdf saved as PDF format

754
755 ** Within-respondent analyses: see below after dataset is re-shaped to long format
```

```
756
757
758 *** SOM. 7.b: Testing the role of emotional reactions to the war using factor score
  > trait variables
759 reg Comp_PCA_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff c
  > .ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include_PCA==1
```

Source	SS	df	MS		
Model	25.1549777	7	3.59356824	Number of obs	683
				F(7, 675)	4.23
Residual	573.196918	675	.84918062	Prob > F	0.0001
				R-squared	0.0420
				Adj R-squared	0.0321
Total	598.351896	682	.877348821	Root MSE	.92151

Comp_PCA_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.36725	.1938587	-1.89	0.059	-.7478886	.0133885
aggressive_diff	.4766669	.1914833	2.49	0.013	.1006924	.8526414
sadness_diff	.5060609	.2007945	2.52	0.012	.111804	.9003177
selfconf_diff	.3134092	.2031848	1.54	0.123	-.0855409	.7123594
ID_Ukraine_diff	.7697779	.328259	2.35	0.019	.1252464	1.414309
ID_Europe_diff	.1507477	.1640946	0.92	0.359	-.1714495	.4729449
ID_Russia_diff	.0235283	.2386548	0.10	0.921	-.4450667	.4921234
_cons	-.0873647	.0391463	-2.23	0.026	-.164228	-.0105015

```
760
761 reg Warm_PCA_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff c
  > .ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include_PCA==1
```

Source	SS	df	MS		
Model	19.6506272	7	2.80723245	Number of obs	683
				F(7, 675)	3.38
Residual	560.722134	675	.830699458	Prob > F	0.0015
				R-squared	0.0339
				Adj R-squared	0.0238
Total	580.372761	682	.850986453	Root MSE	.91143

Warm_PCA_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	-.2765789	.1917376	-1.44	0.150	-.6530526	.0998949
aggressive_diff	.2124068	.1893882	1.12	0.262	-.1594539	.5842675
sadness_diff	.5518197	.1985974	2.78	0.006	.1618766	.9417627
selfconf_diff	-.1436557	.2009616	-0.71	0.475	-.5382408	.2509293
ID_Ukraine_diff	.2350241	.3246673	0.72	0.469	-.4024552	.8725034
ID_Europe_diff	.5211665	.1622991	3.21	0.001	.2024947	.8398383
ID_Russia_diff	-.1254654	.2360435	-0.53	0.595	-.5889333	.3380024
_cons	-.0162903	.038718	-0.42	0.674	-.0923125	.0597319

```
762
763 reg Domi_PCA_diff c.fearfull_diff c.aggressive_diff c.sadness_diff c.selfconf_diff c
  > .ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include_PCA==1
```

Source	SS	df	MS		
Model	8.22526711	7	1.17503816	Number of obs	683
				F(7, 675)	1.58
Residual	502.920666	675	.745067653	Prob > F	0.1390
				R-squared	0.0161
				Adj R-squared	0.0059
Total	511.145933	682	.74948084	Root MSE	.86317

Domi_PCA_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
fearfull_diff	.0987235	.1815863	0.54	0.587	-.2578185	.4552654
aggressive_diff	.3819084	.1793613	2.13	0.034	.0297353	.7340816
sadness_diff	.0433761	.188083	0.23	0.818	-.325922	.4126741
selfconf_diff	.1620485	.190322	0.85	0.395	-.2116458	.5357428
ID_Ukraine_diff	.2895549	.3074783	0.94	0.347	-.3141741	.8932838
ID_Europe_diff	-.0105256	.1537064	-0.07	0.945	-.3123258	.2912746
ID_Russia_diff	.331105	.2235466	1.48	0.139	-.1078252	.7700353
_cons	.0707088	.0366681	1.93	0.054	-.0012885	.1427061

```
764
765
766 ****************************** SOM.8. ANALYSES OF RELATIVE IMPORTANCE USING ALL AVAI
  > LABLE RESPONDENTS IN WAVE 1 ********************************************************
  > *****
767 *** SOM 8: Testing the relative importance of leader competence, warmth and dominanc
  > e in Wave 1 using all available respondents
768 reg Comp_scale_1 if Conflict_1 == 1
```

Source	SS	df	MS	Number of obs	=	528
				F(0, 527)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	13.078971	527	.024817782	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	13.078971	527	.024817782	Root MSE	=	.15754

Comp_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.8824179	.0068559	128.71	0.000	.8689497	.8958862

```
769 reg Warm_scale_1 if Conflict_1 == 1
```

Source	SS	df	MS	Number of obs	=	525
				F(0, 524)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	29.2134109	524	.055750784	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	29.2134109	524	.055750784	Root MSE	=	.23612

Warm_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.6990476	.0103049	67.84	0.000	.6788035	.7192917

```
770 reg Domi_scale_1 if Conflict_1 == 1
```

Source	SS	df	MS	Number of obs	=	522
				F(0, 521)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	35.6961065	521	.0685146	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	35.6961065	521	.0685146	Root MSE	=	.26175

Domi_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.5288953	.0114566	46.17	0.000	.5063884	.5514021

```
771
772 reg Comp_scale_1 if Context== 1
```

Source	SS	df	MS	Number of obs	=	528
				F(0, 527)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	13.078971	527	.024817782	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	13.078971	527	.024817782	Root MSE	=	.15754

Comp_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.8824179	.0068559	128.71	0.000	.8689497	.8958862

```
773 margins, level(95)
```

Predictive margins Number of obs = **528**
Model VCE: **OLS**

Expression: **Linear prediction, predict()**

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_cons	.8824179	.0068559	128.71	0.000	.8689497	.8958862

```
774 marginsplot,  recastci(rcap) yscale(range(0(.1)1)) ylabel(0(.1)1) recast(scatter) yl
  > ine(0) ///
  > xtitle("Competence") ytitle("Competence Importance") title("Competence") legend(off)
  >  scheme(s1mono) name(Comp_war_mean_SOM4, replace)
```

Variables that uniquely identify margins:

```
775
776 reg Warm_scale_1 if Context== 1
```

Source	SS	df	MS	Number of obs	=	525
				F(0, 524)	=	0.00
Model	0	0	.	Prob > F	=	.
Residual	29.2134109	524	.055750784	R-squared	=	0.0000
				Adj R-squared	=	0.0000
Total	29.2134109	524	.055750784	Root MSE	=	.23612

Warm_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.6990476	.0103049	67.84	0.000	.6788035	.7192917

```
777 margins, level(95)
```

Predictive margins Number of obs = **525**
Model VCE: **OLS**

Expression: **Linear prediction, predict()**

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_cons	.6990476	.0103049	67.84	0.000	.6788035	.7192917

```
778 marginsplot,  recastci(rcap) yscale(range(0(.1)1)) ylabel(0(.1)1) recast(scatter) yl
  > ine(0) ///
  > xtitle("Warmth") ytitle("Warmth Importance") title("Warmth") legend(off) scheme(s1mo
  > no) name(Warm_war_mean_SOM4, replace)

  Variables that uniquely identify margins:

779
780 reg Domi_scale_1 if Context== 1
```

Source	SS	df	MS		
Model	**0**	**0**	**.**	Number of obs	= **522**
Residual	**35.6961065**	**521**	**.0685146**	F(0, 521)	= **0.00**
				Prob > F	= **.**
Total	**35.6961065**	**521**	**.0685146**	R-squared	= **0.0000**
				Adj R-squared	= **0.0000**
				Root MSE	= **.26175**

Domi_scale_1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	**.5288953**	**.0114566**	**46.17**	**0.000**	**.5063884**	**.5514021**

```
781 margins, level(95)

  Predictive margins                                      Number of obs = 522
  Model VCE: OLS

  Expression: Linear prediction, predict()
```

	Margin	Delta-method std. err.	t	P>|t|	[95% conf.	interval]
_cons	**.5288953**	**.0114566**	**46.17**	**0.000**	**.5063884**	**.5514021**

```
782 marginsplot,  recastci(rcap) yscale(range(0(.1)1)) ylabel(0(.1)1) recast(scatter) yl
  > ine(0) ///
  > xtitle("Dominance") ytitle("Dominance Importance") title("Dominance") legend(off) sc
  > heme(s1mono) name(Domi_war_mean_SOM4, replace)

  Variables that uniquely identify margins:

783
784 graph combine Comp_war_mean_SOM4 Warm_war_mean_SOM4 Domi_war_mean_SOM4, scheme(s1mon
  > o) cols(3)

785
786
787
788 ************************************************* SOM.9. ASSESSMENT OF POTENTIAL A
  > TTRITION ***********************************************************************
789 *** SOM 9: Assessing potential attrition bias
790 tab _merge
```

Matching result from merge	Freq.	Percent	Cum.
Master only (1)	**270**	**24.98**	**24.98**
Matched (3)	**811**	**75.02**	**100.00**
Total	**1,081**	**100.00**	

```
791 *[Overall, we reinterviewed 75.02% of Wave 1 sample]
792
793
794 *Generate dropout variable
795 tab _merge, nolab
```

Matching result from merge	Freq.	Percent	Cum.
1	**270**	**24.98**	**24.98**
3	**811**	**75.02**	**100.00**
Total	**1,081**	**100.00**	

```
796 gen dropout = .
  (1,081 missing values generated)

797 replace dropout = 1 if _merge == 1
  (270 real changes made)

798 replace dropout = 0 if _merge == 3
  (811 real changes made)

799 tab dropout
```

dropout	Freq.	Percent	Cum.
0	**811**	**75.02**	**75.02**
1	**270**	**24.98**	**100.00**
Total	**1,081**	**100.00**	

```
800
801 *** ATTRITION TESTS
802 *Lagged Comp_scale_1 (measured at Wave 1) alone
803 logit dropout Comp_scale_1

  Iteration 0:  Log likelihood = -591.90357
  Iteration 1:  Log likelihood = -583.42704
  Iteration 2:  Log likelihood = -583.26807
  Iteration 3:  Log likelihood = -583.26807

  Logistic regression                                     Number of obs =  1,057
                                                          LR chi2(1)    =  17.27
                                                          Prob > chi2   = 0.0000
  Log likelihood = -583.26807                             Pseudo R2     = 0.0146
```

dropout	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
Comp_scale_1	**-1.902788**	**.4557757**	**-4.17**	**0.000**	**-2.796092**	**-1.009484**
_cons	**.5738623**	**.4067128**	**1.41**	**0.158**	**-.2232802**	**1.371005**

```
804 *Lagged Comp_scale_1 together with controls (all measured at Wave 1)
805 logit dropout Comp_scale_1 c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_scale_1
  >  c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 c.w1_victim_self i.se
  > x c.age i.edu

  Iteration 0:  Log likelihood = -526.79547
  Iteration 1:  Log likelihood = -495.44566
  Iteration 2:  Log likelihood = -494.63897
  Iteration 3:  Log likelihood = -494.63861
  Iteration 4:  Log likelihood = -494.63861

  Logistic regression                                     Number of obs =    962
                                                          LR chi2(15)   =  64.31
                                                          Prob > chi2   = 0.0000
  Log likelihood = -494.63861                             Pseudo R2     = 0.0610
```

dropout	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
Comp_scale_1	-1.141371	.5375453	-2.12	0.034	-2.194941	-.0878021
fearfull_scale_1	.1528675	.4278174	0.36	0.721	-.685639	.9913742
aggressive_scale_1	-.7483885	.4069631	-1.84	0.066	-1.546021	.0492444
sadness_scale_1	-.1882361	.4891973	-0.38	0.700	-1.147045	.7705729
selfconf_scale_1	.7100792	.4367422	1.63	0.104	-.145919	1.566078
ID_Ukraine_1	-.262607	.5558235	-0.47	0.637	-1.352001	.826787
ID_Europe_1	-.4031576	.3053884	-1.32	0.187	-1.001708	.1953927
ID_Russia_1	1.266145	.3635673	3.48	0.000	.5535657	1.978723
w1_victim_self	.1160616	.0752109	1.54	0.123	-.031349	.2634722
sex						
Female	-.2925704	.1768579	-1.65	0.098	-.639282	.0540647
age	-.0288053	.009446	-3.05	0.002	-.047319	-.0102916
education						
Professional-technical (vocational)	-.2299669	.3183674	-0.72	0.470	-.853949	.3940218
Incomplete higher	-.3861523	.389221	-0.99	0.321	-1.149011	.3767069
Bachelor degree	-.1371803	.3162278	-0.43	0.664	-.756975	.4826149
Master degree & Doctorate	-.1555782	.2870328	-0.54	0.588	-.718136	.4069958
_cons	1.39074	.7394895	1.88	0.060	-.058113	2.840113

