

# Soil microbial community structure and functionality changes in response to long-term metal and radionuclide pollution

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## Summary

**Microbial communities are essential for a healthy soil ecosystem. Metals and radionuclides can exert a persistent pressure on the soil microbial community. However, little is known on the effect of long-term co-contamination of metals and radionuclides on the microbial community structure and functionality. We investigated the impact of historical discharges of the phosphate and nuclear industry on the microbial community in the Grote Nete river basin in Belgium. Eight locations were sampled along a transect to the river edge and one location further in the field. Chemical analysis demonstrated a metal and radionuclide contamination gradient and revealed a distinct clustering of the locations based on all metadata. Moreover, a relation between the chemical parameters and the bacterial community structure was demonstrated. Although no difference in biomass was observed between locations, cultivation-dependent experiments showed that communities from contaminated locations survived better on singular metals than communities from control locations. Furthermore,**

**nitrification, a key soil ecosystem process seemed affected in contaminated locations when combining metadata with microbial profiling. These results indicate that long-term metal and radionuclide pollution impacts the microbial community structure and functionality and provides important fundamental insights into microbial community dynamics in co-metal-radionuclide contaminated sites.**

## Introduction

Microorganisms are essential for a healthy soil ecosystem. They have primary functions in biogeochemical cycles (Chen *et al.*, 2003; Hayat *et al.*, 2010; Madsen, 2011), organic matter turnover (Six *et al.*, 2004; Kindler *et al.*, 2009; Miltner *et al.*, 2012), formation and maintenance of soil structure and fertility (Bronick and Lal, 2005; Schulz *et al.*, 2013). For example, nitrogen fixation and nitrification are necessary to supply soluble nitrogenous compounds for plant growth (Kowalchuk and Stephen, 2001; Chen *et al.*, 2003; Xia *et al.*, 2011). Consequently, changes in the microbial community structure or inhibition of its functions can alter the soil ecosystem.

Anthropogenic activities such as mining and metallurgical processes, phosphate industry, agriculture, and industry in general contribute to a widespread pollution of soils with metals and radionuclides (UNSCEAR, 1993; IAEA, 2003; Loska *et al.*, 2004; Práválie, 2014). *In situ* bioremediation has several advantages compared to standard physicochemical and mechanical techniques with respect to eco-friendliness or cost. However, it necessitates the presence of metal-resistant microbial communities (Groudev *et al.*, 2001; Prakash *et al.*, 2013; Khalid *et al.*, 2017).

Metals and radionuclides can exert a persistent pressure on the microbial community (Frostegard *et al.*, 1993; Simonoff *et al.*, 2007; Wang *et al.*, 2007). Indeed, numerous studies using a culture-dependent or -independent approach demonstrated diverse effects of metal or radionuclide contamination on microbial communities (Sandaa *et al.*, 2001; Ellis *et al.*, 2003; Oliveira and Pampulha, 2006; Golebiewski *et al.*, 2014; Azarbad *et al.*, 2015; Sitte

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et al., 2015; Guo et al., 2017; Li et al., 2017; Beattie et al., 2018). However, it is difficult to generalize the response of the residing microbial community to the metal or radionuclide contamination. For instance, some studies report a reduction of soil microbial biomass (Oliveira and Pampulha, 2006; Zhang et al., 2016; Song et al., 2018), while in other studies, no clear association between contamination and microbial biomass was observed (Gillan et al., 2005; Rousk and Rousk, 2018). Although the bacterial diversity frequently remained similar to non-contaminated soils, changes in microbial community structure or physiological status have been demonstrated (Sandaa et al., 2001; Ellis et al., 2003; Azarbad et al., 2015; Sitte et al., 2015; Guo et al., 2017; Li et al., 2017; Theodorakopoulos et al., 2017; Beattie et al., 2018). On the other hand, microbes can influence metal/radionuclide mobility, toxicity and distribution, hence play an important role in the cycling of metals and radionuclides (Simonoff et al., 2007). Key processes are reduction, uptake and accumulation by cells, biosorption and complexation with proteins, polysaccharides and microbial biomolecules, and biomineralization with phosphates and carbonates (Groudev et al., 2001; Lloyd and Lovley, 2001; Lloyd, 2003; Ruggiero et al., 2005). Besides, resistance mechanisms can be transferred throughout the community (Aminov, 2011). Consequently, long-term exposure enables microbial communities to adapt to metal or radionuclide contamination (Shi et al., 2002; Martinez et al., 2006). Therefore, detailed knowledge of the species and functional diversity, and physiology of the indigenous microbial population is of paramount importance. Although several studies have shown the effect of metals and/or uranium on the microbial community (Sandaa et al., 1999, 2001; Azarbad et al., 2015; Sitte et al., 2015; Boteva et al., 2016; Guo et al., 2017; Li et al., 2017; Sutcliffe et al., 2017; Beattie et al., 2018; Jacquiod et al., 2018), studies where the combination of metals and several radionuclides is investigated remain scarce.

We hypothesize that long-term co-contamination of metals, natural and artificial radionuclides affects the microbial community structure and functionality. To address this hypothesis, we studied the soil microbiome at various locations in a contaminated meadow of the Grote Nete river basin near Lier, a historically contaminated site due to discharges from the phosphate industry and the nuclear industry. We focused on the following questions: (i) is there a difference in the microbial community structure among locations with varying contamination levels? (ii) Can we link the observed differences to the contamination profile? (iii) Are there functional changes in the microbial community from the contaminated locations? Results provide important fundamental

insights into microbial community dynamics in co-metal-radionuclide contaminated sites.

## Results

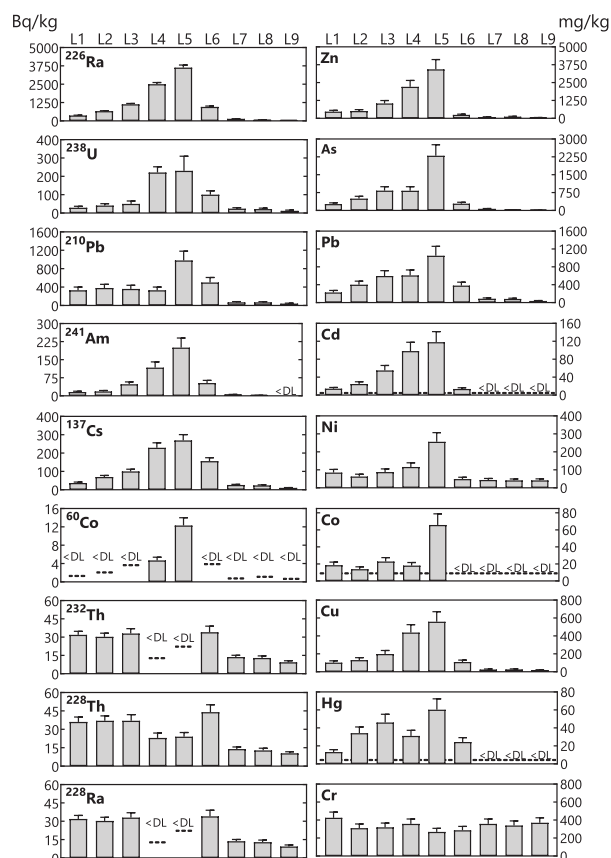
### *Experimental site description*

A contaminated meadow of the Grote Nete basin near Lier was sampled in eight locations (L1–L8) as well as one adjacent unpolluted location (L9). The meadow is located in the Campine region of Belgium, approximately 47 and 37 km downstream the outflow of the Molsse Nete and the Grote Laak respectively (Fig. S1). The Molsse Nete is contaminated with artificial radionuclides (e.g.  $^{137}\text{Cs}$ ,  $^{241}\text{Am}$ ,  $^{60}\text{Co}$ ) due to cumulative historical discharges between 1960 and 1980 from the nuclear industries located in Mol and Dessel, although daily discharge limits were never exceeded. In addition, the former phosphate industry located in Tessenderlo is responsible for the contamination in the Grote Laak. Discharges until the early 1990s resulted in high concentrations of  $^{226}\text{Ra}$  and several metals (e.g. Cd, Zn, As) in the water bottom and river banks.

