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Metagenomic insights into density-driven plastic stratification in open dumpsites: implications for waste management and bioremediation

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Open dumpsites in developing countries constitute a critical nexus of plastic contamination and microbial ecology. However, the mechanistic interplay between polymer density-driven stratification and niche-specific microbial adaptation remains unexplored. This study presents the first integrated metagenomic investigation linking density-mediated plastic segregation to depth-stratified microbial community assembly and metabolic functionality in a tropical open dumpsite. Depth profiling revealed that low-density polyethylene dominated the layers closer to the surface, whereas polypropylene and polyethylene terephthalate were concentrated in the bottom layers. A stratification pattern was subsequently detected in adjacent soil microplastic profiles, indicating density-dependent dispersal mechanisms. Metagenomic sequencing revealed distinct microbial niches aligned with plastic layers: surface zones favoured aerobic *Proteobacteria* and *Actinobacteria*, whereas anoxic depths enriched anaerobic taxa and pathways for methane metabolism, alternative carbon fixation, and dissimilatory processes. These findings highlight plastic density as an important driver of vertical microplastic migration and indigenous microbial community structuring, informing predictive modelling of contamination risk trajectories and the design of depth-targeted bioremediation strategies using adapted native consortia. This work advances fundamental understanding of density-driven ecological structuring in anthropogenic waste systems while providing an evidence-based framework for optimizing site-specific waste stabilization and developing indigenous microbial solutions for plastic remediation in regions dependent on uncontrolled dumping.

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Environmental significance

Open dumpsites in developing countries are major, yet understudied, sources of plastic pollution that disrupt soil ecosystems and threaten environmental health. Understanding how plastic types and microbial communities distribute and interact vertically within these waste mounds is critical to addressing pollution impacts and devising effective bioremediation strategies. This study reveals significant depth-dependent stratification of plastics and associated bacterial communities accompanied by distinct metabolic functions linked to oxygen gradients. Key findings show that native microbial populations adapt to environmental conditions to potentially degrade plastics, offering promising avenues for targeted bioremediation. These insights highlight the urgent need to incorporate microbial ecology into waste management frameworks, helping mitigate plastic contamination and restore soil ecosystem functions globally, especially in regions burdened by uncontrolled dumping.

1. Introduction

Plastic pollution is widely recognized as a critical global challenge demanding immediate action. Since 1950, over 9 billion tonnes of plastic including additives and synthetic fibres have been produced,¹ yet an estimated 23 million metric tonnes (MMT) entered the aquatic systems since 2016 alone, with

projections reaching 53 MMT by 2030.² Despite worldwide efforts to reduce plastic pollution through initiatives like the “Beat Plastic Pollution” campaign by the United Nations Environment Programme (UNEP), drawing attention towards the root causes of MPW still remains a challenge.

The burden of MPW, dominated by packaging waste, is particularly severe in developing regions with inadequate waste management infrastructure.³ Global assessments highlight that land-based sources contribute over 80 percent of marine plastic inputs *via* riverine and atmospheric pathways.^{4,5} In India, only 60 percent of the 3.5 MMT of annual plastic waste is collected; the remainder accumulates in over 2400 active dumpsites. According to the annual CPCB report 2021–22, West Bengal

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Fig. 1 (a) Sites for the study. (b) Method for extracting plastic waste from the dump site.

Samples from each layer of each mound were mixed and used for metagenomic analysis ($n = 3$).

2.3.2 Soil processing. The soil sample was sieved with a 365 μm sieve and plastic films and fibres that were visible in the soil and sieve were separated manually. These plastic particles (1 mm to 100 μm in size) were then characterized using an FTIR and matched against the Bruker Polymer Library. Plastic waste samples were compared with reference plastic samples and characterized based on >80% similarity. Soil samples were then characterized for pH, electrical conductivity, and moisture and then treated to remove any organic matter by heating at 400 $^{\circ}\text{C}$ and H_2O_2 digestion. The samples were then stained using Nile Red to visualize small microplastic particles that could not be removed manually. Stained samples were viewed under a fluorescence microscope (Motic BA400, Hong Kong, China) to confirm the presence of small microplastics (<365 μm).

2.4 Physical characterisation

2.4.1 Soil pH, electrical conductivity, and moisture. Electrical conductivity and pH of soil were recorded with the help of a conductivity meter (Hanna Instruments, Rhode Island, USA). Moisture content was calculated based on the change in percent weight between dried and original samples. Samples were dried at 50 $^{\circ}\text{C}$ to constant weight. All processes were carried out following ASTM methods – ASTM D4972, ASTM D7100, and ASTM D2216.