```
806
807 *Lagged Warm_scale_1 (measured at Wave 1) alone
808 logit dropout Warm_scale_1

  Iteration 0:  Log likelihood = -589.10812
  Iteration 1:  Log likelihood = -588.83995
  Iteration 2:  Log likelihood = -588.83989
  Iteration 3:  Log likelihood = -588.83989

  Logistic regression                                     Number of obs =  1,055
                                                          LR chi2(1)    =   0.54
                                                          Prob > chi2   = 0.4639
  Log likelihood = -588.83989                             Pseudo R2     = 0.0005

------------------------------------------------------------------------------
     dropout | Coefficient  Std. err.      z    P>|z|     [95% conf. interval]
-------------+----------------------------------------------------------------
Warm_scale_1 |  -.2299565   .3130999    -0.73   0.463    -.8436211    .3837082
       _cons |  -.9531062   .2342877    -4.07   0.000    -1.412302   -.4939108
------------------------------------------------------------------------------

809 *Lagged Warm_scale_1 together with controls (all measured at Wave 1)
810 logit dropout Warm_scale_1  c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_scale_
  > 1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 c.w1_victim_self i.s
  > ex c.age i.edu

  Iteration 0:  Log likelihood = -526.79547
  Iteration 1:  Log likelihood = -497.64132
  Iteration 2:  Log likelihood = -496.93375
  Iteration 3:  Log likelihood = -496.93346
  Iteration 4:  Log likelihood = -496.93346

  Logistic regression                                     Number of obs =    962
                                                          LR chi2(15)   =  59.72
                                                          Prob > chi2   = 0.0000
  Log likelihood = -496.93346                             Pseudo R2     = 0.0567

------------------------------------------------------------------------------
  ------------------
                     dropout | Coefficient  Std. err.      z    P>|z|     [95%
>  con
>     f. interval]
-------------+----------------------------------------------------------------
  ------------------
                Warm_scale_1 |  -.0960439   .3627443    -0.26   0.791    -.807
> 0098
>         .6149219
             fearfull_scale_1 |   .2124884   .4270631     0.50   0.619      -.6
> 2454
>         1.049517
           aggressive_scale_1 |  -.8242786    .403782    -2.04   0.041    -1.61
> 5677
>        -.0328803
            sadness_scale_1 |  -.2522384   .4867595    -0.52   0.604    -1.20
> 6269
>         .7017927
           selfconf_scale_1 |   .6110285   .4331884     1.41   0.158    -.238
> 0052
>         1.460062
              ID_Ukraine_1 |   -.393159   .5455725    -0.72   0.471    -1.46
> 2462
>         .6761435
               ID_Europe_1 |   -.459691   .3053844    -1.51   0.132    -1.05
> 8233
>         .1388514
               ID_Russia_1 |   1.236036   .3621214     3.41   0.001     .526
> 2912
>         1.945781
            w1_victim_self |   .1382234   .0744625     1.86   0.063    -.007
> 7204
>         .2841672
                           |
```

```
                                sex |
                             Female |  -.3324541   .1758482    -1.89   0.059    -.677
> 1102
>         .0122021
                                age |  -.0293568    .009575    -3.07   0.002    -.048
> 1234
>        -.0105902
                                    |
                          education |
Professional-technical (vocational) |  -.2805353   .3149069    -0.89   0.373    -.897
> 7415
>         .3366708
                  Incomplete higher |  -.4496327   .3855694    -1.17   0.244    -1.20
> 5335
>         .3060694
                    Bachelor degree |  -.2069804   .3125174    -0.66   0.508    -.819
> 5033
>         .4055425
           Master degree & Doctorate |  -.2341197   .2824316    -0.83   0.407    -.787
> 6754
>         .3194359
                                    |
                              _cons |   .7929506   .6993965     1.13   0.257    -.577
> 8414
>         2.163743
------------------------------------+----------------------------------------------
```

```
811
812 *Lagged Warm_scale_1 (measured at Wave 1) alone
813 logit dropout Domi_scale_1

Iteration 0:  Log likelihood = -586.85865
Iteration 1:  Log likelihood = -586.69847
Iteration 2:  Log likelihood = -586.69846

Logistic regression                                     Number of obs =  1,051
                                                        LR chi2(1)    =   0.32
                                                        Prob > chi2   = 0.5714
Log likelihood = -586.69846                             Pseudo R2     = 0.0003

------------------------------------------------------------------------------
     dropout | Coefficient  Std. err.      z    P>|z|     [95% conf. interval]
-------------+----------------------------------------------------------------
Domi_scale_1 |   .1566942   .2769298     0.57   0.572    -.3860784    .6994667
       _cons |  -1.198385   .1601652    -7.48   0.000    -1.512303   -.8844674
------------------------------------------------------------------------------

814 *Lagged Domi_scale_1 together with controls (all measured at Wave 1)
815 logit dropout Domi_scale_1  c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_scale_
  > 1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 c.w1_victim_self i.s
  > ex c.age i.edu

Iteration 0:  Log likelihood = -526.79547
Iteration 1:  Log likelihood = -496.76019
Iteration 2:  Log likelihood =  -496.0155
Iteration 3:  Log likelihood = -496.01512
Iteration 4:  Log likelihood = -496.01512

Logistic regression                                     Number of obs =    962
                                                        LR chi2(15)   =  61.56
                                                        Prob > chi2   = 0.0000
Log likelihood = -496.01512                             Pseudo R2     = 0.0584
```

dropout	Coefficient	Std. err.	z	P>|z|	[95% conf.	interval]
Domi_scale_1	.4529705	.3140765	1.44	0.149	-.162	1.068549
fearfull_scale_1	.1761816	.4262723	0.41	0.679	-.659	1.01166
aggressive_scale_1	-.8707035	.403767	-2.16	0.031	-1.66	-.0793346
sadness_scale_1	-.2732188	.4868099	-0.56	0.575	-1.22	.6809111
selfconf_scale_1	.5542374	.4342595	1.28	0.202	-.296	1.40537
ID_Ukraine_1	-.4211515	.5453761	-0.77	0.440	-1.49	.6477661
ID_Europe_1	-.4533843	.3042895	-1.49	0.136	-1.04	.1430122
ID_Russia_1	1.187188	.3623684	3.28	0.001	.476	1.897417
w1_victim_self	.133141	.0745468	1.79	0.074	-.012	.2792501
sex						
Female	-.2984131	.1777495	-1.68	0.093	-.646	.0499696
age	-.0302317	.0094769	-3.19	0.001	-.04	-.0116573
education						
Professional-technical (vocational)	-.2986024	.3144013	-0.95	0.342	-.914	.3176127
Incomplete higher	-.4363078	.3852685	-1.13	0.257	-1.1	.3188045
Bachelor degree	-.2050183	.3121747	-0.66	0.511	-.816	.4068329
Master degree & Doctorate	-.2452661	.2819104	-0.87	0.384	-.797	.3072682
_cons	.6357274	.6728737	0.94	0.345	-.683	1.954536

```
816
817
818 ************************************************ SOM.10. ALTERNATIVE 'RALLY AROUN
  > D THE FLAG' EXPLANATION ************************************************************
  > ********
819 *** SOM 10: Explores traits ratings of President Zelenskyy
820 * Do average ratings of Zelensky closely mirror stated trait preferences in ideal le
  > ader?
821 summ Comp_scale_Zel1 Warm_scale_Zel1 Domi_scale_Zel1
```

Variable	Obs	Mean	Std. dev.	Min	Max
Comp_scal~l1	1,063	.7637713	.2908337	0	1
Warm_scal~l1	1,054	.7145003	.2931831	0	1
Domi_scal~l1	1,052	.4771863	.2808222	0	1

```
822
823 reg Comp_scale_Zel1 if Conflict_1 == 1
```

Source	SS	df	MS		
				Number of obs	= 531
				F(0, 530)	= 0.00
Model	0	0	.	Prob > F	= .
Residual	46.2469941	530	.087258479	R-squared	= 0.0000
				Adj R-squared	= 0.0000
Total	46.2469941	530	.087258479	Root MSE	= .2954

Comp_scal~l1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.769094	.0128191	60.00	0.000	.7439115	.7942764

```
824 reg Warm_scale_Zel1 if Conflict_1 == 1
```

Source	SS	df	MS		
				Number of obs	= 526
				F(0, 525)	= 0.00
Model	0	0	.	Prob > F	= .
Residual	47.2798888	525	.090056931	R-squared	= 0.0000
				Adj R-squared	= 0.0000
Total	47.2798888	525	.090056931	Root MSE	= .30009

Warm_scal~l1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.7148289	.0130848	54.63	0.000	.689124	.7405338

```
825 reg Domi_scale_Zel1 if Conflict_1 == 1
```

Source	SS	df	MS		
				Number of obs	= 526
				F(0, 525)	= 0.00
Model	0	0	.	Prob > F	= .
Residual	45.1691083	525	.086036397	R-squared	= 0.0000
				Adj R-squared	= 0.0000
Total	45.1691083	525	.086036397	Root MSE	= .29332

Domi_scal~l1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
_cons	.4786122	.0127893	37.42	0.000	.4534876	.5037367

```
826 // Ranking of traits is similar as for ideal leader. But rating of Z's competence is
  >  0.15 scale points lower than for ideal leader.
827
828 * Does context (war vs. peace) affect ratings of Zalenskyy
829 reg Comp_scale_Zel1 i.Conflict_1
```

Source	SS	df	MS		
Model	.030058886	1	.030058886	Number of obs	1,063
				F(1, 1061)	0.36
Residual	89.7984354	1,061	.08463566	Prob > F	0.5513
				R-squared	0.0003
				Adj R-squared	-0.0006
Total	89.8284943	1,062	.08458427	Root MSE	.29092

Comp_scale_Z~1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	-.0106353	.017846	-0.60	0.551	-.0456527	.0243821
_cons	.769094	.0126249	60.92	0.000	.7443213	.7938666

```
830 reg Warm_scale_Zel1 i.Conflict_1
```

Source	SS	df	MS		
Model	.000113364	1	.000113364	Number of obs	1,054
				F(1, 1052)	0.00
Residual	90.511882	1,052	.086037911	Prob > F	0.9711
				R-squared	0.0000
				Adj R-squared	-0.0009
Total	90.5119954	1,053	.085956311	Root MSE	.29332

Warm_scale_Z~1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	-.0006559	.0180699	-0.04	0.971	-.036113	.0348012
_cons	.7148289	.0127895	55.89	0.000	.6897332	.7399246

```
831 reg Domi_scale_Zel1 i.Conflict_1
```

Source	SS	df	MS		
Model	.00213878	1	.00213878	Number of obs	1,052
				F(1, 1050)	0.03
Residual	82.8808866	1,050	.078934178	Prob > F	0.8693
				R-squared	0.0000
				Adj R-squared	-0.0009
Total	82.8830254	1,051	.078861109	Root MSE	.28095

Domi_scale_Z~1	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Conflict_1						
Peace, future	-.0028517	.0173243	-0.16	0.869	-.0368458	.0311424
_cons	.4786122	.0122501	39.07	0.000	.4545747	.5026496

```
832 // Assigned context does not affect ratings of Zelensky
833
834 * Do Ratings of Zelensky correlate with stated trait preferences in ideal leader? NO
  > T USED IN SOM.6
835 pwcorr Comp_scale_1 Comp_scale_Zel1, sig // r=0.113
```

	Com~le_1	Comp_~l1
Comp_scale_1	1.0000	
Comp_scal~l1	0.1130	1.0000
	0.0002	