### *A contamination gradient was observed among the different sample locations*

General soil characteristics and the total and dissolved concentration of several major ions were determined for each sampling point (Fig. S2). The pH was slightly acidic and ranged from  $5.08 \pm 0.23$  to  $5.49 \pm 0.25$ , but did not differ significantly across locations. Moisture content gradually increased from L1 ( $17.21\% \pm 0.19\%$  wt/wt) to L5 ( $53.8\% \pm 0.08\%$  wt/wt). Location L6 showed a slightly lower moisture content than L5 ( $40.18\% \pm 1.19\%$  wt/wt) and levels further decreased in L7–L9 ( $11.82\% \pm 0.3\%$  wt/wt,  $16.8\% \pm 0.24\%$  wt/wt and  $7.19\% \pm 0.01\%$  wt/wt, respectively). Locations L7–L9 also contained the least amount of total organic carbon (TOC) with an average of  $3.6\% \pm 0.4\%$ . Location L1 contained the highest TOC ( $17.1\% \pm 0.3\%$  wt/wt), while values of L2–L6 ranged from  $9.0\% \pm 0.3\%$  (wt/wt) to  $13\% \pm 0.3\%$  (wt/wt). These high TOC levels could be explained by flooding of the river banks and the consequent deposition of river sediments. Total inorganic carbon (TIC) was not detected in the samples.

Total concentrations of nine radionuclides ( $^{137}\text{Cs}$ ,  $^{210}\text{Pb}$ ,  $^{226}\text{Ra}$ ,  $^{228}\text{Ra}$ ,  $^{60}\text{Co}$ ,  $^{241}\text{Am}$ ,  $^{238}\text{U}$ ,  $^{228}\text{Th}$ , and  $^{232}\text{Th}$ ) and nine metals (Co, Ni, Cu, Zn, As, Cd, Hg, Pb and Cr) were measured with gamma spectroscopy and EDXRF, respectively (Fig. 1). The levels of the natural radionuclide  $^{232}\text{Th}$  and its daughter products  $^{228}\text{Ra}$  and  $^{228}\text{Th}$  ranged within natural levels for Belgian subsoil (5–50 Bq/kg (FANC, 2020)) in all samples. The natural radionuclide



**Fig. 1.** Overview of the total levels of radionuclides and metals at the different locations (L1–L9). Radionuclide levels are presented in Bq/kg soil and metal levels are shown as mg/kg soil. Error bars represent the measurement uncertainty. Values under the detection limit are shown as <DL, which is indicated with dotted line.

$^{238}\text{U}$  was also within the expected range (5–50 Bq/kg (FANC, 2020)) in most locations, however, a two- to five-fold increase was measured at L4, L5 and L6. The  $^{226}\text{Ra}$  level in L9 was as expected between 10 and 30 Bq/kg (Paridaens and Vanmarcke, 2001; Licour *et al.*, 2017), while a two- and four-fold increase were measured in L8 and L7, respectively. The  $^{226}\text{Ra}$  levels in all other samples were 30 to 120 times higher compared to background levels. Similarly, L1–L6 contained up to nine times higher levels of the natural isotope  $^{210}\text{Pb}$  compared to the expected level in soil (19–111 Bq/kg (Ibrahim and Whicker, 1987)).

Next to natural radionuclides, artificial radionuclides ( $^{60}\text{Co}$ ,  $^{137}\text{Cs}$  and  $^{241}\text{Am}$ ) were measured. Although  $^{60}\text{Co}$  levels were below the detection limit in seven locations, levels of  $^{241}\text{Am}$  and  $^{137}\text{Cs}$  increased from L1 to L5. Levels decreased from L6 onwards and L9 can be seen as non-contaminated. Most metals showed a similar trend as the radionuclides, that is, levels gradually increased from L1 to L5 and decreased again up to the non-contaminated L9. The highest levels were found for

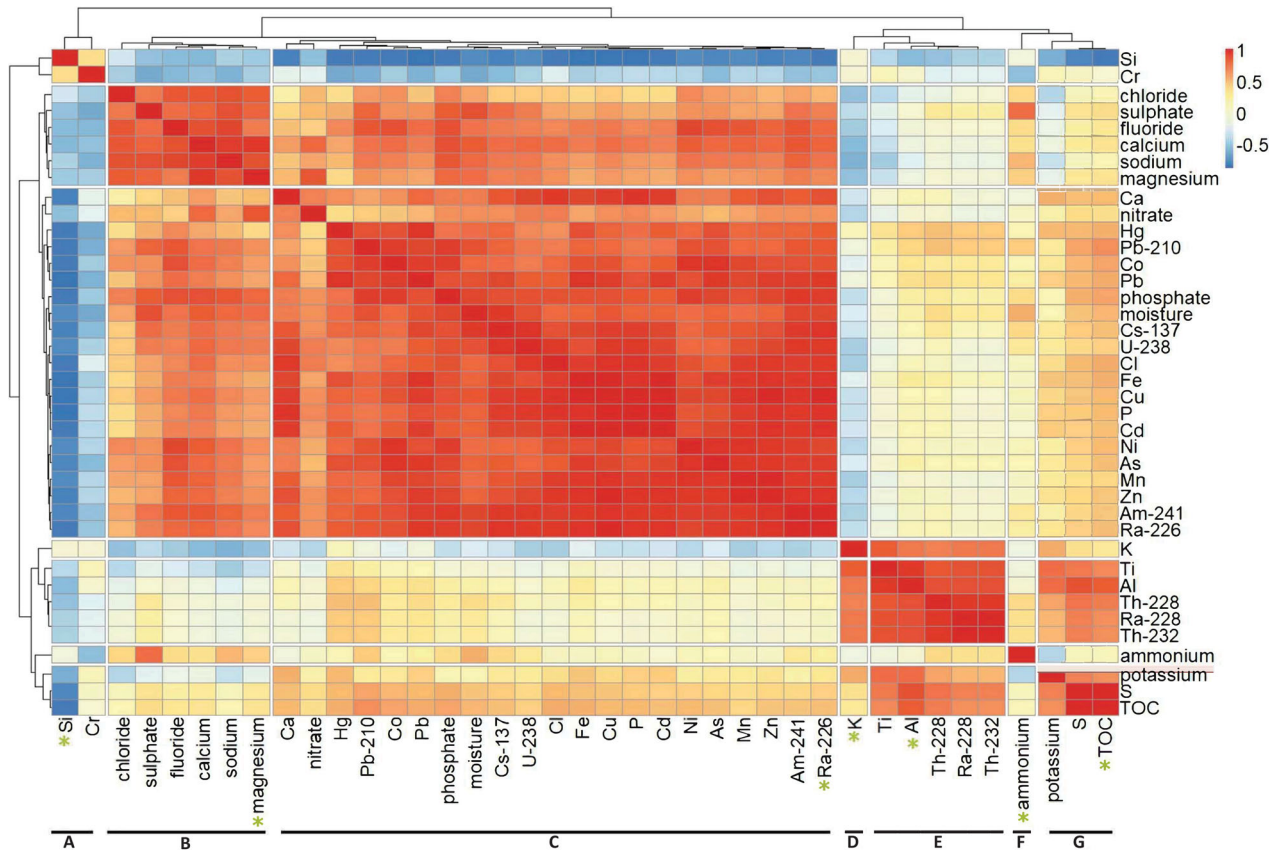
As and Zn, ranging from 9.6 mg/kg to 2300 mg/kg and from 49.9 mg/kg to 3430 mg/kg, respectively (Fig. 1).

