Depletion risks of essential trace metals in a transboundary drain and its surrounding soil using GIS techniques

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Hudaira drain, a big sewage water body originating from India, is an important tributary of river Ravi in Pakistan. In present study, role of the drain in build-up/depletion of metal micronutrients Cu, Fe, Mn and Zn in surrounding agricultural lands has been investigated. Soil samples, up to a vertical distance of 1200 m on both sides of the drain and drain water samples were collected and analysed for metal micronutrients. Soil analysis results were interpolated by using the best optimized interpolator to generate continuous variation of the selected metals. Further patterns have been identified using proximity stat with consecutive neighbouring zones of 100 m extent. Cu and Mn concentrations were higher in some drain water samples, whereas Fe and Zn were found below guidelines in all collected water samples. In adjacent agricultural land the metals did not show uniform dispersal pattern along the drain length. However, vertical pattern of distribution indicated that drain water was controlling Cu and Zn concentrations, positively, whereas it was causing a dilution effect on Fe and Mn build-up in the surrounding agricultural lands.

Keywords: copper, iron, manganese, zinc, geographic information system, spatial interpolation, proximity analysis.

INTRODUCTION

Copper (Cu), Iron (Fe), Manganese (Mn) and Zinc (Zn) are essential plant micronutrients that perform various vital functions in plant body. They act as cofactors of several enzymes involved in photosynthesis, oxidation-reduction reactions, nitrogen fixation, and metabolism of proteins and carbohydrates etc. (Osman, 2013). Initially, these metal nutrients were enough in most soils of the world to support plant growth. However, intensive exhaustive cropping has resulted in the deficiency of one or more micronutrients particularly in alkaline calcareous soils and hence these need to be applied through fertilizers (Ryan *et al.*, 2013).

On the other hand, owing to irrigation with urban or industrial effluents and other anthropogenic activities, toxic levels of Cu, Fe, Mn and Zn may be accumulated in agricultural lands to drastically affect plant growth (Wuana and Okiyeimen, 2011). From polluted soils, high concentrations of these metals may enter the food chain to cause various disorders in human body like anaemia, neurodegeneration in brain, and

hair and skin discolorations, dermatitis, and respiration related problems (Shakir *et al.*, 2016). Keeping in view agricultural and environmental importance of these metals, it is necessary to keep an eye on their build up in soils and their possible sources.

Hudaira Drain (HD) is an international drain, which originates from Batala in India. after covering nearly 55 km on Indian side it enters Pakistan and flows about 63 km before joining the River Ravi. In terms of latitude, the drain extends from 31°31'4.601"N to 31°24'23.612"N and in terms of longitude, it extends from 74°36'24.413"E to 74°6'15.751"E in the outskirts of Lahore city with an average annual discharge of about 180 cusecs (Yamin and Ahmad, 2007). It was initially a storm water drain but now a perennial drain, carries untreated sewage water and industrial effluent all the yearlong (Khan *et al.*, 2003). There are about 250 industries of different sizes which throw their discharge in HD on Indian and Pakistani sides (Qureshi and Syed, 2014). The drain water is used for irrigation in adjacent agricultural lands.

In recent years, several studies have been carried out on HD water, adjacent agricultural land, and vegetation. In 2007, in the drain water, concentrations of Cu, Fe, Mn, Cd, Cr and Ni were found higher than the guidelines suggested by WWF for Pakistan (WWF, 2007; Kashif *et al.*, 2009). The concentration of some metals was also found higher than permissible limits in soil and vegetable samples collected from the fields irrigated with the drain water (Kashif *et al.*, 2009). However, in 2003, the drain water had low heavy metal load and Cu, Cr, and Cd concentrations in the water were below permissible limits (Khan *et al.*, 2003). Some other studies focus on ground water quality (Khattak *et al.*, 2012), distribution of herbaceous vegetation in surrounding of HD, and impact of irrigation with HD water on plants like *Eucalyptus camaldulensis* (Shah *et al.*, 2010), *Brassica campestris*, *Psidium guajava* (Muhammad *et al.*, 2013) and *Spinacia oleracea* (Yamin and Ahmad, 2007).