```
836 pwcorr Warm_scale_1 Warm_scale_Zel1, sig // r=0.398
```

	Warm~e_1	Warm_~l1
Warm_scale_1	**1.0000**	
Warm_scal~l1	**0.3984** **0.0000**	**1.0000**

```
837 pwcorr Domi_scale_1 Domi_scale_Zel1, sig // r=0.443
```

	Domi~e_1	Domi_~l1
Domi_scale_1	**1.0000**	
Domi_scal~l1	**0.4431** **0.0000**	**1.0000**

```
838 // Ratings of Z and ideal leader correlate, but not overwhelmingly so
839
840
841 ************************************************* SOM.11. INDIVIDUAL VICTIMIZATION
  >  OF RUSSIAN ATTACKS ************************************************************
  > ****
842
843 *** SOM.11: Exploring if self-reported victimization by Russian attacks affect leade
  > r trait preferences
844 * Models controlling for changes in identification variables (models 1-3 in SOM.11)
845 reg Comp_scale_diff c.Victimization_diff c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Rus
  > sia_diff if include==1
```

Source	SS	df	MS		
Model	**.283610561**	**4**	**.07090264**	Number of obs	= **711**
Residual	**11.8871864**	**706**	**.016837375**	F(4, 706)	= **4.21**
				Prob > F	= **0.0022**
				R-squared	= **0.0233**
				Adj R-squared	= **0.0178**
Total	**12.170797**	**710**	**.017141968**	Root MSE	= **.12976**

Comp_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Victimization_diff	**-.0475526**	**.0189375**	**-2.51**	**0.012**	**-.0847331**	**-.010372**
ID_Ukraine_diff	**.1021717**	**.0435817**	**2.34**	**0.019**	**.0166065**	**.1877369**
ID_Europe_diff	**.0231535**	**.02197**	**1.05**	**0.292**	**-.0199809**	**.0662879**
ID_Russia_diff	**-.0488814**	**.0311958**	**-1.57**	**0.118**	**-.1101291**	**.0123663**
_cons	**.0003546**	**.005162**	**0.07**	**0.945**	**-.0097801**	**.0104893**

```
846 reg Warm_scale_diff c.Victimization_diff c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Rus
  > sia_diff if include==1
```

Source	SS	df	MS		
Model	**.876387947**	**4**	**.219096987**	Number of obs	= **711**
Residual	**33.4391868**	**706**	**.047364287**	F(4, 706)	= **4.63**
				Prob > F	= **0.0011**
				R-squared	= **0.0255**
				Adj R-squared	= **0.0200**
Total	**34.3155747**	**710**	**.048331795**	Root MSE	= **.21763**

Warm_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Victimization_diff	**-.032306**	**.0317622**	**-1.02**	**0.309**	**-.0946657**	**.0300537**
ID_Ukraine_diff	**.0335988**	**.0730958**	**0.46**	**0.646**	**-.1099123**	**.1771099**
ID_Europe_diff	**.1496797**	**.0368484**	**4.06**	**0.000**	**.0773341**	**.2220253**
ID_Russia_diff	**-.0635483**	**.052322**	**-1.21**	**0.225**	**-.1662737**	**.0391771**
_cons	**-.0451014**	**.0086578**	**-5.21**	**0.000**	**-.0620995**	**-.0281034**

```
847 reg Domi_scale_diff c.Victimization_diff c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Rus
  > sia_diff if include==1
```

Source	SS	df	MS	Number of obs	=	711
				F(4, 706)	=	0.54
Model	.109844408	4	.027461102	Prob > F	=	0.7092
Residual	36.1560557	706	.051212543	R-squared	=	0.0030
				Adj R-squared	=	-0.0026
Total	36.2659001	710	.051078733	Root MSE	=	.2263

Domi_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Victimization_diff	-.0260402	.0330273	-0.79	0.431	-.0908838	.0388033
ID_Ukraine_diff	.0590337	.0760072	0.78	0.438	-.0901936	.2082609
ID_Europe_diff	-.0209786	.0383161	-0.55	0.584	-.0962058	.0542486
ID_Russia_diff	.0524634	.0544061	0.96	0.335	-.0543537	.1592804
_cons	.0229764	.0090026	2.55	0.011	.0053013	.0406514

```
848
849 * Models also controlling for changes in emotional reactions (models 4-6 in SOM.11)
850 reg Comp_scale_diff c.Victimization_diff c.fearfull_diff c.aggressive_diff c.sadness
  > _diff c.selfconf_diff c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include
  > ==1
```

Source	SS	df	MS	Number of obs	=	706
				F(8, 697)	=	4.24
Model	.563055925	8	.070381991	Prob > F	=	0.0001
Residual	11.5636544	697	.016590609	R-squared	=	0.0464
				Adj R-squared	=	0.0355
Total	12.1267104	705	.017201008	Root MSE	=	.1288

Comp_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Victimization_diff	-.0506781	.0189094	-2.68	0.008	-.0878044	-.0135519
fearfull_diff	-.0563143	.0266272	-2.11	0.035	-.1085934	-.0040352
aggressive_diff	.0679219	.0261436	2.60	0.010	.0165922	.1192516
sadness_diff	.0437024	.0271144	1.61	0.107	-.0095332	.0969381
selfconf_diff	.0396815	.0277011	1.43	0.152	-.0147061	.0940691
ID_Ukraine_diff	.086504	.0438741	1.97	0.049	.0003627	.1726453
ID_Europe_diff	.0156595	.0219826	0.71	0.476	-.0275005	.0588196
ID_Russia_diff	-.0424028	.0312252	-1.36	0.175	-.1037095	.018904
_cons	-.0035746	.0053924	-0.66	0.508	-.014162	.0070127

```
851 reg Warm_scale_diff c.Victimization_diff c.fearfull_diff c.aggressive_diff c.sadness
  > _diff c.selfconf_diff c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include
  > ==1
```

Source	SS	df	MS	Number of obs	=	706
				F(8, 697)	=	3.83
Model	1.44162484	8	.180203105	Prob > F	=	0.0002
Residual	32.8261573	697	.047096352	R-squared	=	0.0421
				Adj R-squared	=	0.0311
Total	34.2677822	705	.048606783	Root MSE	=	.21702

Warm_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Victimization_diff	-.0334488	.0318596	-1.05	0.294	-.0960011	.0291036
fearfull_diff	-.077192	.0448629	-1.72	0.086	-.1652747	.0108907
aggressive_diff	.0904302	.0440482	2.05	0.040	.0039471	.1769133
sadness_diff	.1093213	.0456838	2.39	0.017	.019627	.1990156
selfconf_diff	-.048069	.0466723	-1.03	0.303	-.1397042	.0435661
ID_Ukraine_diff	.0310155	.0739215	0.42	0.675	-.1141201	.1761511
ID_Europe_diff	.1389846	.0370375	3.75	0.000	.0662662	.2117029
ID_Russia_diff	-.0566594	.0526099	-1.08	0.282	-.1599523	.0466336
_cons	-.050552	.0090855	-5.56	0.000	-.0683902	-.0327138

```
852 reg Domi_scale_diff c.Victimization_diff c.fearfull_diff c.aggressive_diff c.sadness
  > _diff c.selfconf_diff c.ID_Ukraine_diff c.ID_Europe_diff c.ID_Russia_diff if include
  > ==1
```

Source	SS	df	MS	Number of obs	=	706
				F(8, 697)	=	1.02
Model	.410676465	8	.051334558	Prob > F	=	0.4212
Residual	35.1710921	697	.050460677	R-squared	=	0.0115
				Adj R-squared	=	0.0002
Total	35.5817685	705	.050470594	Root MSE	=	.22463

Domi_scale_diff	Coefficient	Std. err.	t	P>|t|	[95% conf.	interval]
Victimization_diff	-.0334112	.0329779	-1.01	0.311	-.0981592	.0313368
fearfull_diff	.0123753	.0464377	0.27	0.790	-.0787993	.1035498
aggressive_diff	.097824	.0455944	2.15	0.032	.0083052	.1873428
sadness_diff	.0277847	.0472873	0.59	0.557	-.0650581	.1206274
selfconf_diff	.0072708	.0483106	0.15	0.880	-.0875809	.1021225
ID_Ukraine_diff	.0475118	.0765163	0.62	0.535	-.1027182	.1977419
ID_Europe_diff	-.0275922	.0383375	-0.72	0.472	-.1028631	.0476786
ID_Russia_diff	.0520365	.0544566	0.96	0.340	-.0548822	.1589552
_cons	.0233883	.0094044	2.49	0.013	.0049239	.0418526

```
853
854
855
856 ************************************************** SOM.12. OBLAST-LEVEL ANALYSES OF
  >  RUSSIAN ATTACKS ********************************************************************
  > *
857
858 *** SOM.12: Exploring if oblast-level differences in attack intensity relates to tra
  > it preferences in leaders
859 ** Adding VIINA events to the dataset
860 * Wave 1: N of events per oblast
861 tab w1_q5
```

5.1 Oblast	Freq.	Percent	Cum.
Vinnytsia	25	2.31	2.31
Volyn	10	0.93	3.24
Dnipropetrovsk	122	11.29	14.52
Donetsk	33	3.05	17.58
Zhytomyr	27	2.50	20.07
Transcarpathian	15	1.39	21.46
Zaporizhzhia	62	5.74	27.20
Ivano-Frankivsk	28	2.59	29.79
Kyiv	37	3.42	33.21
Kyiv	195	18.04	51.25
Kirovohrad	21	1.94	53.19
Luhansk	10	0.93	54.12
Lviv	62	5.74	59.85
Mykolaiv	39	3.61	63.46
Odesa	71	6.57	70.03
Poltava	40	3.70	73.73
Rivne	21	1.94	75.67
Sumy	25	2.31	77.98
Ternopil	18	1.67	79.65
Kharkiv	94	8.70	88.34
Kherson	21	1.94	90.29
Khmelnytsky	28	2.59	92.88
Cherkasy	37	3.42	96.30
Chernivtsi	17	1.57	97.87
Chernihiv	23	2.13	100.00
Total	1,081	100.00	

862 tab w1_q5, nolab

5.1 Oblast	Freq.	Percent	Cum.
2	**25**	**2.31**	**2.31**
3	**10**	**0.93**	**3.24**
4	**122**	**11.29**	**14.52**
5	**33**	**3.05**	**17.58**
6	**27**	**2.50**	**20.07**
7	**15**	**1.39**	**21.46**
8	**62**	**5.74**	**27.20**
9	**28**	**2.59**	**29.79**
10	**37**	**3.42**	**33.21**
11	**195**	**18.04**	**51.25**
12	**21**	**1.94**	**53.19**
13	**10**	**0.93**	**54.12**
14	**62**	**5.74**	**59.85**
15	**39**	**3.61**	**63.46**
16	**71**	**6.57**	**70.03**
17	**40**	**3.70**	**73.73**
18	**21**	**1.94**	**75.67**
19	**25**	**2.31**	**77.98**
20	**18**	**1.67**	**79.65**
21	**94**	**8.70**	**88.34**
22	**21**	**1.94**	**90.29**
23	**28**	**2.59**	**92.88**
24	**37**	**3.42**	**96.30**
25	**17**	**1.57**	**97.87**
26	**23**	**2.13**	**100.00**
Total	**1,081**	**100.00**	

863
864 ** N of VIINA events (all types) in the 2-week period before Wave 1
865 * 2 Vinnytsya 279
866 * 3 Volyn 120
867 * 4 Dnipropetrovsk 279
868 * 5 Donetsk 2101
869 * 6 Zhytomyr 454
870 * 7 Transcarpathian 76
871 * 8 Zaporizhzhia 1280
872 * 9 Ivano-Frankivsk 51
873 * 10 Kyiv 1995
874 * 11 Kyiv city 3704
875 * 12 Kirovohrad 40
876 * 13 Luhansk 720
877 * 14 Lviv 393
878 * 15 Mykolayiv 805
879 * 16 Odessa 695
880 * 17 Poltava 126
881 * 18 Rivne 90
882 * 19 Sumy 1091
883 * 20 Ternopil 57
884 * 21 Kharkiv 2436
885 * 22 Kherson 938
886 * 23 Khmelnytsky 121
887 * 24 Cherkasy 139
888 * 25 Chernivtsi 29
889 * 26 Chernihiv 817
890 * Crimea 387 [no respondents reached in Crimea]