A Pearson correlation-based clustering of the measured compounds was performed with parameters that were significantly different among the locations and above the detection limit in most locations (Fig. 2). Seven distinct clusters were identified, with the largest cluster (cluster C) containing most metals and radionuclides, moisture, nitrate, total calcium, phosphate and total chloride (Fig. 2). The three radionuclides originating from the  $^{232}\text{Th}$  decay series, Al, and Ti clustered together (cluster E). Most dissolved ions comprised one cluster (cluster B), except potassium that correlated with S and TOC (cluster G) and ammonium that did not cluster with any other compound (cluster F). Another cluster was formed by Si and Cr (cluster A). Finally, total potassium did not cluster with any other compound (cluster D) (Fig. 2).

A principal component analysis (PCA) was performed to reveal possible similarities among the locations. The first three axes of the PCA could explain 90.3% of the observed variation. Moreover, it showed that L1, L2 and L3 were similar in chemical composition (Fig. 3). Location L6 appeared to be closely related to L1–L3 based on the first and second principal component, but it was different on the third principal component (Fig. 3B). The higher ammonium level in L6 compared to all other locations possibly explains this difference. Although L4 was most comparable to L5, the latter differed from all other locations. Despite a slight contamination, L7 and L8 grouped with the non-contaminated L9.

#### Relationship between the microbial community and the chemical compounds

We identified the indigenous microbial community with 16S rRNA gene amplicon sequencing, which resulted in an average of 1099 OTUs per location, totalling to 6765 OTUs (Table S1 and Table S2). Alpha diversities (within sample diversity) were significantly lower in the contaminated locations compared to the control locations (Table S1). The five most relative abundant phyla were *Proteobacteria*, *Acidobacteria*, *Bacteroidetes*, *Actinobacteria* and *Verrucomicrobia* and ranged from 23.8% to 43.0%, 14.6% to 34.5%, 2.6% to 17.8%, 2.6% to 27.2%, and 2.4% to 10.2%, respectively (Fig. S3). 2.9% to 18.3% of the reads could not be classified at the phylum level. In general, no clear correlation between *Proteobacteria* classes and the measured compounds was observed (Fig. S4). L6 and L5 contained the most *Betaproteobacteria* followed by L8 and L9 (Fig. S5). The 20 *Acidobacteria* subgroups showed different correlation profiles (Fig. S4). The relative abundance of Gp18 positively correlated with all compounds from cluster B and all compounds from cluster C, except Ca and nitrate.



**Fig. 2.** Clustered heatmap of the measured compounds using Pearson's correlation coefficients showing seven distinct clusters. The heatmap is coloured according to the correlation coefficients: red means positively correlated ( $\rho > 0$ ), while blue means negatively correlated ( $\rho < 0$ ). The representative of each cluster is indicated with a green asterisk. Dissolved ions are written in full, whereas abbreviations are used to indicate total element concentrations.

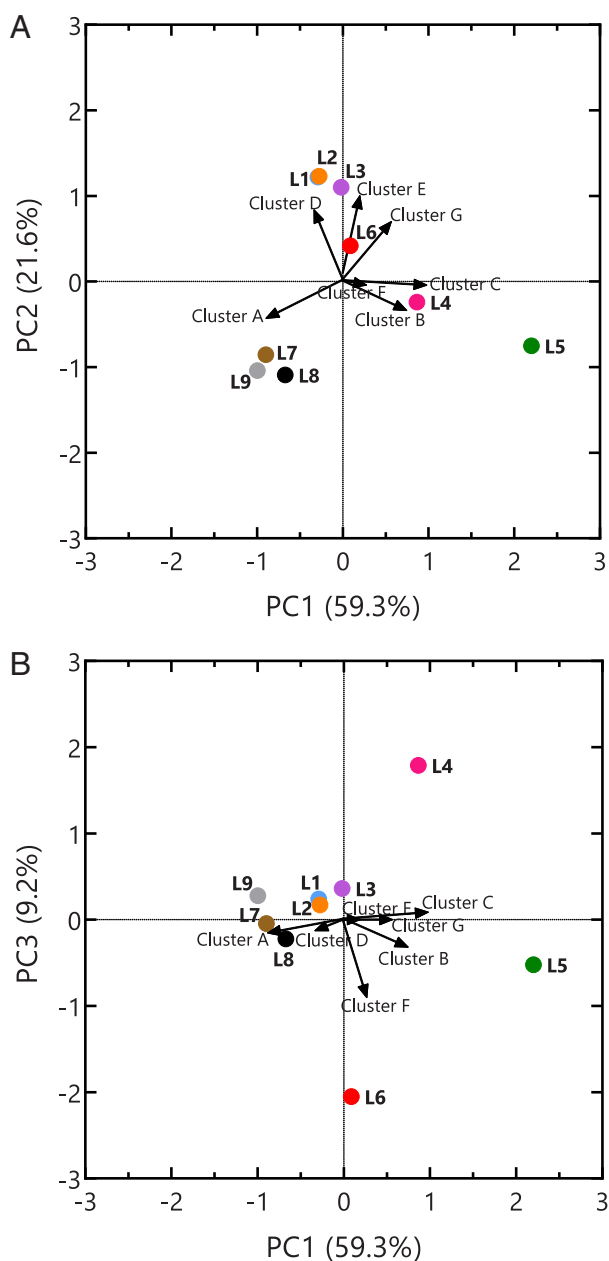
Subgroups Gp4 and Gp17 positively correlated with several compounds from cluster C, while the abundance of Gp5 negatively correlated with compounds from cluster B (all except chloride) and cluster C. *Verrucomicrobia* correlated negatively with compounds from cluster C, mainly due to the *Spartobacteria* class. Similarly, *Sphingobacteria* within the *Bacteroidetes* also correlated negatively with compounds from clusters B and C (Fig. S4). Notably, unclassified bacteria positively correlated with all compounds from Cluster C, except nitrate and  $^{210}\text{Pb}$  (Fig. S4).

To reveal possible differences within the bacterial community structure, an unconstrained RC(M) analysis on the OTU level was performed (Hawinkel *et al.*, 2019). This indicated clear differences in the bacterial community structure (Fig. 4A). Samples from a single location clustered closely together, indicating only little variation within each sample location. More variation was observed between the different sample locations. A gradual change was seen from locations L1 to L4, while L5 was more

secluded. The bacterial community of location L6 resembled the closest to the control locations L7, L8 and L9.