However, there is a knowledge gap about the build-up or depletion of Cu, Fe, Mn and Zn in agricultural lands adjacent to the drain. In the present study, advanced GIS techniques have been used to identify the role of HD on these micronutrients' status in surrounding soils. The results of the study will help to find the element(s) which need to focus on nutrient management of crops irrigated with water from the drain or from the tube wells installed on aquifer getting its refill possibly from the drain.

MATERIALS AND METHODS

Study Area: On Pakistan side, HD passes through the southern outskirts of metropolitan Lahore; provincial capital of Punjab and the second largest population unit of Pakistan with semi-arid subtropical climate. It is situated over flat alluvial plains and has a latitudinal extent of 31°15' N - 31°45' N and longitudinal extent of 74°01' E - 74°39' E. Associations of the study area are given in Fig. 1.

Figure 1. Study area and its associations.

In context of temperature and rainfall Lahore experiences large variations. It has very long and acute summer season, dry winters, and rainy monsoon season. On average it has annual rainfall of about 680 mm, with peak from June to September (monsoon season). This region is part of inter-fluvial Bari Doab between, Sutlej and Ravi rivers (Muhammad *et al.*, 2013). Geologically the study area formed over alluvial deposits of Quaternary period, with thickness varies maximum to 300 m and contains different types of rocks largely silt, sand and clay (Khattak *et al.*, 2012). Surroundings of HD are being used for agricultural, industrial, and residential purposes. Rice, wheat, maize, and potatoes are major crops of the area; however, some farmers also grow vegetables and fruits to meet their needs.

Soil and Drain Water Sampling and Analysis: Soil samples, from 0-15 cm depth, were collected along with their spatial coordinates in sets of nine. Each set was separated by about 6000 m from others and comprised of three composite samples: one near the HD, one to its right side within 1200 m and similarly one to its left side (Figure 1). In this way the data used comprised of 27 composite sample points. Collected soil samples were air-dried, ground, and sieved through 2 mm diameter. Known amount of each sample was wet digested with nitric acid and perchloric acid mixed in 2:1 volume ratio (Hseu, 2004). The digested contents were diluted to 50 mL volume and run-on Atomic Absorption Spectroscopy (Model PG - 990) for Cu, Mn, Zn and Fe. The metal concentrations along with their spatial coordinates were used to create geographic layer in the form of shapefile for further spatial analysis.

From sites in drain, parallel to nine soil sampling set points, water samples were also collected in April (pre-monsoon) and October (post-monsoon), and analyzed for Cu, Fe, Mn, and Zn on the same instrument as described in soil analysis.

GIS Analysis

Interpolation: As the measured metal concentrations in soil were location specific so, to make a continuous spatial distribution of the parameters, interpolated surfaces were generated over the study area that was extended to 1200 m on both sides of HD.

Among the deterministic families of interpolators Inverse Distance Weighted (IDW) is the simplest and the most used technique that is purely based on Tobler's first law of geography (Ketata-Rokbani *et al.*, 2011; Bairu *et al.*, 2013; Selvam *et al.*, 2014; Mahmood *et al.*, 2016). This law is stated as "Everything is related to everything else, but near things are more related than distant things". Geo-statistics and spatial autocorrelation are based upon this concept. More advanced interpolation techniques have been developed that are estimating parameter's values at un-sampled location with a better accuracy. Kriging family of interpolators is an example of such modern techniques that is recommended as a better alternate for creating continuous geographic layers from limited samples by recent studies (Alqadi *et al.*, 2014; Taghizadeh-Mehrjardi, 2014; Mahmood *et al.*, 2016).

In IDW sampled points nearer to locations of estimations get more importance in the estimating formula than those that are

Table 1. Optimization of interpolation method on the basis of RMSE.