```
891 * Sevastopol 21 [no respondents reached in Sevastapol]
892
893 * Enters VIINA events/observations for Wave 1 to dataset
894 gen w1_VIINA_events = .
  (1,081 missing values generated)

895 replace w1_VIINA_events = 279 if w1_q5 == 2
  (25 real changes made)

896 replace w1_VIINA_events = 120 if w1_q5 == 3
  (10 real changes made)

897 replace w1_VIINA_events = 279 if w1_q5 == 4
  (122 real changes made)

898 replace w1_VIINA_events = 2101 if w1_q5 == 5
  (33 real changes made)

899 replace w1_VIINA_events = 454 if w1_q5 == 6
  (27 real changes made)

900 replace w1_VIINA_events = 76 if w1_q5 == 7
  (15 real changes made)

901 replace w1_VIINA_events = 1280 if w1_q5 == 8
  (62 real changes made)

902 replace w1_VIINA_events = 51 if w1_q5 == 9
  (28 real changes made)

903 replace w1_VIINA_events = 1995 if w1_q5 == 10
  (37 real changes made)

904 replace w1_VIINA_events = 3704 if w1_q5 == 11
  (195 real changes made)

905 replace w1_VIINA_events = 40 if w1_q5 == 12
  (21 real changes made)

906 replace w1_VIINA_events = 720 if w1_q5 == 13
  (10 real changes made)

907 replace w1_VIINA_events = 393 if w1_q5 == 14
  (62 real changes made)

908 replace w1_VIINA_events = 805 if w1_q5 == 15
  (39 real changes made)

909 replace w1_VIINA_events = 695 if w1_q5 == 16
  (71 real changes made)

910 replace w1_VIINA_events = 126 if w1_q5 == 17
  (40 real changes made)

911 replace w1_VIINA_events = 90 if w1_q5 == 18
  (21 real changes made)

912 replace w1_VIINA_events = 1091 if w1_q5 == 19
  (25 real changes made)
```

```
. replace w1_VIINA_events = 57 if w1_q5 == 20
  (18 real changes made)

. replace w1_VIINA_events = 2436 if w1_q5 == 21
  (94 real changes made)

. replace w1_VIINA_events = 938 if w1_q5 == 22
  (21 real changes made)

. replace w1_VIINA_events = 121 if w1_q5 == 23
  (28 real changes made)

. replace w1_VIINA_events = 139 if w1_q5 == 24
  (37 real changes made)

. replace w1_VIINA_events = 29 if w1_q5 == 25
  (17 real changes made)

. replace w1_VIINA_events = 817 if w1_q5 == 26
  (23 real changes made)

.
. ** Normalizes VIINA variable for Wave 1
. summ w1_VIINA_events
```

Variable	Obs	Mean	Std. dev.	Min	Max
w1_VIINA_e~s	**1,081**	**1319.797**	**1327.05**	**29**	**3704**

```
. gen VIINA_W1_norm = (w1_VIINA_events - r(min)) / (r(max) - r(min))

. summ VIINA_W1_norm
```

Variable	Obs	Mean	Std. dev.	Min	Max
VIINA_W1_n~m	**1,081**	**.3512374**	**.3611019**	**0**	**1**

```
.
. ** Creates log-transformed version of the normalized VIINA variable for Wave 1
. gen ln_VIINA_W1_norm = ln(VIINA_W1_norm)
  (17 missing values generated)

.
.
.
. ** N of VIINA events (all types) in the 2-week period before Wave 2
. * 2 Vinnytsya 72
. * 3 Volyn       150
. * 4 Dnipropetrovsk 349
. * 5 Donetsk 2274
. * 6 Zhytomyr 180
. * 7 Transcarpathian 117
. * 8 Zaporizhzhia 763
. * 9 Ivano-Frankivsk 37
. * 10 Kyiv 1435
. * 11 Kyiv city 1995
. * 12 Kirovohrad 20
. * 13 Luhansk 840
. * 14 Lviv 429
. * 15 Mykolayiv 415
```

```
946 * 16 Odessa 486
947 * 17 Poltava 86
948 * 18 Rivne 120
949 * 19 Sumy 465
950 * 20 Ternopil 27
951 * 21 Kharkiv 1258
952 * 22 Kherson 709
953 * 23 Khmelnytsky 82
954 * 24 Cherkasy 32
955 * 25 Chernivtsi 43
956 * 26 Chernihiv 680
957 * Crimea 335 [no respondents reached in Crimea]
958 * Sevastopol 41 [no respondents reached in Sevastapol]
959
960 * Enters VIINA events/observations for Wave 2 to dataset
961 gen w2_VIINA_events = .
  (1,081 missing values generated)

962 replace w2_VIINA_events = 72 if w2_q4 == 2
  (22 real changes made)

963 replace w2_VIINA_events = 150 if w2_q4 == 3
  (6 real changes made)

964 replace w2_VIINA_events = 349 if w2_q4 == 4
  (95 real changes made)

965 replace w2_VIINA_events = 2274 if w2_q4 == 5
  (19 real changes made)

966 replace w2_VIINA_events = 180 if w2_q4 == 6
  (20 real changes made)

967 replace w2_VIINA_events = 117 if w2_q4 == 7
  (10 real changes made)

968 replace w2_VIINA_events = 763 if w2_q4 == 8
  (48 real changes made)

969 replace w2_VIINA_events = 37 if w2_q4 == 9
  (32 real changes made)

970 replace w2_VIINA_events = 1435 if w2_q4 == 10
  (31 real changes made)

971 replace w2_VIINA_events = 1995 if w2_q4 == 11
  (110 real changes made)

972 replace w2_VIINA_events = 20 if w2_q4 == 12
  (18 real changes made)

973 replace w2_VIINA_events = 840 if w2_q4 == 13
  (5 real changes made)

974 replace w2_VIINA_events = 429 if w2_q4 == 14
  (60 real changes made)

975 replace w2_VIINA_events = 415 if w2_q4 == 15
  (22 real changes made)
```

```
976 replace w2_VIINA_events = 486 if w2_q4 == 16
  (50 real changes made)

977 replace w2_VIINA_events = 86 if w2_q4 == 17
  (42 real changes made)

978 replace w2_VIINA_events = 120 if w2_q4 == 18
  (20 real changes made)

979 replace w2_VIINA_events = 465 if w2_q4 == 19
  (21 real changes made)

980 replace w2_VIINA_events = 27 if w2_q4 == 20
  (17 real changes made)

981 replace w2_VIINA_events = 1258 if w2_q4 == 21
  (36 real changes made)

982 replace w2_VIINA_events = 709 if w2_q4 == 22
  (18 real changes made)

983 replace w2_VIINA_events = 82 if w2_q4 == 23
  (16 real changes made)

984 replace w2_VIINA_events = 32 if w2_q4 == 24
  (32 real changes made)

985 replace w2_VIINA_events = 43 if w2_q4 == 25
  (13 real changes made)

986 replace w2_VIINA_events = 680 if w2_q4 == 26
  (20 real changes made)

987
988 ** Normalizes VIINA variable for Wave 1
989 summ w2_VIINA_events
```

Variable	Obs	Mean	Std. dev.	Min	Max
w2_VIINA_e~s	**783**	**689.4879**	**696.1808**	**20**	**2274**

```
990 gen VIINA_W2_norm = (w2_VIINA_events - r(min)) / (r(max) - r(min))
  (298 missing values generated)

991 summ VIINA_W2_norm
```

Variable	Obs	Mean	Std. dev.	Min	Max
VIINA_W2_n~m	**783**	**.2970221**	**.3088646**	**0**	**1**

```
992
993 ** Creates log-transformed version of the normalized VIINA variable for Wave 1
994 gen ln_VIINA_W2_norm = ln(VIINA_W2_norm)
  (316 missing values generated)

995
996
997
998
999 *** Predictions of leader trait preferences from VIINA events (standard errors clust
  > ered at oblast-level)
```

```
1000** Wave 1 (Table SOM.12a)
1001reg Comp_scale_1 VIINA_W1_norm c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_sca
  > le_1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 i.sex c.age i.edu
  > cation, cluster(w1_q5)

  Linear regression                               Number of obs     =      1,012
                                                  F(14, 24)         =      59.33
                                                  Prob > F          =     0.0000
                                                  R-squared         =     0.1296
                                                  Root MSE          =     .13651

                                                          (Std. err. adjusted for 25 cl
  > usters in w1_q5)
  ------------------------------------------------------------------------------------
    ----------------
                                          |               Robust
                         Comp_scale_1 | Coefficient  std. err.      t    P>|t|     [95%
  >  con
  >     f. interval]
  ------------------------------------------------------------------------------------
    ----------------
                        VIINA_W1_norm |  -.0096757   .0081714    -1.18   0.248    -.026
  > 5407
  >         .0071893
                     fearfull_scale_1 |  -.0425799   .0282878    -1.51   0.145    -.100
  > 9631
  >         .0158032
                   aggressive_scale_1 |    .079017   .0246296     3.21   0.004      .02
  > 8184
  >         .1298499
                      sadness_scale_1 |   .0551254   .0269799     2.04   0.052    -.000
  > 5583
  >         .1108091
                     selfconf_scale_1 |   .0697216   .0291557     2.39   0.025     .009
  > 5472
  >          .129896
                         ID_Ukraine_1 |    .131721   .0479649     2.75   0.011     .032
  > 7264
  >         .2307156
                          ID_Europe_1 |   .0437404   .0244811     1.79   0.087    -.006
  > 7862
  >          .094267
                          ID_Russia_1 |  -.0061298   .0325483    -0.19   0.852    -.073
  > 3062
  >         .0610466
                                      |
                                  sex |
                               Female |   .0224803   .0081674     2.75   0.011     .005
  > 6236
  >         .0393371
                                  age |  -.0002181   .0005233    -0.42   0.681    -.001
  > 2982
  >          .000862
                                      |
                            education |
  Professional-technical (vocational) |    .027332   .0200714     1.36   0.186    -.014
  > 0933
  >         .0687572
                   Incomplete higher  |   .0418732   .0203612     2.06   0.051    -.000
  > 1503
  >         .0838968
                     Bachelor degree  |   .0420831   .0230371     1.83   0.080    -.005
  > 4631
  >         .0896292
            Master degree & Doctorate |   .0562099   .0158051     3.56   0.002     .023
  > 5898
  >           .08883
                                      |
                                _cons |   .5962104   .0614431     9.70   0.000      .46
  > 9398
  >         .7230228
  ------------------------------------------------------------------------------------
```

```
1002reg Warm_scale_1 VIINA_W1_norm c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_sca
  > le_1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 i.sex c.age i.edu
  > cation, cluster(w1_q5)

Linear regression                               Number of obs     =      1,010
                                                F(14, 24)         =      38.15
                                                Prob > F          =     0.0000
                                                R-squared         =     0.1123
                                                Root MSE          =     .21519

                                                               (Std. err. adjusted for 25 cl
  > usters in w1_q5)
------------------------------------------------------------------------------
             |               Robust
  Warm_scale_1 | Coefficient  std. err.      t    P>|t|     [95%
  >  con
  >     f. interval]
-------------+----------------------------------------------------------------
  VIINA_W1_norm |  -.0184231   .0128877    -1.43   0.166     -.04
  > 5022
  >         .0081758
  fearfull_scale_1 |   .0990457   .0441623     2.24   0.034     .007
  > 8993
  >         .1901921
  aggressive_scale_1 |  -.0043112   .0432222    -0.10   0.921    -.093
  > 5176
  >         .0848951
  sadness_scale_1 |   .0991296   .0651628     1.52   0.141    -.035
  > 3598
  >          .233619
  selfconf_scale_1 |   .1052093   .0368444     2.86   0.009     .029
  > 1663
  >         .1812523
  ID_Ukraine_1 |   .1046643   .0707795     1.48   0.152    -.041
  > 4174
  >         .2507461
  ID_Europe_1 |   .0646321   .0319918     2.02   0.055    -.001
  > 3956
  >         .1306598
  ID_Russia_1 |    .005405    .033667     0.16   0.874    -.064
  > 0803
  >         .0748904
              |
  sex |
  Female |   .0171565   .0152081     1.13   0.270    -.014
  > 2314
  >         .0485445
  age |  -.0048571    .000626    -7.76   0.000    -.006
  > 1491
  >        -.0035651
              |
  education |
  Professional-technical (vocational) |  -.0027496    .025919    -0.11   0.916    -.056
  > 2438
  >         .0507445
  Incomplete higher |   -.018459   .0332708    -0.55   0.584    -.087
  > 1266
  >         .0502087
  Bachelor degree |  -.0574968   .0311708    -1.84   0.077    -.121
  > 8302
  >         .0068367
  Master degree & Doctorate |  -.0532765   .0254322    -2.09   0.047    -.105
  > 7659
  >         -.000787
              |
  _cons |   .6220129   .0607372    10.24   0.000     .496
  > 6573
  >         .7473684
```