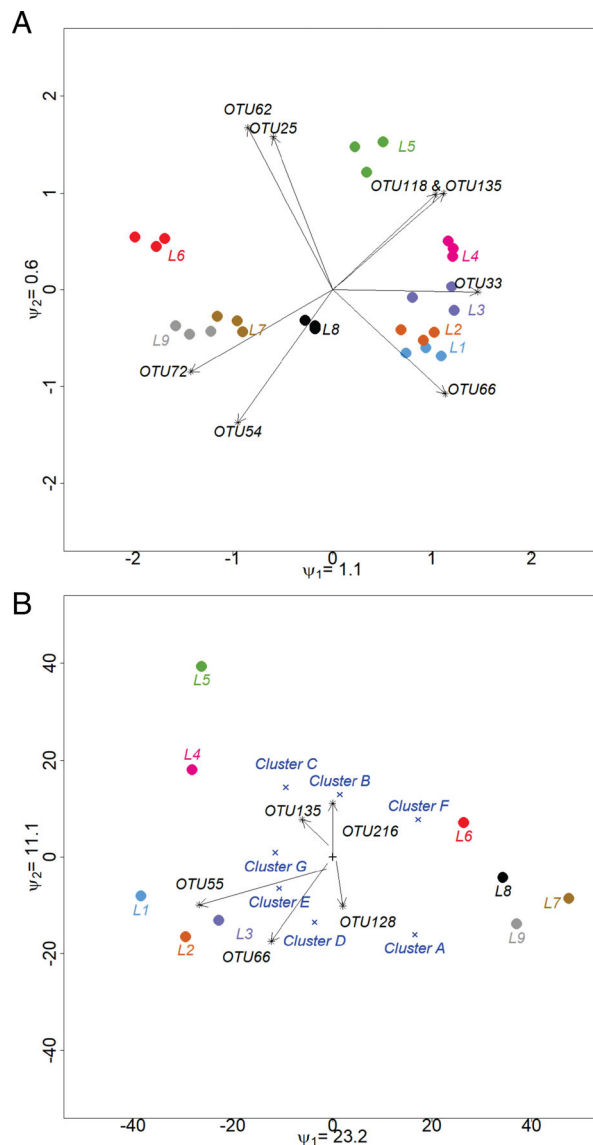
Constrained RC(M) illustrated that the observed differences in the microbial community could be explained by the metadata (Fig. 4B). Overall, a similar distribution can be seen as in the PCA of the metadata (Fig. 3) and the unconstrained RCM (Fig. 4A). Locations L1–L3 as well as L7–L9 were similar in both bacterial communities and soil characteristics. Although the contamination profile of L6 was more similar to that of L1–L3, its microbial community seemed to resemble that of the control locations (Fig. 4A). Location L5 appeared to be more unique when taking both OTUs and metadata into account, with L4 being the most closely related.

OTU55, OTU66, OTU128, OTU216 and OTU268 explain most of the variance within the dataset (Fig. 4B). OTU268 and OTU216 were exclusively present in L5 and L6 for all three replicates (Table S2). OTU268, an unclassified bacterial OTU, correlated negatively with cluster A and positively with cluster B, cluster C and cluster F



**Fig. 3.** Biplot showing the (A) PC1 and PC2 axes, and (B) PC1 and PC3 axes of a principal component analysis of all measured compounds demonstrating the different groups among the locations (L1–L9) based on the contamination. One representative compound from Fig. 2 was chosen to represent the cluster's orientation. Values of the contamination cluster are multiplied with the standard deviation of the respective principal component.

(Fig. S6). OTU216, member of the *Myxococcales* order, correlated with cluster B, cluster C (only  $^{210}\text{Pb}$ , moisture and phosphate) and cluster F. OTU55, belonging to the genus *Pedobacter*, correlated exclusively with cluster A and OTU66 (*Acidobacteria* subgroup Gp2) correlated with cluster D and Ti from cluster E. Lastly, OTU128, an unclassified *Betaproteobacterium*, was mainly present in L7, but did not correlate with any compound (Fig. S6).



**Fig. 4.** A. The unconstrained RC(M) biplot shows differences in the microbial community. The eight OTUs are more abundant than average in the locations to which the arrow points, and less abundant in the opposite direction. The importance parameters  $\psi$  (1.1 for the x-axis and 0.6 for the y-axis) reflect the relative importance of the dimensions. B. The constrained RC(M) triplot shows the relation between the microbial communities and the locations (L1–L9), and the relation between the chemical parameters and the identified microorganisms (cluster A–F). The abundance of the OTUs increases in the direction of the arrow and decreases in the opposite direction. The importance parameters  $\psi$  (23.2 for the x-axis and 11.1 for the y-axis) reflect the relative importance of the dimensions. OTU25: *Thiobacillus*; OTU33: *Bacteria\_unclassified*; OTU54: *Betaproteobacteria\_unclassified*; OTU55: *Pedobacter*; OTU62: *Burkholderiales\_unclassified*; OTU66: *Gp2\_unclassified*; OTU72: *Betaproteobacteria\_unclassified*; OTU118: *Bacteria\_unclassified*; OTU128: *Betaproteobacteria\_unclassified*; OTU135: *Gp4\_unclassified*; OTU216: *Myxococcales\_unclassified*.

Overall, the top 50 most explanatory OTUs belong to the five most abundant phyla (*Proteobacteria*, *Acidobacteria*, *Bacteroidetes*, *Actinobacteria* and *Verrucomicrobia*),

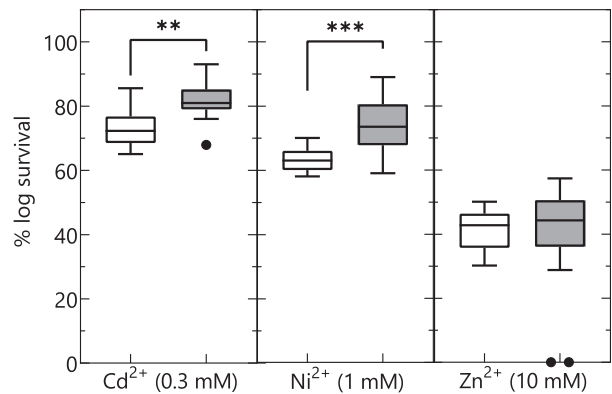
*Gemmatimonadetes*, *Chloroflexi* and unclassified bacteria reads. OTU139 (unclassified *Deltaproteobacteria*), OTU129 (unclassified *Betaproteobacteria*), OTU135 (*Acidobacteria* Gp4) and OTU118 (unclassified bacteria) correlated with all compounds from Cluster B and with all compounds from cluster C, except Ca and nitrate. OTU136 (*Acidobacteria* Gp4), OTU18 (*Acidobacteria* Gp1), OTU71 (*Sinobacteriaceae*) and OTU33 (unclassified bacteria) positively correlated with several compounds from cluster C. Contrary, OTU131 (*Chitinophagaceae*), correlated negatively with several compounds of cluster C and with cluster G, except for potassium. A negative correlation with cluster G was also observed for OTU6 (*Acidobacteria* Gp2), OTU87 (unclassified *Betaproteobacteria*) and OTU102 (*Roseiarcus*) (Fig. S6). Furthermore, 12 OTUs correlated exclusively with ammonium (Fig. S6). Lastly, OTU25 (*Thiobacillus*) and OTU101 (*Candidatus Koribacter*) correlated exclusively with ammonium and sulphate (Fig. S6).