Soil Quality Parameter	Transformation and data properties		Optimal Interpolation Methods	Mean	RMSE
Zn	Transformation:	Log	Inverse Distance Weighted	0.4226	36.46
			Global Polynomial	0.0673	36.80
	Skewness:	0.2652	Local Polynomial	0.7032	37.66
			Radial Basis Function	0.4928	36.81
	Kurtosis:	2.1136	Ordinary Kriging with CSv	0.0018	35.05
			Ordinary Kriging with SSv	0.0663	35.05
	1st Quartile:	4.2550	Ordinary Kriging with ESv	0.0537	35.03
			Ordinary Kriging with GSv	0.0018	35.05
	Median:	4.4362	Simple Kriging with CSv	0.9732	34.00
			Simple Kriging with SSv	0.7982	34.06
	3rd Quartile:	4.8086	Simple Kriging with ESv	0.9951	34.29
			Simple Kriging with GSv	0.9708	34.15
	Mean:	4.5116	Universal Kriging with CSv	-0.0018	35.05
			Universal Kriging with SSv	0.0663	35.05
Fe	Std div:	0.3419	Universal Kriging with ESv	0.0538	35.03
			Universal Kriging with GSv	-0.0018	35.05
	Transformation:	Box-cox	Inverse Distance Weighted	-0.1476	24.93
			Global Polynomial	1.9745	26.64
	Skewness:	-1.690	Local Polynomial	-0.1533	24.48
			Radial Basis Function	0.3580	27.01
	Kurtosis:	6.159	Ordinary Kriging with CSv	-0.0013	23.68
			Ordinary Kriging with SSv	-0.0112	23.74
	1st Quartile:	880.450	Ordinary Kriging with ESv	-0.0023	23.80
			Ordinary Kriging with GSv	-0.0013	23.72
	Median:	887.200	Simple Kriging with CVg	1.3996	22.71
			Simple Kriging with SSv	0.9216	22.89
	3rd Quartile:	896.840	Simple Kriging with ESv	1.4231	22.78
			Simple Kriging with GSv	0.9921	22.83
Cu	Mean:	884.970	Universal Kriging with CSv	0.0072	23.56
			Universal Kriging with SSv	0.0132	23.63
	Std div:	23.101	Universal Kriging with ESv	0.1150	23.62
			Universal Kriging with GSv	1.0042	23.76
	Transformation:	Log	Inverse Distance Weighted	0.3819	24.09
			Global Polynomial	-0.0455	24.18
	Skewness:	0.8249	Local Polynomial	0.5468	25.30
			Radial Basis Function	0.9851	25.64
	Kurtosis:	2.7514	Ordinary Kriging with CSv	-0.5731	22.91
			Ordinary Kriging with SSv	-0.5761	22.91
	1st Quartile:	3.4609	Ordinary Kriging with ESv	-0.4685	22.89
			Ordinary Kriging with GSv	-0.5794	22.91
	Median:	3.6623	Simple Kriging with CSv	-0.3689	21.47
			Simple Kriging with SSv	0.2075	22.11
Mn	3rd Quartile:	3.9051	Simple Kriging with ESv	0.0348	22.15
			Simple Kriging with GSv	0.0763	22.14
	Mean:	3.7479	Universal Kriging with CSv	-0.5731	22.91
			Universal Kriging with SSv	-0.5761	22.91
	Std div:	0.4049	Universal Kriging with ESv	-0.4685	22.89
			Universal Kriging with GSv	-0.5795	22.91
	Transformation:	Box-cox	Inverse Distance Weighted	0.9905	18.70
			Global Polynomial	-0.1369	19.84
	Skewness:	0.457	Local Polynomial	-1.1384	17.90
			Radial Basis Function	0.7474	18.59
	Kurtosis:	2.208	Ordinary Kriging with CSv	-0.2449	17.47
			Ordinary Kriging with SSv	-0.2807	17.46
	1st Quartile:	83.400	Ordinary Kriging with ESv	-0.2874	17.47
			Ordinary Kriging with GSv	-0.2563	17.47
Mn	Median:	95.000	Simple Kriging with CSv	0.2682	17.99
			Simple Kriging with SSv	0.3279	18.01
	3rd Quartile:	117.560	Simple Kriging with ESv	0.4319	18.26
			Simple Kriging with GSv	0.3532	17.98
	Mean:	100.70	Universal Kriging with CSv	-0.2449	17.47
			Universal Kriging with SSv	-0.2807	17.46
	Std div:	25.852	Universal Kriging with ESv	-0.2874	17.47
			Universal Kriging with GSv	-0.2563	17.47