```
1003reg Domi_scale_1 VIINA_W1_norm c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_sca
  > le_1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 i.sex c.age i.edu
  > cation, cluster(w1_q5)

Linear regression                               Number of obs     =      1,009
                                                F(14, 24)         =      16.84
                                                Prob > F          =     0.0000
                                                R-squared         =     0.0617
                                                Root MSE          =     .25267

                                                               (Std. err. adjusted for 25 cl
  > usters in w1_q5)
------------------------------------------------------------------------------
             |               Robust
  Domi_scale_1 | Coefficient  std. err.      t    P>|t|     [95%
  >  con
  >     f. interval]
-------------+----------------------------------------------------------------
  VIINA_W1_norm |  -.0402314   .0276414    -1.46   0.158    -.097
  > 2804
  >         .0168176
  fearfull_scale_1 |   .0562479   .0502032     1.12   0.274    -.047
  > 3664
  >         .1598623
  aggressive_scale_1 |   .0678551   .0365506     1.86   0.076    -.007
  > 5817
  >         .1432918
  sadness_scale_1 |   .0546184   .0648623     0.84   0.408    -.079
  > 2509
  >         .1884876
  selfconf_scale_1 |   .1348081   .0453993     2.97   0.007     .041
  > 1085
  >         .2285077
  ID_Ukraine_1 |   .0077534   .0442257     0.18   0.862     -.08
  > 3524
  >         .0990309
  ID_Europe_1 |  -.0361663   .0313045    -1.16   0.259    -.100
  > 7757
  >         .0284431
  ID_Russia_1 |   .0686272   .0318387     2.16   0.041     .002
  > 9154
  >          .134339
              |
  sex |
  Female |  -.0940792   .0155864    -6.04   0.000     -.12
  > 6248
  >        -.0619103
  age |   .0019488   .0011702     1.67   0.109    -.000
  > 4663
  >         .0043639
              |
  education |
  Professional-technical (vocational) |   .0258005    .035342     0.73   0.472    -.047
  > 1418
  >         .0987428
  Incomplete higher |  -.0220183   .0339259    -0.65   0.522    -.092
  > 0379
  >         .0480013
  Bachelor degree |   .0373062   .0333662     1.12   0.275    -.031
  > 5583
  >         .1061708
  Master degree & Doctorate |   .0460874   .0253667     1.82   0.082    -.006
  > 2669
  >         .0984418
              |
  _cons |     .30615   .0759566     4.03   0.000     .149
  > 3832
```

```
>         .4629167
------------------------------------------------------------------------------
------------------

1004
1005* Analyses based on log-transformed VIINA-variable (not reported in manuscript or SO
  > M.10)
1006reg Comp_scale_1 ln_VIINA_W1_norm c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_
  > scale_1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 i.sex c.age i.
  > education, cluster(w1_q5)

Linear regression                               Number of obs     =        998
                                                F(14, 23)         =      83.79
                                                Prob > F          =     0.0000
                                                R-squared         =     0.1291
                                                Root MSE          =     .13706

                                                      (Std. err. adjusted for 24 cl
> usters in w1_q5)
------------------------------------------------------------------------------
------------------
                             |               Robust
                Comp_scale_1 | Coefficient  std. err.      t    P>|t|     [95%
>  con
>     f. interval]
-------------------------------------------------------------------------------
------------------
             ln_VIINA_W1_norm |  -.0018457   .0031777    -0.58   0.567    -.008
> 4192
>         .0047278
             fearfull_scale_1 |   -.045936   .0285429    -1.61   0.121    -.104
> 9815
>         .0131095
           aggressive_scale_1 |   .0765568   .0247587     3.09   0.005     .025
> 3395
>         .1277741
            sadness_scale_1 |   .0595509   .0265987     2.24   0.035     .004
> 5273
>         .1145745
             selfconf_scale_1 |   .0671073   .0290721     2.31   0.030     .006
> 9671
>         .1272474
                 ID_Ukraine_1 |   .1351711    .048285     2.80   0.010      .03
> 5286
>         .2350562
                  ID_Europe_1 |   .0436148   .0248253     1.76   0.092    -.007
> 7402
>         .0949698
                  ID_Russia_1 |  -.0058767   .0328068    -0.18   0.859    -.073
> 7427
>         .0619894
                              |
                          sex |
                       Female |   .0233848   .0081736     2.86   0.009     .006
> 4764
>         .0402932
                          age |  -.0001856    .000527    -0.35   0.728    -.001
> 2757
>         .0009045
                              |
                    education |
Professional-technical (vocational) |   .0266658   .0205398     1.30   0.207     -.01
> 5824
>         .0691557
           Incomplete higher |   .0414762    .020729     2.00   0.057    -.001
> 4049
>         .0843573
            Bachelor degree |   .0411325   .0232588     1.77   0.090     -.00
> 6982
>         .0892469
          Master degree & Doctorate |   .0553801   .0161017     3.44   0.002     .022
> 0712
```

```
>          .088689
                                    _cons |   .5873941   .0618648     9.49   0.000      .459
> 4169
>          .7153713
------------------------------------------+------------------------------------------------
  ------------------

1007reg Warm_scale_1 ln_VIINA_W1_norm c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_
  > scale_1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 i.sex c.age i.
  > education, cluster(w1_q5)

Linear regression                               Number of obs     =        996
                                                F(14, 23)         =      27.01
                                                Prob > F          =     0.0000
                                                R-squared         =     0.1120
                                                Root MSE          =     .21547

                                                          (Std. err. adjusted for 24 cl
> usters in w1_q5)
------------------------------------------+------------------------------------------------
  ------------------
                                          |               Robust
                       Warm_scale_1 | Coefficient  std. err.      t    P>|t|     [95%
>  con
>     f. interval]
------------------------------------------+------------------------------------------------
  ------------------
                  ln_VIINA_W1_norm |  -.0003591   .0035266    -0.10   0.920    -.007
> 6544
>          .0069363
                  fearfull_scale_1 |   .0945964   .0443955     2.13   0.044     .002
> 7573
>          .1864355
          aggressive_scale_1 |  -.0082757   .0436312    -0.19   0.851    -.098
> 5338
>          .0819824
           sadness_scale_1 |   .1043609   .0645404     1.62   0.120     -.02
> 9151
>          .2378728
                  selfconf_scale_1 |    .103574   .0369416     2.80   0.010     .027
> 1545
>          .1799936
                     ID_Ukraine_1 |   .1014041   .0722151     1.40   0.174    -.047
> 9841
>          .2507923
                      ID_Europe_1 |    .065931   .0318286     2.07   0.050     .000
> 0886
>          .1317734
                      ID_Russia_1 |    .002571   .0335688     0.08   0.940    -.066
> 8714
>          .0720134
                                  |
                              sex |
                           Female |   .0200629   .0148841     1.35   0.191    -.010
> 7272
>          .0508531
                              age |  -.0050409   .0006165    -8.18   0.000    -.006
> 3163
>         -.0037655
                                  |
                        education |
Professional-technical (vocational) |  -.0061837   .0263486    -0.23   0.817    -.060
> 6899
>          .0483225
                 Incomplete higher |  -.0234977   .0338063    -0.70   0.494    -.093
> 4313
>          .0464359
                   Bachelor degree |  -.0605594   .0311683    -1.94   0.064     -.12
> 5036
>          .0039172
          Master degree & Doctorate |  -.0532127   .0256987    -2.07   0.050    -.106
```

```
> 3745
>       -.0000509
                                    _cons |    .626588   .0623239    10.05   0.000     .497
> 6613
>        .7555148
------------------------------------------+------------------------------------------------
  ----------------

1008reg Domi_scale_1 ln_VIINA_W1_norm c.fearfull_scale_1 c.aggressive_scale_1 c.sadness_
  > scale_1 c.selfconf_scale_1 c.ID_Ukraine_1 c.ID_Europe_1 c.ID_Russia_1 i.sex c.age i.
  > education, cluster(w1_q5)

Linear regression                               Number of obs     =        995
                                                F(14, 23)         =      30.74
                                                Prob > F          =     0.0000
                                                R-squared         =     0.0676
                                                Root MSE          =     .25171

                                                          (Std. err. adjusted for 24 cl
> usters in w1_q5)
------------------------------------------+------------------------------------------------
  ----------------
                                          |               Robust
                      Domi_scale_1 | Coefficient  std. err.      t    P>|t|     [95%
>  con
>     f. interval]
------------------------------------------+------------------------------------------------
  ----------------
                  ln_VIINA_W1_norm |  -.0125897   .0067036    -1.88   0.073    -.026
> 4572
>        .0012777
                  fearfull_scale_1 |    .074113   .0507521     1.46   0.158    -.030
> 8758
>        .1791017
                aggressive_scale_1 |   .0667022    .037226     1.79   0.086    -.010
> 3056
>        .1437101
                   sadness_scale_1 |   .0342702   .0646542     0.53   0.601    -.099
> 4772
>        .1680177
                  selfconf_scale_1 |    .136295   .0450642     3.02   0.006     .043
> 0727
>        .2295173
                      ID_Ukraine_1 |   .0118174   .0439589     0.27   0.790    -.079
> 1185
>        .1027533
                       ID_Europe_1 |  -.0385554   .0305963    -1.26   0.220    -.101
> 8486
>        .0247378
                       ID_Russia_1 |   .0760227   .0321126     2.37   0.027     .009
> 5927
>        .1424526
                                   |
                               sex |
                            Female |  -.0971882   .0155146    -6.26   0.000    -.129
> 2826
>       -.0650939
                               age |   .0021419   .0011851     1.81   0.084    -.000
> 3097
>        .0045935
                                   |
                         education |
Professional-technical (vocational) |    .024921   .0363144     0.69   0.499     -.05
> 0201
>         .100043
                 Incomplete higher |  -.0279341   .0338698    -0.82   0.418    -.097
> 9992
>        .0421309
                   Bachelor degree |    .036013   .0342361     1.05   0.304    -.034
> 8097
>        .1068357
```

```
          Master degree & Doctorate |   .0431387   .0259745     1.66   0.110    -.010
> 5937
>          .096871
                                    |
                              _cons |   .2647037   .0822248     3.22   0.004     .094
> 6087
>          .4347987
------------------------------------+----------------------------------------------
  ------------------

1009
1010** Based on wave 2 data (Table SOM.12b)
1011reg Comp_scale_2 VIINA_W2_norm c.fearfull_scale_2 c.aggressive_scale_2 c.sadness_sca
  > le_2 c.selfconf_scale_2 c.ID_Ukraine_2 c.ID_Europe_2 c.ID_Russia_2 i.sex c.age i.edu
  > cation, cluster(w2_q4)

  Linear regression                               Number of obs     =        740
                                                  F(14, 24)         =      42.09
                                                  Prob > F          =     0.0000
                                                  R-squared         =     0.1241
                                                  Root MSE          =     .13133