#### Functional changes within the microbial community

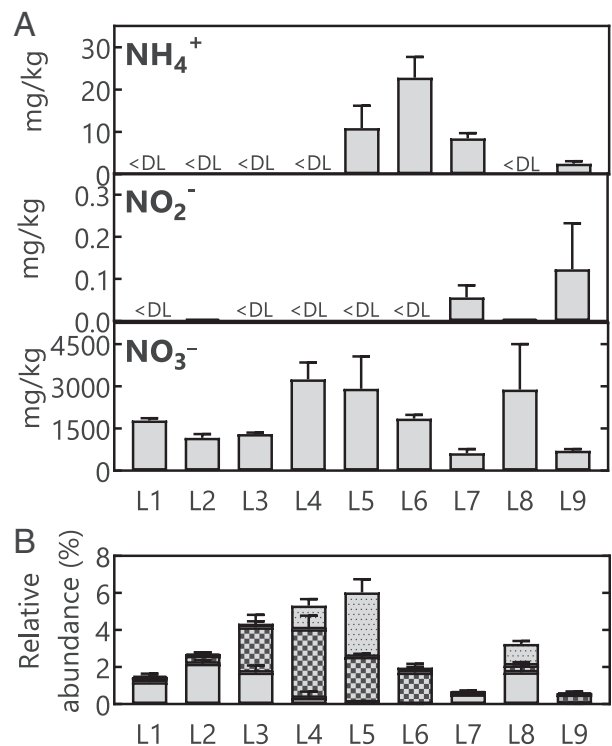
Total cell counts by flow cytometry and targeted Cd, Ni and Zn toxicity experiments were performed in an attempt to correlate differences in microbial biomass and the contamination level. No difference in total cell counts was observed between contaminated locations (L1–L6) and the control locations (L7–L9), ranging from  $2.18 \times 10^6$  to  $5.27 \times 10^7$  cells/g soil. Likewise, total viable count on R2A agar and a mineral salts medium (RM) supplemented with gluconate as sole carbon source were similar for both contaminated and control locations (Fig. S7). The latter was selected as it is highly suitable for screening metal resistance (Mergeay *et al.*, 1985).

Three metals were selected to identify whether the microbial community has adapted to their chemically perturbed environment. Half of the maximum measured total concentration of Zn, Cd and Ni was supplemented to RM mineral salts agar medium. The results indicated that a larger fraction of the microbial communities from the contaminated locations was able to grow on 0.3 mM Cd<sup>2+</sup> and 1 mM Ni<sup>2+</sup>, but not on 10 mM Zn<sup>2+</sup> (Fig. 5).

To investigate the effect of the contamination on nitrification, we first looked at correlations between nitrifying bacteria and the metal and radionuclide concentration profile. Interestingly, the abundance of known autotrophic nitrifying bacteria (predominantly *Nitrosospira* and *Nitrospira*, with only 3 reads of *Nitrosomonas* present in 1 out of 27 samples) resembled the contamination profile (Fig. 6B). A significant positive correlation with 20 out of the 21 compounds (except <sup>210</sup>Pb) of cluster C was observed, with Cd showing the highest correlation (Table S3). In addition, the OTU composition of the nitrifying bacteria changed among the locations (Fig. 6B). To



**Fig. 5.** Survival of the microbial community from the control (L7–L9) (white) and contaminated (L1–L6) (grey) locations on RM agar supplemented with 0.3 mM Cd<sup>2+</sup>, 1 mM Ni<sup>2+</sup> or 10 mM Zn<sup>2+</sup> visualized in boxplots. The box has lines at the lower quartile, median and upper quartile values. The whiskers are lines extending from each end of the box with a maximum length of 1.5 times the interquartile range. Outliers are data with values beyond the end of the whiskers and plotted with a black circle. Values represent percent log<sub>10</sub> survival compared to RM agar. Significant differences obtained with unequal variances T-test between control and contaminated locations are shown with \*\* if  $p < 0.01$  and \*\*\* if  $p < 0.001$ .



**Fig. 6.** A. Overview of the dissolved NH<sub>4</sub><sup>+</sup>, NO<sub>2</sub><sup>-</sup> and NO<sub>3</sub><sup>-</sup> levels in all locations. Values represent the average and standard deviation of 4–6 technical replicates. B. Relative abundance of the sum of all reads belonging to the dominant OTUs representing *Nitrosospira* and *Nitrospira* in all locations. *Nitrosospira* (OTU19) is shown as dotted bar, and *Nitrospira* is presented as full bar (OTU15) and blocked bar (OTU103). Data represent the average and standard deviation of three biological triplicates.

explicate the relationship between nitrification and metal contamination across our sample locations, we assessed the correlations between the relative abundances of nitrifying bacteria with ammonia, nitrite and nitrate levels (Fig. 6A and B). A clear relationship between the relative abundance of nitrifying bacteria and the differences in ammonium, nitrite and nitrate levels was only found for the control locations L7–L9. Ammonium was below the detection limit in L8, while L7 and L9 contained  $8.5 \pm 1.2$  and  $2.5 \pm 0.6$  mg dissolved  $\text{NH}_4^+$ /kg dry soil, respectively. Moreover, a low level of nitrite could only be measured in L7 and L9, and the nitrate level was higher in L8 compared to L7 and L9 (Fig. 6A). Consequently, ammonium could be oxidized faster to nitrate in L8 compared to L7 and L9. Indeed, these findings coincide with a higher abundance of *Nitrosospira*, a known genus involved in ammonium oxidation, and *Nitrospira*, a genus known to be involved in autotrophic nitrite oxidation, in L8 ( $1.16\% \pm 0.15\%$  and  $2.12 \pm 0.49\%$ ) compared to L7 ( $0.11\% \pm 0.05\%$  and  $0.58\% \pm 0.04\%$ ) and L9 ( $0.08\% \pm 0.09\%$  and  $0.51\% \pm 0.08\%$ ) (Table S2).

The relationship between N compounds and nitrifying microorganisms was less clear for the contaminated samples. Ammonium was only detected in L5 and L6. Nitrite was not detected in the contaminated locations, except in L2. Nitrate levels were the highest in L4 and L5, and those of L1, L2, L3 and L6 were similar to each other (Fig. 6A). The build-up of ammonium in L6 could be explained by the low abundance of *Nitrosospira* ( $0.27\% \pm 0.19\%$ ), similar to that observed for L7 and L9. However, the abundance of *Nitrosospira* is the highest in L5 ( $3.48\% \pm 0.71\%$ ), where the ammonium level is similar to that of the control site L7 (Fig. 6).

## Discussion

### *Relationship between the microbial community and the chemical compounds*

Areas of the Grote Nete river basin (Belgium) are contaminated by historical discharges from nuclear industries and from the phosphate industry. This resulted in a unique contamination complexity of natural and artificial radionuclides ( $^{238}\text{U}$ ,  $^{232}\text{Th}$ ,  $^{228}\text{Th}$ ,  $^{210}\text{Pb}$ ,  $^{226}\text{Ra}$ ,  $^{228}\text{Ra}$ ,  $^{60}\text{Co}$ ,  $^{137}\text{Cs}$  and  $^{241}\text{Am}$ ), metals (Co, Ni, Cu, Zn, As, Cd, Hg, Pb and Cr) and other ions (e.g.  $\text{NO}_3^-$ ,  $\text{NO}_2^-$ ,  $\text{NH}_4^+$ ,  $\text{PO}_4^{3-}$ ,  $\text{SO}_4^{2-}$ ,  $\text{Ca}^{2+}$ ,  $\text{Cl}^-$ ) in areas prone to flooding (because of acute flooding and tidal movement) (Paridaens and Vanmarcke, 2001; FANC, 2019). We sampled eight locations (L1–L8) with varying contamination levels and one adjacent unpolluted location (L9) to investigate the impact of this long-term contamination on the residing microbial community. Locations L1–L6 were clearly contaminated, L5 was a contamination hot spot,

while L7, L8 and L9 can be seen as control locations (Figs 1 and 3).

Taxonomic profiling indicated a high number of OTUs in each sample as can be expected for soil bacterial communities (Torsvik *et al.*, 1996). Moreover, the five most relative abundant phyla of this study (*Proteobacteria*, *Acidobacteria*, *Bacteroidetes*, *Actinobacteria* and *Verucomicrobia*) are typically observed in soil (Janssen, 2006; Gołębiewski *et al.*, 2014; Li *et al.*, 2017). The significant co-contamination of metals and radionuclides could have a synergistic toxic impact on the indigenous microbial community affecting diversity and richness. This could explain the lower alpha diversity in the contaminated locations (L1–L6) compared to the control locations (L7–L9) (Table S1), contrary to several studies where metal contamination did not affect bacterial diversity and evenness (Azarbad *et al.*, 2015; Sitte *et al.*, 2015; Li *et al.*, 2017; Beattie *et al.*, 2018).