CVs: Circular Semivariogram; SVg: Spherical semivariogram; EVg: Exponential semivariogram; GSv: Gaussian semivariogram

far apart, how much more importance they will get depends on the power value in the formula. Although default value of this power is 2 in almost all the software, but for better results this value should be optimized to natural setting and local

conditions of the phenomenon. Likewise, there are different types of kriging i.e., ordinary, simple, and universal kriging, belongs to stochastic family of interpolators, that use different methods for surface estimation.

Further variations in kriging methods are being made by fitting various semi-variogram models i.e., circular, exponential etc. So, to make the best possible interpolated continuous surfaces of the metals' concentrations in soil all the interpolation techniques were tested for each metal by verifying the suitability using cross-validation. Cross-validation provided values of mean as well as root mean square error (RMSE) to determine accuracy of the created surfaces. It is also worth mentioning here that almost all the interpolation methods assume that the sampled data is normally distributed. So, to comply with the assumption, the data was first checked for its distribution and transformed where needed. Normalized distribution was being standardized using the best values of skewness and kurtosis. Results of this exercise are provided in Table 1.

Analysis of proximity: To have spatial distribution patterns of soil Cu, Fe, Mn and Zn, maps were prepared over a common scale. To check the study hypothesis that wastewater flowing in HD is controlling concentration of selected metals in surrounding soil, an additional spatial analysis of proximity was designed. For this analysis study area around the source was divided into 12 neighbors based on vertical distance from the source. Each of the zones was of 100 m radial distance from the source on both sides. Spatial distribution of proximity zones is shown in Fig. 2.

Figure 2. Spatial distribution of proximity zones.

Finally, the predicted values in each of the zones were averaged to reduce the effects of other contributory sources, and to get the pattern of distance dependent variations, if exists. Although this way of elongated zones averaging over both sides will surely result into very small variation of parametric values, but even then, it can be helpful to investigate dependence of variations over distance from the source. On the other hand, this averaging over large area is the only way to overcome any local variation that may exists at certain points along length of the area. In addition to the common scale maps of the averaging results, distance dependent variation profiles of the proximity analysis have also been prepared to explore trends and patterns of variation. To further remove effects of soil quality controlling sources other than HD, an additional trend function of EMA (Exponential Moving Average) was applied to the graphs of distance dependent variations. EMA is a moving average function that assigns weights to measured data values based on inverse distance law while estimation for a location. So, infect it is a momentum indicator that suppress insignificant anomalies, using weighted moving average function (Mahmood *et al.*, 2017).

RESULTS

Permissible limits of Cu, Fe, Mn and Zn given in Table 2 shows that Cu and Mn concentrations were higher in some drain water samples, whereas Fe and Zn were found below guidelines in all the collected water samples. In September (post-monsoon season), possible dilution effect of monsoon season was not observed on these metal concentrations.

Spatial Distribution: Spatial distributions of soil Cu, Fe, Mn and Zn along the length of HD arranged in Fig. 3 reveal that Zn and Cu did not show any pattern that varies along the drain length. However, spot patterns of high and low values for both the parameters were almost similar (Fig. 3).

Intensity of total Fe in soil has been seen intensifying along downstream length of the source with maximum value in western edge of the study area where the drain falls into the River Ravi (Fig. 3). Spatial distribution of Mn shows that its intrusion into surrounding soils intensified just ahead of the middle of study area, whereas relatively less concentration was present at both eastern and western edges (Fig. 3).

Table 2. Metal concentrations in soil and Hudiara drain water samples.