                                                           (Std. err. adjusted for 25 cl
  > usters in w2_q4)
------------------------------------+----------------------------------------------
  ------------------
                                    |               Robust
                       Comp_scale_2 | Coefficient  std. err.      t    P>|t|     [95%
  >  con
  >     f. interval]
------------------------------------+----------------------------------------------
  ------------------
                     VIINA_W2_norm |   .0013251   .0101465     0.13   0.897    -.019
> 6163
>          .0222665
                  fearfull_scale_2 |  -.0855858   .0172142    -4.97   0.000    -.121
> 1142
>         -.0500573
                aggressive_scale_2 |   .0573742   .0252004     2.28   0.032     .005
> 3631
>          .1093854
                   sadness_scale_2 |    .086151   .0257919     3.34   0.003     .032
> 9192
>          .1393828
                  selfconf_scale_2 |   .0622659   .0289909     2.15   0.042     .002
> 4317
>          .1221001
                     ID_Ukraine_2 |   .1593679   .0377353     4.22   0.000      .08
> 1486
>          .2372497
                      ID_Europe_2 |   .0496091   .0259615     1.91   0.068    -.003
> 9729
>           .103191
                      ID_Russia_2 |  -.0129365   .0485734    -0.27   0.792     -.11
> 3187
>          .0873141
                                  |
                              sex |
                           Female |   .0257735   .0077827     3.31   0.003     .009
> 7108
>          .0418362
                              age |  -.0005816   .0005705    -1.02   0.318     -.00
> 1759
>          .0005959
                                  |
                        education |
Professional-technical (vocational) |  -.0037926   .0281257    -0.13   0.894    -.061
> 8413
>           .054256
                 Incomplete higher |   .0080863   .0273433     0.30   0.770    -.048
> 3475
>          .0645202
```

```
                    Bachelor degree |   .0166586   .0255861     0.65   0.521    -.036
> 1485
>         .0694658
          Master degree & Doctorate |   .0193869   .0229983     0.84   0.408    -.028
> 0792
>          .066853
                                    |
                              _cons |   .6310441   .0533802    11.82   0.000     .520
> 8729
>         .7412154
------------------------------------+----------------------------------------------
  ----------------

1012reg Warm_scale_2 VIINA_W2_norm c.fearfull_scale_2 c.aggressive_scale_2 c.sadness_sca
  > le_2 c.selfconf_scale_2 c.ID_Ukraine_2 c.ID_Europe_2 c.ID_Russia_2 i.sex c.age i.edu
  > cation, cluster(w2_q4)

  Linear regression                               Number of obs     =        728
                                                  F(14, 24)         =      24.64
                                                  Prob > F          =     0.0000
                                                  R-squared         =     0.1206
                                                  Root MSE          =     .23544

                                                            (Std. err. adjusted for 25 cl
  > usters in w2_q4)
------------------------------------+----------------------------------------------
  ----------------
                                    |               Robust
                       Warm_scale_2 | Coefficient  std. err.      t    P>|t|     [95%
  >  con
  >     f. interval]
------------------------------------+----------------------------------------------
  ----------------
                     VIINA_W2_norm |   .0191862   .0202804     0.95   0.354    -.022
> 6705
>         .0610429
                  fearfull_scale_2 |   .1061628   .0456334     2.33   0.029     .011
> 9801
>         .2003454
                aggressive_scale_2 |  -.0685957   .0511487    -1.34   0.192    -.174
> 1615
>           .03697
                  sadness_scale_2 |   .0501229   .0580696     0.86   0.397     -.06
> 9727
>         .1699727
                  selfconf_scale_2 |   .1474133   .0433574     3.40   0.002     .057
> 9279
>         .2368987
                     ID_Ukraine_2 |   .2016106   .0588632     3.43   0.002      .08
> 0123
>         .3230982
                      ID_Europe_2 |   .0645799   .0378985     1.70   0.101    -.013
> 6388
>         .1427986
                      ID_Russia_2 |   .0568291   .0465705     1.22   0.234    -.039
> 2877
>         .1529459
                                  |
                              sex |
                           Female |     .03858   .0155066     2.49   0.020     .006
> 5759
>         .0705841
                              age |  -.0064922   .0009544    -6.80   0.000    -.00
> 8462
>        -.0045223
                                  |
                        education |
  Professional-technical (vocational) |  -.0502897    .037747    -1.33   0.195    -.128
> 1957
>         .0276164
                 Incomplete higher |  -.0862869   .0641405    -1.35   0.191    -.218
> 6663
```

```
>         .0460926
                    Bachelor degree |  -.0627934   .0461403    -1.36   0.186    -.158
> 0222
>         .0324355
          Master degree & Doctorate |  -.0775398    .045346    -1.71   0.100    -.171
> 1293
>         .0160496
                                    |
                              _cons |   .5920556   .0775404     7.64   0.000     .432
> 0201
>         .7520911
------------------------------------+----------------------------------------------
                  ------------------

1013reg Domi_scale_2 VIINA_W2_norm c.fearfull_scale_2 c.aggressive_scale_2 c.sadness_sca
  > le_2 c.selfconf_scale_2 c.ID_Ukraine_2 c.ID_Europe_2 c.ID_Russia_2 i.sex c.age i.edu
  > cation, cluster(w2_q4)

Linear regression                               Number of obs     =        728
                                                F(14, 24)         =      13.90
                                                Prob > F          =     0.0000
                                                R-squared         =     0.0562
                                                Root MSE          =     .25348

                                                               (Std. err. adjusted for 25 cl
> usters in w2_q4)
------------------------------------+----------------------------------------------
                  ------------------
                                    |               Robust
                       Domi_scale_2 | Coefficient  std. err.      t    P>|t|     [95%
>  con
>     f. interval]
------------------------------------+----------------------------------------------
                  ------------------
                       VIINA_W2_norm |  -.0797794   .0171241    -4.66   0.000    -.115
> 1218
>        -.0444369
                  fearfull_scale_2 |   .0712021   .0453258     1.57   0.129    -.022
> 3458
>           .16475
             aggressive_scale_2 |   .0620388   .0657703     0.94   0.355    -.073
> 7044
>          .197782
              sadness_scale_2 |   .0454363   .0370624     1.23   0.232    -.031
> 0567
>         .1219293
                  selfconf_scale_2 |   .1046817   .0510098     2.05   0.051    -.000
> 5975
>         .2099608
                     ID_Ukraine_2 |  -.0339921   .0487065    -0.70   0.492    -.134
> 5175
>         .0665333
                      ID_Europe_2 |   .0057272   .0433362     0.13   0.896    -.083
> 7143
>         .0951687
                      ID_Russia_2 |   .0355095   .0786136     0.45   0.656     -.12
> 6741
>           .19776
                                  |
                              sex |
                           Female |  -.0913048   .0183105    -4.99   0.000    -.129
> 0958
>        -.0535138
                              age |   .0006676   .0010249     0.65   0.521    -.001
> 4477
>         .0027828
                                  |
                        education |
Professional-technical (vocational) |  -.0383892   .0405658    -0.95   0.353    -.122
> 1128
>         .0453344
                 Incomplete higher |  -.0746896   .0531394    -1.41   0.173     -.18
```

```
> 4364
>         .0349847
                  Bachelor degree |  -.0241032    .035892    -0.67   0.508    -.098
> 1808
>         .0499743
         Master degree & Doctorate |  -.0445354   .0369568    -1.21   0.240    -.120
> 8104
>         .0317396
                                   |
                             _cons |   .4846569   .0585157     8.28   0.000     .363
> 8863
>         .6054274
------------------------------------+----------------------------------------------
  ------------------

1014
1015* Analyses based on log-transformed VIINA-variable (not reported in manuscript or SO
  > M.10)
1016reg Comp_scale_2 ln_VIINA_W2_norm c.fearfull_scale_2 c.aggressive_scale_2 c.sadness_
  > scale_2 c.selfconf_scale_2 c.ID_Ukraine_2 c.ID_Europe_2 c.ID_Russia_2 i.sex c.age i.
  > education, cluster(w2_q4)

Linear regression                               Number of obs     =        727
                                                F(14, 23)         =      32.15
                                                Prob > F          =     0.0000
                                                R-squared         =     0.1339
                                                Root MSE          =     .13041

                                                          (Std. err. adjusted for 24 cl
> usters in w2_q4)
------------------------------------+----------------------------------------------
  ------------------
                                    |               Robust
                       Comp_scale_2 | Coefficient  std. err.      t    P>|t|     [95%
>  con
>     f. interval]
------------------------------------+----------------------------------------------
  ------------------
                  ln_VIINA_W2_norm |   .0025742   .0028938     0.89   0.383    -.003
> 4122
>         .0085605
                  fearfull_scale_2 |     -.0855   .0180433    -4.74   0.000    -.122
> 8254
>        -.0481745
            aggressive_scale_2 |   .0579847   .0260042     2.23   0.036     .004
> 1909
>         .1117786
             sadness_scale_2 |   .0892843   .0265635     3.36   0.003     .034
> 3335
>         .1442351
                  selfconf_scale_2 |   .0692507   .0285262     2.43   0.023     .010
> 2396
>         .1282617
                      ID_Ukraine_2 |   .1648015   .0379748     4.34   0.000     .086
> 2445
>         .2433584
                       ID_Europe_2 |   .0481153   .0264057     1.82   0.081    -.00
> 6509
>         .1027396
                       ID_Russia_2 |  -.0153309   .0485989    -0.32   0.755    -.115
> 8653
>         .0852034
                                   |
                               sex |
                            Female |   .0276918    .007823     3.54   0.002     .011
> 5087
>         .0438749
                               age |  -.0006752   .0005732    -1.18   0.251    -.00
> 1861
>         .0005106
                                   |
                         education |
```

```
Professional-technical (vocational) |  -.0040943   .0285078    -0.14   0.887    -.063
> 0672
>         .0548786
                Incomplete higher |   .0084034   .0272693     0.31   0.761    -.048
> 0074
>         .0648143
                  Bachelor degree |   .0169186   .0257265     0.66   0.517    -.036
> 3007
>          .070138
        Master degree & Doctorate |   .0195904   .0232152     0.84   0.407     -.02
> 8434
>         .0676148
                                  |
                            _cons |   .6283565   .0525998    11.95   0.000     .519
> 5454
>         .7371675
------------------------------------------------------------------------------
```

1017 `reg Warm_scale_2 ln_VIINA_W2_norm c.fearfull_scale_2 c.aggressive_scale_2 c.sadness_`
`> scale_2 c.selfconf_scale_2 c.ID_Ukraine_2 c.ID_Europe_2 c.ID_Russia_2 i.sex c.age i.`
`> education, cluster(w2_q4)`

```
Linear regression                               Number of obs     =        715
                                                F(14, 23)         =      26.02
                                                Prob > F          =     0.0000
                                                R-squared         =     0.1240
                                                Root MSE          =     .23538

                                                          (Std. err. adjusted for 24 cl
> usters in w2_q4)
------------------------------------------------------------------------------
                             |               Robust
                Warm_scale_2 | Coefficient  std. err.      t    P>|t|     [95%
>  con
>     f. interval]
-------------------------------------------------------------------------------
             ln_VIINA_W2_norm |   .0029692   .0060121     0.49   0.626    -.009
> 4679
>         .0154062
              fearfull_scale_2 |   .1032017    .044984     2.29   0.031     .010
> 1451
>         .1962582
            aggressive_scale_2 |  -.0807987   .0504046    -1.60   0.123    -.185
> 0684
>         .0234711
               sadness_scale_2 |   .0592763   .0566496     1.05   0.306    -.057
> 9123
>         .1764649
              selfconf_scale_2 |   .1550581   .0433985     3.57   0.002     .065
> 2815
>         .2448347
                 ID_Ukraine_2 |   .2068077   .0590759     3.50   0.002     .084
> 5998
>         .3290156
                  ID_Europe_2 |   .0634301   .0383242     1.66   0.111    -.015
> 8496
>         .1427097
                  ID_Russia_2 |   .0493542   .0452149     1.09   0.286    -.044
> 1801
>         .1428884
                              |
                          sex |
                       Female |   .0408103   .0156274     2.61   0.016     .008
> 4827
>          .073138
                          age |  -.0064855   .0009774    -6.64   0.000    -.008
> 5073
>        -.0044637
                              |
```