The community structure varied among the locations and was defined by the chemical parameters (Fig. 4b). Locations L5 and L4 contain the first and second highest levels of most metals and radionuclides, respectively, and were the most unique when taken into account both OTUs and metadata. This could indicate that the microbial community was affected by the contamination level. The ammonium level was the highest in L6 and 12 out of the 50 most explanatory OTUs correlated exclusively with ammonium (cluster F) (Fig. 1 and Fig. S6), suggesting an important influence of ammonium on the microbial community. However, if the ammonium level was solely responsible for the observed differences, L5 should cluster with L7 as levels were similar in both locations (Fig. S2). Metal and radionuclide levels in L6 were at least in the range of L1–L3, except for Co and Zn (Fig. 1), suggesting that Zn and Co could be important drivers in shaping the microbial community. Nonetheless, it remains difficult to find a causal effect due to a single compound since a lot of co-correlation was observed among the chemical parameters. The synergistic effects of metals on the bacterial community can be more pronounced than those of the single metals (Song *et al.*, 2018). Moreover, the bioavailability of the different compounds could differ among the sample locations. In general, the bioavailability of different metals and radionuclides depends strongly on their speciation, the local geochemistry, soil solution and the soil microbiome. A complex interplay between abiotic and biotic reactions, affected by the physico-chemical parameters in the soil (pH,  $E_h$ , organic matter type and content) and the soil microbiome control the soil-solution equilibrium and bioavailability of each metal and radionuclide (Tack, 2010). Microorganisms can directly and/or in concertation with the physico-chemical soil parameters influence metal and radionuclide mobility, toxicity and distribution. Key

processes are reduction, uptake and accumulation by cells, biosorption, complexation with proteins, polysaccharides and microbial biomolecules, and biomineralization with phosphates and carbonates (Groudev *et al.*, 2001; Lloyd and Lovley, 2001; Lloyd, 2003; Ruggiero *et al.*, 2005).

Unclassified bacteria were highly present and positively correlated with all compounds from cluster C, except for nitrate and  $^{210}\text{Pb}$  (Fig. S4). This could be due to the fact that these mixed polluted environments are not well characterized. From the classified reads, several of the most explanatory OTUs positively correlated with the contaminants. OTU129, an unclassified *Betaproteobacteria* positively correlated with cluster B and 14 out of 21 compounds from cluster C. However, in general, no clear correlation of the *Betaproteobacteria* class was found (Fig. S4) although this class has been reported to be more abundant under metal contamination (Jacquiod *et al.*, 2018). This could be explained by the presence of 195 OTUs that correspond to unclassified *Betaproteobacteria* and exhibit different abundance profiles (Fig. S3).

*Acidobacteria*, suggested to be metabolically diverse like *Proteobacteria* (Kielak *et al.*, 2016), are often associated with a slightly acidic pH (Kielak *et al.*, 2016), which could explain their high abundance in this meadow. Of the most explanatory OTUs, OTU135 and OTU136 (subgroup Gp4) were positively correlated with compounds from cluster C. In general, the relative abundance of Gp17 correlated positively with 12 compounds from cluster C. Gp18 correlated positively with cluster B and cluster C (except Ca and nitrate), while Gp5 correlated negatively with the cluster B (except chloride) and 6 compounds from cluster C (Fig. S4). Gp17 was also more abundant in soils with higher multi-metal contamination as compared to lower (Hg, As, Co, Cd, Cr, Ni, Pb, Cu, Mn and Zn), indicating a possible general trend for this subgroup to be metal resistant (Yin *et al.*, 2015). Metal resistance genes have been found in this phylum (Thomas *et al.*, 2020), but further studies are necessary to identify their role in metal interactions. Nevertheless, these results indicate that this phylum is clearly diverse in their response to environmental factors.

The higher abundance of several OTUs in the most contaminated locations could indicate that resistance mechanisms are more prevalent in these locations. Furthermore, possible indirect mechanisms could also be at play. OTU216, member of the *Myxococcales* order, was exclusively present in L5 and L6 (Table S2) and showed the strongest correlation with sulphate (cluster B). This order has been reported as potentially sulphate-reducing bacteria (Reese *et al.*, 2014), resulting in the production of  $\text{HS}^-$ . On its turn,  $\text{HS}^-$  can result in the formation of insoluble metal-sulphide complexes of dissolved metals

such as  $\text{Cd}^{2+}$ ,  $\text{U}^{6+}$ ,  $\text{Ni}^{2+}$  and  $\text{Co}^{2+}$  (Yin *et al.*, 2019). On the other hand, OTU25, a sulphur-oxidizing betaproteobacterium *Thiobacillus* is also mainly present in L5 and L6. Sulphur oxidation can result in leaching of metal-sulphide complexes (Schippers and Sand, 1999). Interplay between these two sulphur-based mechanisms and the physico-chemical parameters could lead to different equilibria in the different locations.

Besides an enrichment of several groups of bacteria in the most contaminated locations, several OTUs are absent in these locations. OTU131, an unclassified *Chitinophagaceae*, correlated negatively with compounds from cluster C and is absent in L4 and L5 (Fig. S6, Table S2). Members of the *Chitinophagaceae* family have previously been suggested as indicators of environmental perturbations (Hermans *et al.*, 2017). The *Sphingobacteria* class was negatively correlated with compounds from clusters B and C and has been shown to be less abundant in more contaminated soils (Yin *et al.*, 2015). Furthermore, OTU102 (*Roseiarcus*) was exclusively present in L6–L9 (Table S2) and correlated negatively with potassium, S and TOC (Fig. S6). The role of this genus is still largely unknown and only one strain has been characterized (Kulichevskaya *et al.*, 2014). Putatively, OTU102 is highly sensitive to Co and Zn, explaining its absence in L1–L5. Similarly, OTU213 of the *Verrucomicrobia* class *Spartobacteria* was solely found in L6–L9 and significantly correlated with ammonium (Fig. S6). Overall, *Spartobacteria* correlated negatively with compounds from cluster C (Fig. S2) and have recently been found sensitive to toxic elements in antimony mining sites (Deng *et al.*, 2020). This could explain the negative correlation and complete absence of some of the OTUs in the contaminated locations.

#### Functional changes within the microbial community

A reduction in microbial biomass has been observed in metal-contaminated environments (Oliveira and Pampulha, 2006; Zhang *et al.*, 2016; Song *et al.*, 2018), however, the correlation between microbial biomass and metal contamination is not always straightforward (Gillan *et al.*, 2005; Rousk and Rousk, 2018). In this study, no difference in total and viable cell counts were observed between contaminated locations (L1–L6) and the control locations (L7–L9).

Three metals,  $\text{Zn}^{2+}$ ,  $\text{Cd}^{2+}$  and  $\text{Ni}^{2+}$ , were selected to identify whether the microbial community has adapted to their chemically perturbed environment.  $\text{Zn}^{2+}$  was chosen as concentrations were the highest of all metals and radionuclides and it was identified as one of the possible drivers explaining the observed differences in the microbial community.  $\text{Cd}^{2+}$  was selected as it has been well-documented to be highly toxic for living organisms



(Zhang and Reynolds, 2019) and  $\text{Ni}^{2+}$  was chosen as no difference in Ni levels were observed between contaminated locations L2 and L6 and control locations L7–L9 (Fig. 1). The survival of microbial communities from the contaminated locations on 0.3 mM  $\text{Cd}^{2+}$  and 1 mM  $\text{Ni}^{2+}$  was higher compared to that of the control locations but no difference was observed on 10 mM  $\text{Zn}^{2+}$  (Fig. 5). Although only a few singular compounds were selected, these results suggest that the microbial community from the contaminated locations is adapted to the *in situ* conditions, which other studies also showed in similar experiments (Turpeinen *et al.*, 2004; Margesin *et al.*, 2011).