Metal	Total conc. in soil (ppm)		Conc. in drain water (ppm)			
			Pre-monsoon		Post-monsoon	
	Range	Mean±SD	Range	Mean±SD	Range	Mean±SD
Cu	25-104	46±22	0.01-0.21	0.06±0.056	0.01-0.24	0.09±0.09
Fe	812-918	886±23	0.02-0.13	0.08±0.04	0.02-0.11	0.07±0.04
Mn	60-148	102±26	0.07-0.60	0.26±0.18	0.67-1.05	0.84±0.11
Zn	47-166	96±34	0.04-0.14	0.07±0.03	0.04-0.13	0.06±0.03

Figure 3. Spatial distribution of total Cu, Fe, Mn and Zn in soil along the length of Hudiara drain.

Proximity Analysis: Fig. 4 shows the results of proximity averaging of metal concentrations in soil with zoom-in at middle of the study area. As averaging was done along length of the study area so the resultant patterns of distant dependent variations away from the source were not only same but also the resultant variation was very small. For detailed insight into variations of soil Cu, Fe, Mn and Zn away from the source, distance dependent graphs of parametric averages are also given in Fig. 5.

Figure 4. Results of the proximity stat analysis.

Both Zn and Cu are showing decrease in their average total concentration away from the source. However, the behaviour of Fe in going away from HD was entirely opposite to that of Zn and Cu and was increasing with distance away from the drain. Like that of Fe, Mn concentration in soils was found increasing with distance away from HD. However, dilution effect lasts after about 900 m away from the drain and its concentration got independent of the distance from the source (Figure 4 and 5).

Figure 5. Distance dependent vertical profiles of selected metals away from Hudiara drain.

DISCUSSION

Common distribution spot pattern of Cu and Zn in soils seems to be due to common toxicity or deficiency of Zn and Cu in drain water used for irrigation as mean concentrations of both the metals were comparable to each other in drain water in pre-monsoon and post-monsoon seasons (Table 2). Besides direct irrigation, another possible way HD can contaminate surrounding agricultural land is through contaminating underground aquifer, which is then pumped for irrigation purposes (Khattak *et al.*, 2012). Due to the influence of soil profile, both Cu and Zn can accompany each other in percolating water to reach aquifer. This hypothesis gets support from the fact that both the metals have synergistic effect on their extraction from soil to aqueous medium (Luo and Rimmer, 1995). In this interaction the effect of Cu on Zn solubility is more prominent than that of Zn on Cu (Luo and Rimmer, 1995).

There are several tributaries of HD and important ones are Minhala, Charrar, Ferozepur road and Sattu Katla drain. The Charrar drain meets HD at about the middle of our study area and might be a cause of Mn infestation in the surrounding before and after entering to HD. The drain Sattu Katla joins the HD near Bhobattian chowk on the Raiwind road. This is the same place where Mn concentration decreased and that of Fe increased in the surrounding soils. This indicates that Sattu Katla drain, running through Quaid-e-Azam industrial estate and carrying about 7500 m³ hr⁻¹ water to HD, causing a dilution of Mn and infestation of Fe in agricultural land surrounding HD. This hypothesis is supported by a Sattu Katla water quality report claiming mean concentration of Fe in water as 0.529 ppm (Majeed *et al.*, 2018), whereas overall average concentration of Fe in HD was 0.06 ppm (Table 2). However, we found no report on Mn status of Sattu Katla and Charrar drains water.

When some source is responsible for intrusion of something in surrounding than maximum concentration of the intruded element has been found near the source and vice versa. According to this rule, the concentrations of Cu and Zn seem to be positively controlled by HD. On the other hand, total concentrations of Fe and Mn in soil were also very much dependent on the drain which tended to decrease it. This indicates that HD water has caused a dilution effect on Fe and Mn build up in the surrounding soils. This is possible because the soils already have a reasonable concentration of total Fe ranging from 812 to 918 ppm with an average of 886 ppm (Table 2), and the drain water was only slightly contaminated with the metal (0.08 ppm; Table 2); the concentration was far below the permissible limit of 5 ppm (WWF, 2007).

Conclusion: Our results oppose the general perception that drain water always add metals to soils, and management of metal micronutrients in the fertilizer plan is mostly neglected in drain-influenced soils. Instead, these soils also need site

specific fertilizer recommendations for micronutrients along with macro nutrients.

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