2.4.2 Plastic waste. Plastic wastes (macroplastics and microplastics) were sorted visually and isolated from each layer, and this process was repeated ten times. All plastic particles were characterised using an FTIR and the Bruker Polymer Library, and four different types of plastics (PP, PE, PET, and PS) in groups of 10 were segregated from each layer of each mound ($n = 12/\text{layer}$), resulting in a grand total of 360 samples from the waste mounds. One way ANOVA was carried out with Microsoft Excel to establish the significant difference in the abundance of plastic types across layers.

2.5 Metagenomic analysis

Two grams of 2 soil samples each ($n = 2$) and 4 grams of plastic waste ($n = 5$) were used to extract bacterial cells. The microbial community was characterized using 16S rRNA sequencing. Using the primers 16SF (5'AGAGTTTGATGTTGGCTCAG3') and 16SR (5'TTACCGCGGCMGCSGGCAC3'), the bacterial V3–V4 region of the 16S rRNA gene was amplified and sequenced using the Illumina MiSeq platform, with a read length of 2×300 bp paired end. An average read count of ~ 0.1 million bp was maintained for all samples. Bioinformatics analysis and quality check of the demultiplexed data were conducted using FastQC and MultiQC, while the adapters and the low-quality reads were trimmed using TRIMGALORE, which includes merging paired-end reads, chimera removal, operational taxonomic unit (OTU) abundance calculation, and estimation correction. This was achieved using Uclust (QIIME 2 pipeline), which enables highly accurate investigations at the genus level. Taxonomic classification was done using the SILVA database. KEGG pathway analysis was carried out using PICRUST.¹⁷

Data filtering was done to remove low quality features and improve downstream statistical analysis. Data were further scaled using the Total Sum Scaling (TSS) factor to normalise information. In this study, various biodiversity assays were used to shed light on the microbial communities that colonise and adapt to the plastic surface under different environmental conditions. Alpha diversity in metagenomic studies reflects the bacterial dynamics in the different layers of the open dumpsite and health of a particular ecosystem. Metrics like Abundance-based Coverage Estimator (ACE) and Chao1 were implemented to assess species richness within the groups, while the Shannon diversity index and phylogenetic diversity index were used to assess the diversity within the 2 groups.^{18,19} Beta diversity metrics compare the differences between bacterial communities in different groups. The assessment explores if the samples share similar taxa or differ significantly in their bacterial communities.¹⁸ Here we used Principal Coordinates



Analysis (PCoA) to understand how each sample within the groups differs along with PERMANOVA to establish set-significance. The Bray–Curtis method was adopted and the analysis was conducted at the genus level. All assays were conducted with the help of EZ-BioCloud.²⁰

3 Results and discussion

3.1 Soil characterization

The average pH of dumpsite soil was 8.01 while that of the control soil was around 6.67. The electrical conductivity of the dumpsite soil was $824 \mu\text{S m}^{-1}$ and that of the control soil was $542 \mu\text{S m}^{-1}$.

3.2 Spatial profiling of plastic content in the soil samples

The soil samples collected from the dumpsite were abundant in microplastic particles (less than 5 mm) when compared to the control site. Soil was sieved with a 365μ sieve and plastic particles that were visible in the soil were separated. All plastic particles were characterized with the help of FTIR. A total of 50 microplastic particles were identified. Around 60% of the particles were polyethylene (PE), around 20% samples were polypropylene (PP), and 10% samples were polyethylene terephthalate (PET) whereas the soil sample from the control site had a few PE film particles only (Fig. 2a). Around 5 + 2 particles per g of soil were visualized under the fluorescence microscope (Fig. 2b); however, due to their small size range the polymer type of the particles could not be identified using an FTIR.

3.3 Vertical profiling of plastic waste in the dumpsite

The plastic waste from each layer (20 cm height) was segregated and characterised using an FTIR. The types of plastics identified were polyethylene (PE), polypropylene (PP), polyethylene terephthalate (PET), and polystyrene (PS) (Table 1). One-way ANOVA yielded a *p*-value of 0.0001 which is <0.05 , (d.f. = 2); hence there is a significant difference between the types of plastic in the different layers.