These metals were discharged by phosphate industry, which could explain why soluble phosphate and total phosphorus correlated together with the metal and radionuclide contaminants (Fig. 2). Furthermore, the presence of phosphate in the soil groundwater could shift bacterial resistance mechanisms more towards intra- and/or extracellular phosphate mediated complexation resulting in the precipitation and immobilization of metals and radionuclides. Although phosphate mediated complexation has been shown for  $\text{Cd}^{2+}$ ,  $\text{Zn}^{2+}$  and several other metals and radionuclides (Basnakova *et al.*, 1998; Andrade *et al.*, 2004; Renninger *et al.*, 2004; Zeng *et al.*, 2020), the best known  $\text{Ni}^{2+}$  resistance mechanism is efflux-mediated resistance (López-Maury *et al.*, 2002; Nies, 2003; Mergeay and Van Houdt, 2015). Nevertheless, our experimental setup did not allow to interpret the individual effect of total phosphorus or phosphate on the microbial community and thus no clear impact was observed.

To further explore the previous observations involving microbial community structure, functionality and metal/radionuclide toxicity, we focused on nitrification, a key soil ecosystem process that is widely reported to be susceptible to metal toxicity (Rother *et al.*, 1982). Overall, although we observed a correlation between the relative abundance of nitrifying bacteria and metal contamination, no significant correlations could be made between nitrifying bacteria and the levels of ammonium, nitrate and nitrite in the contaminated locations, in contrast to the control locations. The composition of nitrifying bacteria changed across locations, therefore an adaptation in activity could have followed. This was shown previously by Mertens *et al.* (2009), who demonstrated that a change in OTU composition of nitrifying bacteria was responsible for restoration of nitrification after Zn contamination. This suggests that nitrification was affected by the contamination, however, it is difficult to elucidate the detailed ongoing mechanism. More experimental data are necessary to corroborate the results and to confirm or refute hypotheses.

### Conclusion

Soil contamination with metals and radionuclides is an important problem and urges efficient remediation

techniques. Bioremediation is a promising cost and eco-friendly strategy to be used *in situ* but requires the presence of metal-resistant microbial communities. Understanding how metal and radionuclide toxicity influences ecosystem health necessitates studies on the microbial ecology of such contaminated sites. This study indicates that long-term metal and radionuclide contamination affects the microbial community structure and functionality and provides important fundamental insights into microbial community dynamics in co-metal-radionuclide contaminated sites. The strong correlation between some OTUs and specific soil parameters could imply their use as bioindicators for changing conditions. Furthermore, our approach could facilitate identifying sites for targeted isolations of adapted strains towards bioremediation applications.

## Experimental procedures

### Sample collection

We organized one sampling campaign on June 7, 2018. We targeted to have samples with a contamination gradient, hence we sampled eight locations (L1–L8) along a transect to the Grote Nete river with L1 and L8 being the closest and furthest, respectively. In addition, one uncontaminated location (L9) was sampled approximately at the same distance from the river as L8 but located in a large uncontaminated area within the same field (based on aerial surveys of gamma radiation (Paridaens *et al.*, 2016)) (Fig. S1B). We measured the external gamma radiation with a dose rate meter (Automess Scintillator Probe 6150AD-b) (mainly originating from the radionuclide  $^{226}\text{Ra}$ ) in the field during sampling to have a first indication of the contamination level and to assure that we sampled various contaminated and uncontaminated locations.

For microbiological analysis, three samples were taken within a 40 × 40 cm square at each location (Fig. S1C). To this end, the upper 10 cm of the soil was taken after removal of the root mat. To avoid external contamination, all sampling equipment was sterilized by autoclaving. In addition, cross-contamination between samples was prevented by using different samplers at each location. All samples were sealed and stored in the dark at 4 °C.

For all chemical analyses, one additional sample was taken at each location (in the same 40 × 40 cm square), including the soil around and in between the three samples. These samples were air-dried in a greenhouse and 2 mm sieved before analyses.

### Analyses of the soil chemistry

General soil characteristics such as pH, moisture content and the total organic and total inorganic carbon were determined for each sampling point. Furthermore, we

analysed the total and dissolved concentration of several major ions and the total concentrations of multiple radionuclides and metals. Detailed protocols of the different analyses can be found in Supporting Information. In brief, the concentrations of the radionuclides  $^{137}\text{Cs}$ ,  $^{210}\text{Pb}$ ,  $^{226}\text{Ra}$ ,  $^{228}\text{Ra}$ ,  $^{60}\text{Co}$ ,  $^{241}\text{Am}$ ,  $^{238}\text{U}$ ,  $^{228}\text{Th}$ , and  $^{232}\text{Th}$  were determined at SCK CEN LRM accredited conform to ISO-17025. Element analysis (Na, Mg, Al, Si, P, K, Ca, Ti, Cr, Mn, Fe, As, Cd, Zn, Pb, Co, Ni, Hg, Cu, Cl and S) was carried out by a certified lab of VITO (Mol, Belgium).

Cations ( $\text{Li}^+$ ,  $\text{Na}^+$ ,  $\text{NH}_4^+$ ,  $\text{K}^+$ ,  $\text{Mg}^{2+}$ , and  $\text{Ca}^{2+}$ ) in soil solution were measured with a Dionex ICS2500 ion chromatography system packed with an IonPac CG12A guard column and IonPac CS12A analytical column. Anions ( $\text{NO}_3^-$ ,  $\text{F}^-$ ,  $\text{Cl}^-$ ,  $\text{NO}_2^-$ ,  $\text{SO}_4^{2-}$  and  $\text{PO}_4^{3-}$ ) were analysed with an IonPac AS11-HS anion exchange column.

To reveal possible correlations between the different chemical parameters, a Pearson-correlation-based clustered heatmap of the metadata was created with the R package *pheatmap* (version 1.0.12) (Kolde, 2019).

To illustrate similarities among the locations, a principal component analysis was performed on normalized metadata (average 0, standard deviation 1) using the *prcomp* function in R (version 3.6.2) (R Core Team, 2019). Coordinates of the locations were divided by the standard deviation of the respective principal component and extracted for visualization. One representative chemical parameter of each cluster identified based on Pearson correlations was used to visualize the orientation of the clusters. The result of the PCA analysis was used for the detailed identification of contaminated and control locations. The latter were identified as those similar in chemical composition as L9.

#### Analysis of the soil microbiome

**DNA extraction.** Five grams of soil were used to extract DNA with the DNeasy PowerMax Soil Kit (Qiagen, The Netherlands) according to the manufacturer's protocol. DNA was extracted within 14 days of storage on 4 °C, which is not expected to influence the results (Lauber *et al.*, 2010). In addition, to further exclude possible storage effects, DNA of the different replicates was extracted on various days. A negative control (i.e. an empty tube) was included to identify possible contaminations originating from the extraction kit or other downstream applications. All samples were subjected to RNase A treatment (1 U/100 µl) (Qiagen, The Netherlands) and extensively purified as soil samples are known to contain several PCR inhibitors and thus could interfere with 16S rRNA gene sequencing (Schrader *et al.*, 2012). Details regarding the purification steps are given in Supporting Information. Finally, the concentration and integrity of the DNA was analysed with the Quantifluor dsDNA sample kit

(Promega, the Netherlands) and gel electrophoresis, respectively.