```
                          education |
Professional-technical (vocational) |  -.0507233   .0378916    -1.34   0.194     -.12
> 9108
>         .0276615
                 Incomplete higher  |  -.0861958   .0636228    -1.35   0.189    -.217
> 8096
>          .045418
                   Bachelor degree  |  -.0589737   .0466465    -1.26   0.219    -.155
> 4695
>          .037522
          Master degree & Doctorate |  -.0791958   .0451556    -1.75   0.093    -.172
> 6073
>         .0142158
                                    |
                              _cons |   .5983831   .0800611     7.47   0.000     .432
> 7642
>         .7640021
------------------------------------+----------------------------------------------------
  ----------------

1018reg Domi_scale_2 ln_VIINA_W2_norm c.fearfull_scale_2 c.aggressive_scale_2 c.sadness_
> scale_2 c.selfconf_scale_2 c.ID_Ukraine_2 c.ID_Europe_2 c.ID_Russia_2 i.sex c.age i.
> education, cluster(w2_q4)

Linear regression                               Number of obs     =        715
                                                F(14, 23)         =      18.78
                                                Prob > F          =     0.0000
                                                R-squared         =     0.0568
                                                Root MSE          =     .25407

                                                         (Std. err. adjusted for 24 cl
> usters in w2_q4)
------------------------------------+----------------------------------------------------
  ----------------
                                    |               Robust
                       Domi_scale_2 | Coefficient  std. err.      t    P>|t|     [95%
>  con
>     f. interval]
------------------------------------+----------------------------------------------------
  ----------------
                   ln_VIINA_W2_norm |  -.0155218   .0046615    -3.33   0.003    -.025
> 1647
>        -.0058788
                   fearfull_scale_2 |   .0834996   .0441809     1.89   0.071    -.007
> 8956
>         .1748948
                 aggressive_scale_2 |   .0548506   .0663899     0.83   0.417    -.082
> 4873
>         .1921885
                    sadness_scale_2 |     .02757   .0374698     0.74   0.469    -.049
> 9422
>         .1050822
                   selfconf_scale_2 |   .1010124   .0517046     1.95   0.063    -.005
> 9467
>         .2079715
                       ID_Ukraine_2 |  -.0267146   .0502332    -0.53   0.600      -.1
> 3063
>         .0772008
                        ID_Europe_2 |   .0081026   .0434242     0.19   0.854    -.081
> 7273
>         .0979325
                        ID_Russia_2 |   .0341445   .0798484     0.43   0.673    -.131
> 0345
>         .1993235
                                    |
                                sex |
                             Female |  -.0956663   .0182986    -5.23   0.000    -.133
> 5197
>        -.0578128
                                age |   .0005498   .0010328     0.53   0.600    -.001
> 5866
>         .0026863
```

```
                                  |
                         education |
Professional-technical (vocational) |  -.0399239   .0412829    -0.97   0.344    -.125
> 3241
>         .0454762
                 Incomplete higher |  -.0734172   .0539684    -1.36   0.187    -.185
> 0594
>          .038225
                   Bachelor degree |  -.0244657   .0360509    -0.68   0.504    -.099
> 0428
>         .0501113
          Master degree & Doctorate |  -.0412472   .0374474    -1.10   0.282     -.11
> 8713
>         .0362186
                                  |
                             _cons |   .4383067   .0555308     7.89   0.000     .323
> 4325
>         .5531809
------------------------------------------------------------------------------------
  ________________

1019
1020
1021
1022*********************************************************************************
  > ************************************************************
1023********************************************************** RESHAPES DATASET TO LONG
  > FORMAT ******************************************************
1024*********************************************************************************
  > ************************************************************
1025reshape long Competence_ Trustworthy_ Dominant_ Generous_ Strong_ Warm_ Toughminded_
  >  Comp_scale_ Warm_scale_ Domi_scale_ Comp_PCA_ Warm_PCA_ Domi_PCA_ ///
  > afraid_ frightened_ scared_ angry_ hostile_ disgusted_ sad_ lonely_ downhearted_ pro
  > ud_ strong_ confident_ anxiety_scale_ aggressive_scale_ sadness_scale_ selfconf_scal
  > e_ Victimization_ ///
  > Conflict_ ID_Ukraine_ ID_Russia_ ID_Europe_, i(ID_random) j(wave)
  (j = 1 2)
  (variable anxiety_scale_1 not found)
  (variable anxiety_scale_2 not found)

  Data                               Wide   ->   Long
  ------------------------------------------------------------------------------
  Number of observations            1,081   ->   2,162
  Number of variables                 324   ->   293
  j variable (2 values)                     ->   wave
  xij variables:
                Competence_1 Competence_2   ->   Competence_
              Trustworthy_1 Trustworthy_2   ->   Trustworthy_
                    Dominant_1 Dominant_2   ->   Dominant_
                    Generous_1 Generous_2   ->   Generous_
                        Strong_1 Strong_2   ->   Strong_
                            Warm_1 Warm_2   ->   Warm_
              Toughminded_1 Toughminded_2   ->   Toughminded_
                Comp_scale_1 Comp_scale_2   ->   Comp_scale_
                Warm_scale_1 Warm_scale_2   ->   Warm_scale_
                Domi_scale_1 Domi_scale_2   ->   Domi_scale_
                    Comp_PCA_1 Comp_PCA_2   ->   Comp_PCA_
                    Warm_PCA_1 Warm_PCA_2   ->   Warm_PCA_
                    Domi_PCA_1 Domi_PCA_2   ->   Domi_PCA_
                        afraid_1 afraid_2   ->   afraid_
                frightened_1 frightened_2   ->   frightened_
                        scared_1 scared_2   ->   scared_
                          angry_1 angry_2   ->   angry_
                      hostile_1 hostile_2   ->   hostile_
                  disgusted_1 disgusted_2   ->   disgusted_
                              sad_1 sad_2   ->   sad_
                        lonely_1 lonely_2   ->   lonely_
              downhearted_1 downhearted_2   ->   downhearted_
                          proud_1 proud_2   ->   proud_
                        strong_1 strong_2   ->   strong_
                  confident_1 confident_2   ->   confident_
          anxiety_scale_1 anxiety_scale_2   ->   anxiety_scale_
```

```
aggressive_scale_1 aggressive_scale_2   ->   aggressive_scale_
      sadness_scale_1 sadness_scale_2   ->   sadness_scale_
    selfconf_scale_1 selfconf_scale_2   ->   selfconf_scale_
      Victimization_1 Victimization_2   ->   Victimization_
                Conflict_1 Conflict_2   ->   Conflict_
           ID_Ukraine_1 ID_Ukraine_2    ->   ID_Ukraine_
             ID_Russia_1 ID_Russia_2    ->   ID_Russia_
             ID_Europe_1 ID_Europe_2    ->   ID_Europe_
```

```
1026
1027** Labels survey round variable
1028label define waveLB 1 "Round 1 (context primed)" 2 "Round 2 (no context primed)"

1029label values wave waveLB

1030
1031*** Sets panelvar to ID_random
1032xtset ID_random

  Panel variable: ID_random (balanced)

1033
1034
1035********************************************** MAPPING WARTIME LEADER TRAIT PREFEREN
  > CES ************************************************************
1036*** Creates Figure 1
1037reg Comp_scale_ i.wave if Context== 1 & include==1, cluster(ID_random)

  Linear regression                               Number of obs     =        748
                                                  F(1, 373)         =       1.84
                                                  Prob > F          =     0.1756
                                                  R-squared         =     0.0014
                                                  Root MSE          =     .13531

                                               (Std. err. adjusted for 374 clusters in ID
  > _random)
------------------------------------------------------------------------------
             |               Robust
  Comp_scale_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
  > nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |   .0101753   .0074981     1.36   0.176    -.0045685
  >  .024919
       _cons |   .8989899   .0071547   125.65   0.000     .8849212
  > .9130586
------------------------------------------------------------------------------

1038test _cons == .7034314

   ( 1)  _cons = .7034314

         F(  1,   373) =  747.08
              Prob > F =    0.0000

1039test _cons == .5274064

   ( 1)  _cons = .5274064

         F(  1,   373) = 2697.27
              Prob > F =    0.0000
```

```
1040margins, at(wave=(1 2)) level(95)

  Adjusted predictions                                  Number of obs = 748
  Model VCE: Robust

  Expression: Linear prediction, predict()
  1._at: wave = 1
  2._at: wave = 2

------------------------------------------------------------------------------
             |            Delta-method
             |     Margin   std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
         _at |
          1  |   .8989899   .0071547   125.65   0.000     .8849212    .9130586
          2  |   .9091652   .0068442   132.84   0.000     .8957071    .9226233
------------------------------------------------------------------------------

1041marginsplot,  recastci(rcap) yscale(range(0(.1)1)) ylabel(0(.1)1) recast(scatter) yl
  > ine(0) ///
  > xtitle("Competence") ytitle("Competence Importance") title("Competence") legend(off)
  >  scheme(s1mono) name(Comp_war_mean_2waves, replace)

  Variables that uniquely identify margins: wave

1042
1043reg Warm_scale_ i.wave if Context== 1 & include==1, cluster(ID_random)

  Linear regression                               Number of obs     =        748
                                                  F(1, 373)         =       8.36
                                                  Prob > F          =     0.0041
                                                  R-squared         =     0.0049
                                                  Root MSE          =     .24422

                                               (Std. err. adjusted for 374 clusters in ID
  > _random)
------------------------------------------------------------------------------
             |               Robust
  Warm_scale_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
  > nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |  -.0343137     .01187    -2.89   0.004    -.0576542   -
  > .0109732
             |
       _cons |   .7034314   .0118986    59.12   0.000     .6800347
  >  .726828
------------------------------------------------------------------------------

1044test _cons == .8989899

   ( 1)  _cons = .8989899

         F(  1,   373) =  270.13
              Prob > F =    0.0000

1045test _cons == .5274064

   ( 1)  _cons = .5274064

         F(  1,   373) =  218.86
              Prob > F =    0.0000
```

```
1046margins, at(wave=(1 2)) level(95)

  Adjusted predictions                                  Number of obs = 748
  Model VCE: Robust

  Expression: Linear prediction, predict()
  1._at: wave = 1
  2._at: wave = 2

------------------------------------------------------------------------------
             |            Delta-method
             |     Margin   std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
         _at |
          1  |   .7034314   .0118986    59.12   0.000     .6800347     .726828
          2  |   .6691176   .0133337    50.18   0.000      .642899    .6953363
------------------------------------------------------------------------------

1047marginsplot,  recastci(rcap) yscale(range(0(.1)1)) ylabel(0(.1)1) recast(scatter) yl
  > ine(0) ///
  > xtitle("Warmth") ytitle("Warmth Importance") title("Warmth") legend(off)  scheme(s1m
  > ono) name(Warmth_war_mean_2waves, replace)

  Variables that uniquely identify margins: wave

1048
1049reg Domi_scale_ i.wave if Context== 1 & include==1, cluster(ID_random)

  Linear regression                               Number of obs     =        748
                                                  F(1, 373)         =       0.29
                                                  Prob > F          =     0.5897
                                                  R-squared         =     0.0001
                                                  Root MSE          =     .26573

                                               (Std. err. adjusted for 374 clusters in ID
  > _random)
------------------------------------------------------------------------------
             |               Robust
  Domi_scale_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
  > nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |  -.0062389   .0115579    -0.54   0.590    -.0289657
  >  .016488
             _cons |   .5274064   .0134822    39.12   0.000     .5008958
  >  .553917
------------------------------------------------------------------------------

1050test _cons == .8989899

   ( 1)  _cons = .8989899

         F(  1,   373) =  759.61
              Prob > F =    0.0000

1051test _cons == .7034314

   ( 1)  _cons = .7034314

         F(  1,   373) =  170.46
              Prob > F =    0.0000
```

```
1052margins, at(wave=(1 2)) level(95)

  Adjusted predictions                                  Number of obs = 748
  Model VCE: Robust

  Expression: Linear prediction, predict()
  1._at: wave = 1
  2._at: wave = 2

------------------------------------------------------------------------------
             |            Delta-method
             |     Margin   std. err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
         _at |
          1  |   .5274064   .0134822    39.12   0.000     .5008958     .553917
          2  |   .5211676   .0140122    37.19   0.000     .4936148    .5487203
------------------------------------------------------------------------------

1053marginsplot,  recastci(rcap) yscale(range(0(.1)1)) ylabel(0(.1)1) recast(scatter) yl
  > ine(0) ///
  > xtitle("Dominance") ytitle("Dominance Importance") title("Dominance") legend(off)  s
  > cheme(s1mono) name(Domi_war_mean_2waves, replace)

  Variables that uniquely identify margins: wave

1054
1055graph combine Comp_war_mean_2waves Warmth_war_mean_2waves Domi_war_mean_2waves, sche
  > me(s1mono) cols(3)

1056
1057
1058********************************************** TESTING THE CONFLICT-SENSITIVITY
  > HYPOTHESIS **********************************************
1059*** Within-respondent test of the conflict-sensitivity hypothesis (testing if trait
  > preferences change across waves for respondents assigned to the peace condition in w
  > ave 1)
1060* Key results reported in main text and full models in SOM.4b
1061reg Comp_scale_ i.wave if include==1 & Context == 2, cluster(ID_random)

  Linear regression                               Number of obs     =        758
                                                  F(1, 378)         =       2.07
                                                  Prob > F          =     0.1506
                                                  R-squared         =     0.0015
                                                  Root MSE          =     .12291

                                               (Std. err. adjusted for 379 clusters in ID
  > _random)
------------------------------------------------------------------------------
             |               Robust
  Comp_scale_ |      Coef.   Std. err.      t    P>|t|     [95% conf. i
  > nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |  -.0093814   .0065135    -1.44   0.151    -.0221887
  > .0034259
             |
       _cons |   .9166667   .0058095   157.79   0.000     .9052437
  > .9280896
------------------------------------------------------------------------------
```

```
1062reg Warm_scale_ i.wave if include==1 & Context == 2, cluster(ID_random)

Linear regression                               Number of obs     =        758
                                                F(1, 378)         =      28.81
                                                Prob > F          =     0.0000
                                                R-squared         =     0.0167
                                                Root MSE          =     .22356

                                               (Std. err. adjusted for 379 clusters in ID
> _random)
------------------------------------------------------------------------------
             |               Robust
 Warm_scale_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
> nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |  -.0582674   .0108552    -5.37   0.000    -.0796116   -
> .0369232
       _cons |   .7365875   .0109741    67.12   0.000     .7150095
> .7581655
------------------------------------------------------------------------------

1063reg Domi_scale_ i.wave if include==1 & Context == 2, cluster(ID_random)

Linear regression                               Number of obs     =        758
                                                F(1, 378)         =      16.19
                                                Prob > F          =     0.0001
                                                R-squared         =     0.0093
                                                Root MSE          =     .24655

                                               (Std. err. adjusted for 379 clusters in ID
> _random)
------------------------------------------------------------------------------
             |               Robust
 Domi_scale_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
> nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |   .0477133   .0118576     4.02   0.000     .0243982
> .0710284
       _cons |   .4872471   .0127499    38.22   0.000     .4621776
> .5123167
------------------------------------------------------------------------------

1064
1065* Testing full interactions between assigned experimental condition (assigned in wav
  > e 1) and wave (all respondents assigned to think of the ongoing war in wave 2)
1066reg Comp_scale_ i.wave##ib(2).Context if include==1, cluster(ID_random)

Linear regression                               Number of obs     =      1,506
                                                F(3, 752)         =       1.72
                                                Prob > F          =     0.1613
                                                R-squared         =     0.0024
                                                Root MSE          =     .12921
```

(Std. err. adjusted for **753** clusters in **ID_random**)

Comp_scale_	Coefficient	Robust std. err.	t	P>|t|	[95% conf.	interval]
wave						
Round 2 (no context primed)	**-.0093814**	**.0065115**	**-1.44**	**0.150**	**-.0221642**	**.0034014**
Context						
Conflict, now	**-.0176768**	**.0092132**	**-1.92**	**0.055**	**-.0357634**	**.0004099**
wave#Context						
Round 2 (no context primed)#Conflict, now	**.0195567**	**.0099288**	**1.97**	**0.049**	**.0000652**	**.0390481**
_cons	**.9166667**	**.0058076**	**157.84**	**0.000**	**.9052656**	**.9280677**

1067 reg Warm_scale_ i.wave##ib(2).Context if include==1, cluster(ID_random)

Linear regression

Number of obs	=	**1,506**
F(3, 752)	=	**13.15**
Prob > F	=	**0.0000**
R-squared	=	**0.0124**
Root MSE	=	**.23405**

(Std. err. adjusted for **753** clusters in **ID_random**)

Warm_scale_	Coefficient	Robust std. err.	t	P>|t|	[95% conf.	interval]
wave						
Round 2 (no context primed)	**-.0582674**	**.0108517**	**-5.37**	**0.000**	**-.0795707**	**-.036964**
Context						
Conflict, now	**-.0331561**	**.0161812**	**-2.05**	**0.041**	**-.0649219**	**-.0013904**
wave#Context						
Round 2 (no context primed)#Conflict, now	**.0239536**	**.0160798**	**1.49**	**0.137**	**-.007613**	**.0555203**
_cons	**.7365875**	**.0109706**	**67.14**	**0.000**	**.7150509**	**.7581242**

```
1068reg Domi_scale_ i.wave##ib(2).Context if include==1, cluster(ID_random)

Linear regression                               Number of obs     =      1,506
                                                F(3, 752)         =       5.69
                                                Prob > F          =     0.0007
                                                R-squared         =     0.0051
                                                Root MSE          =     .25625

                                                          (Std. err. adjusted for 753
> clusters in ID_random)
------------------------------------------------------------------------------
             |               Robust
  Domi_scale_ | Coefficient  std. err.      t    P>|t|     [95% con
>  f. interval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |   .0477133   .0118538     4.03   0.000
>    .0244428
>    .0709837
             |
     Context |
Conflict, now |   .0401593   .0185499     2.16   0.031
>    .0037435
>    .076575
             |
   wave#Context |
Round 2 (no context primed)#Conflict, now |  -.0539521   .0165531    -3.26   0.001
>   -.0864479
>   -.0214564
             |
       _cons |   .4872471   .0127458    38.23   0.000
>    .4622256
>    .5122687
------------------------------------------------------------------------------

1069
1070
1071*** SOM.7: Within-respondent test of the Conflict-Sensitivity Hypothesis using facto
  > r score variables for trait measurement (reported in table SOM.7.a.2)
1072* Testing within-respondent change among respondent assigned to peace condition in w
  > ave 1
1073reg Comp_PCA_ i.wave if include_PCA==1 & Context == 2, cluster(ID_random)

Linear regression                               Number of obs     =        718
                                                F(1, 358)         =       8.01
                                                Prob > F          =     0.0049
                                                R-squared         =     0.0059
                                                Root MSE          =     .88347

                                               (Std. err. adjusted for 359 clusters in ID
> _random)
------------------------------------------------------------------------------
             |               Robust
   Comp_PCA_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
> nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |  -.1362304   .0481478    -2.83   0.005    -.2309185   -
> .0415423
       _cons |   .1545529   .0419426     3.68   0.000     .0720679
> .2370378
------------------------------------------------------------------------------
```

```
1074reg Warm_PCA_ i.wave if include_PCA==1 & Context == 2, cluster(ID_random)

Linear regression                               Number of obs     =        718
                                                F(1, 358)         =       1.10
                                                Prob > F          =     0.2949
                                                R-squared         =     0.0007
                                                Root MSE          =     .9559

                                               (Std. err. adjusted for 359 clusters in ID
> _random)
------------------------------------------------------------------------------
             |               Robust
   Warm_PCA_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
> nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |   -.049396   .0470866    -1.05   0.295    -.141997
>  .043205
       _cons |   .0740856   .0504443     1.47   0.143    -.0251188
> .1732899
------------------------------------------------------------------------------

1075reg Domi_PCA_ i.wave if include_PCA==1 & Context == 2, cluster(ID_random)

Linear regression                               Number of obs     =        718
                                                F(1, 358)         =      11.29
                                                Prob > F          =     0.0009
                                                R-squared         =     0.0065
                                                Root MSE          =     .94069

                                               (Std. err. adjusted for 359 clusters in ID
> _random)
------------------------------------------------------------------------------
             |               Robust
   Domi_PCA_ | Coefficient  std. err.      t    P>|t|     [95% conf. i
> nterval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no context primed) |   .1515254   .0451026     3.36   0.001     .0628261
> .2402248
       _cons |  -.1184738    .049912    -2.37   0.018    -.2166314   -
> .0203162
------------------------------------------------------------------------------

1076
1077* Full interactions between assigned experimental condition (assigned in wave 1) and
  >  wave (all respondents assigned to think of the ongoing war in wave 2)
1078reg Comp_PCA_ i.wave##ib(2).Context if include==1, cluster(ID_random)

Linear regression                               Number of obs     =      1,452
                                                F(3, 747)         =       3.13
                                                Prob > F          =     0.0250
                                                R-squared         =     0.0035
                                                Root MSE          =     .90495
```

(Std. err. adjusted for **748**
> clusters in **ID_random**)

Comp_PCA_	Coefficient	Robust std. err.	t	P>|t|	[95% con f. interval]
wave					
Round 2 (no context primed)	**-.1318403**	**.0478327**	**-2.76**	**0.006**	**-.2257428** **-.0379379**
Context					
Conflict, now	**-.0949738**	**.0625124**	**-1.52**	**0.129**	**-.2176948** **.0277472**
wave#Context					
Round 2 (no context primed)#Conflict, now	**.0984093**	**.0722142**	**1.36**	**0.173**	**-.0433577** **.2401763**
_cons	**.1445638**	**.0413564**	**3.50**	**0.001**	**.0633752** **.2257524**

1079 reg Warm_PCA_ i.wave##ib(2).Context if include==1, cluster(ID_random)

Linear regression

Number of obs	**1,452**
F(3, 747)	**0.98**
Prob > F	**0.4006**
R-squared	**0.0020**
Root MSE	**.98865**

(Std. err. adjusted for **748**
> clusters in **ID_random**)

Warm_PCA_	Coefficient	Robust std. err.	t	P>|t|	[95% con f. interval]
wave					
Round 2 (no context primed)	**-.0504754**	**.0466751**	**-1.08**	**0.280**	**-.1421053** **.0411545**
Context					
Conflict, now	**-.1173588**	**.0729627**	**-1.61**	**0.108**	**-.2605953** **.0258776**
wave#Context					
Round 2 (no context primed)#Conflict, now	**.0751411**	**.0694201**	**1.08**	**0.279**	**-.0611407** **.2114229**
_cons	**.0721408**	**.0496147**	**1.45**	**0.146**	**-.0252599** **.1695416**

```
1080reg Domi_PCA_ i.wave##ib(2).Context if include==1, cluster(ID_random)

Linear regression                               Number of obs     =      1,452
                                                F(3, 747)         =       3.86
                                                Prob > F          =     0.0094
                                                R-squared         =     0.0039
                                                Root MSE          =     .99433

                                                          (Std. err. adjusted for 748
> clusters in ID_random)
------------------------------------------------------------------------------
             |               Robust
    Domi_PCA_ |      Coef.   Std. Err.      t    P>|t|     [95% conf. interval]
-------------+----------------------------------------------------------------
        wave |
Round 2 (no  |   .1238923   .0457027     2.71   0.007     .0341712    .2136133
             |
     Context |
Conflict, now|   .1682441   .0734209     2.29   0.022     .0241083    .3123799
             |
   wave#Cont |
ext          |
Round 2 (no  |
context      |
primed)#Conf |
lict, now    |  -.2086936     .06582    -3.17   0.002    -.3379079   -.0794794
             |
       _cons |  -.1018748   .0493121    -2.07   0.039    -.1986816    -.005068
------------------------------------------------------------------------------

1081
1082
1083*******************************************************************************
  > ************************************************************
1084
1085log close
       name:  <unnamed>
        log:  C:\Users\au206393\OneDrive - Aarhus universitet\Desktop\PSRM acceptance l
  > og-file\trait_preferences_Ukraine.smcl
   log type:  smcl
  closed on:   6 May 2025, 14:12:01
-------------------------------------------------------------------------------
```