**16S rRNA gene sequencing.** The verified DNA was sent to Baseclear (Leiden, the Netherlands) for high-throughput amplicon sequencing of the V3-V4 hypervariable region of the 16S rRNA gene with the Illumina MiSeq platform according to the manufacturer's guidelines. Data were processed using the OCToPUS pipeline (Mysara *et al.*, 2017), which consists of the following steps: quality filtering using HMMER, merging reads using the *make.contigs* command from *mothur*, alignment and filtering of the merged reads following the Standard Operating Protocol (SOP) procedure (v.1.39.0) as described by the authors of *mothur*, error correction using IPED (Mysara *et al.*, 2016), chimera identification using CATCh (Mysara *et al.*, 2015), and OTU clustering using UPARSE (Edgar, 2013). The negative control (i.e. DNA extraction from the empty tube) only had 847 reads and contained eight OTUs. However, none of these were present in all nine locations, indicating that a significant contamination originating from the DNA extraction kit was absent (Fig. S8). Samples were subsampled to a depth of 10,198 reads (resembling smallest sample's depth), which adequately represented the bacterial diversity in the samples (Fig. S9). The datasets generated and analysed during this study are available in the NCBI Sequence Read Archive (SRA) repository (PRJNA630593).

Alpha diversity indices (sobs, Chao1, and Shannon) were calculated using the *mothur summary.single* command index.

The relations between the microbial communities, the sample locations and the chemical parameters were analysed with both the unconstrained and the constrained RC(M) method, using the *RCM* R package (version 1.1.0.) (Hawinkel *et al.*, 2019). Both methods are exploratory ordination methods, which in contrast to other ordination methods use OTU counts instead of OTU frequencies. The unconstrained RC(M) method quantifies the deviation of a given OTU in a sample from all samples, while ignoring all other sources of information, such as the level of all measured compounds. The constrained RC(M) method allows to visualize the variability of the OTU counts that can be explained by the changes in the chemical parameters at every location. To this end, one representative chemical parameter of each cluster based on Pearson correlations (section "Functional changes within the microbial community") was chosen. The RC(M) placed the three biological replicates on each other since only one measurement was performed to obtain the metadata. Both methods used 330 of the most variable OTUs (variance larger than 75).

**Laboratory experiments.** Two grams of soil sample were dissolved in 18 ml 10 mM MgSO<sub>4</sub>, vertically rotated for 2 h on 4 °C followed by 1 h on 4 °C standing upright for phase separation. The upper phase was then taken by gentle pipetting to achieve a representative subsample of the actual soil microbial community.

First, we compared the biomass of these subsamples between contaminated and control locations by estimating the total number of bacteria with flow cytometry (Supporting Information) (Props *et al.*, 2016).

Afterwards, we compared the viable count between control and contaminated locations by plating serial tenfold dilutions on R2A medium (Reasoner and Geldreich, 1985) and a mineral salts RM Noble Agar (1.5%, Sigma Aldrich, Belgium) medium supplemented with 0.2% (w/v) gluconate as sole carbon source (Mergey *et al.*, 1985). Finally, metal resistance was compared between communities originating from contaminated and those from control locations by plating serial tenfold dilutions on RM agar supplemented with one of three key contaminants: 10 mM ZnSO<sub>4</sub>, 1 mM NiCl<sub>2</sub> or 0.3 mM CdCl<sub>2</sub>. Plates were incubated aerobically at 30 °C in dark conditions and colony forming units were counted after 7 days. The log<sub>10</sub> of the viable count on RM agar supplemented with metals divided by the log<sub>10</sub> of the viable count on RM agar medium was used to calculate survival on the specific metals.

#### Statistical analysis

Unequal variances T-test was used to test statistical differences between contaminated and control locations. The Mann–Whitney U test was used if normality could not be assumed based on the Shapiro–Wilk test. Pearson correlations between biological data and the metadata were calculated with R package Hmisc (version 4.3-0) (Harrell Jr *et al.*, 2019). A significance level of 0.05 was used throughout the analyses. All analyses were performed with R (version 3.6.2) (R Core Team, 2019).

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## Supporting Information

Additional Supporting Information may be found in the online version of this article at the publisher's web-site:

### Appendix S1: Supporting Information.

**Figure S1** A. Visualization of the origin of contamination from two distinct industries. Nuclear industry discharged in the Molsse Nete (red line) and phosphate industry discharged in the Grote Laak (green line). Both rivers flow downstream in the Grote Nete river (blue line) and contaminated river banks in the black circle. The map originated from Geopunt Vlaanderen and was adapted based on information from the Federal Agency for Nuclear Control (FANC). B. Schematic representation of the sample locations. Locations L1–L6 were located in the meadow, L7 was on the border of this meadow with a mown field where L8 and L9 were situated. The map was adapted from Google Maps. C. Illustration of the sample method in the sterile manner.

**Fig. S2:** Overview of the general soil characteristics, major ions and total element concentrations measured on the different locations (L1–L9). Moisture, TOC and most total element concentrations are displayed in % (w/w). All ions and total S are presented in mg/kg. Error bars for pH, Moisture and all ions represent variation on technical replicates. Error bars of total element concentrations represent measurement uncertainty. Values below detection limit are indicated with <DL. The detection limit is shown as a dotted line.

**Fig. S3:** Overview of the relative abundance of the top 10 most abundant phyla after subsampling for all 27 samples. Data are presented in relative abundance (%). A legend is provided for the different phyla.

**Fig. S4:** Pearson correlation matrix between the relative abundance of the five most abundant phyla and their classes and unclassified Bacteria. Phyla and classes are listed according to decreasing relative abundance. Significant Pearson correlations ( $p < 0.05$ ) are coloured from red (positive) to blue (negative). Non-significant Pearson correlations are shown in grey.

**Fig. S5:** Overview of the relative abundance of the top 15 most abundant classes after subsampling for all 27 samples. Data are presented in relative abundance (%). A legend is provided for the different classes.

**Fig. S6:** Pearson correlation matrix between the relative abundance of the 50 most explanatory OTUs and the metadata. Significant Pearson correlations ( $p < 0.05$ ) are coloured from red (positive) to blue (negative). Non-significant Pearson correlations are shown in grey.

**Fig. S7:** Boxplot showing the total cell count in control samples (L7–9) (white) compared to contaminated soil samples (L1–6) (grey) determined with Flow Cytometry (left panel), viable count on mineral salts RM agar medium (middle panel) and on R2A agar (right panel). The box has lines at the lower quartile, median and upper quartile values. The whiskers are lines extending from each end of the box to show the extent of the rest of the data, with a maximum length of 1.5 times the interquartile range. Values represent the  $\log_{10}$  of cells/g soil.

**Fig. S8:** Comparison of the OTUs found in the negative (empty tube) control (NC) across all samples (L1.1–L9.3). The total amount of reads was 847 for the NC while all other samples contained >10,000 reads. The number of reads is presented on a 2-segment linear scale ranging from 0 to 1000 and 8000 to 22,000 reads.

**Fig. S9:** Rarefaction curves of all samples showing the number of OTUs in function of the number of reads, subsampling was performed on 10,198 reads (dotted line).

**Table S1** Overview of the 16S read processing in number of raw and high quality reads, subsampling depth and alpha diversities.

**Table S2:** Taxonomy of all OTUs and their number of high quality reads for all 27 samples after subsampling.

**Table S3:** Overview of the significant ( $p < 0.05$ ) Pearson correlations between the sum of the relative abundance of all nitrification OTUs and the metadata. The values shown are the Pearson correlation coefficients ( $r$ ).