The correlation matrix reveals pronounced, statistically robust relationships among plastic types and their vertical distribution across the three layers of the dumpsite (Fig. 3b). Most notably, there is a strong negative correlation between PE and layer (-0.94), indicating that the polyethylene (PE) concentration is highest at the surface and sharply declines with depth. In contrast, PP shows a strong positive correlation with layer ($+0.89$), and PET is moderately positively correlated ($+0.69$), signifying that polypropylene and PET accumulate preferentially in deeper layers.

The PCA biplot visually confirms and amplifies these findings (Fig. 3a) The primary axis (PC1), explaining 58.75% of total variance, is structured by positive loadings for the layers, PP, and PET, directly linking these variables to depth. PE is oriented strongly in the negative PC1 direction, confirming its association with upper layers and suggesting exclusive surface enrichment. PC2 (19.05% variance), driven primarily by PS, captures secondary variability but is less ecologically informative in this context.

Spatial profiling of the soil microplastic content showed a higher abundance of PE in the dumpsite soil. It may be

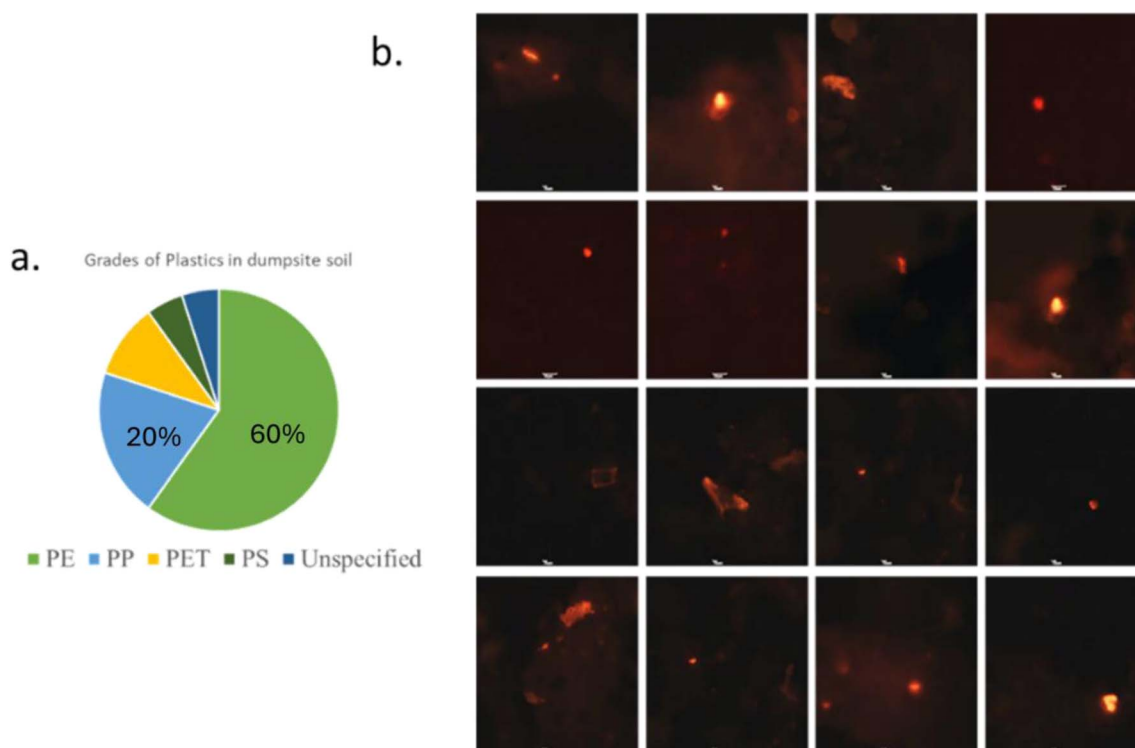


Fig. 2 (a) Distribution of plastic grades in dumpsite soil. (b) Microplastic particles in the soil from the dumpsite viewed under a fluorescence microscope at $10\times$ magnification.



Table 1 Average count of different types of plastics obtained from each layer (total $n = 120$ each layer grand-total $n = 360$, showing average values, $n = 12$)

Layer	PE	PP	PET	PS	PVC
1	8.8 + 0.6	1.7 + 0.5	0.9 + 0.5	0.6 + 0.8	0
2	3.8 + 1.4	5 + 0.8	2.6 + 1.4	0.7 + 0.5	0.2 + 0.4
3	1.5 + 0.7	5.6 + 0.7	3.7 + 0.9	0.9 + 0.9	0.3 + 0.5
Density, g cm^{-3}	0.91 to 0.965	0.895 to 0.92	1.38	0.96 to 1.05	1.1 to 1.45

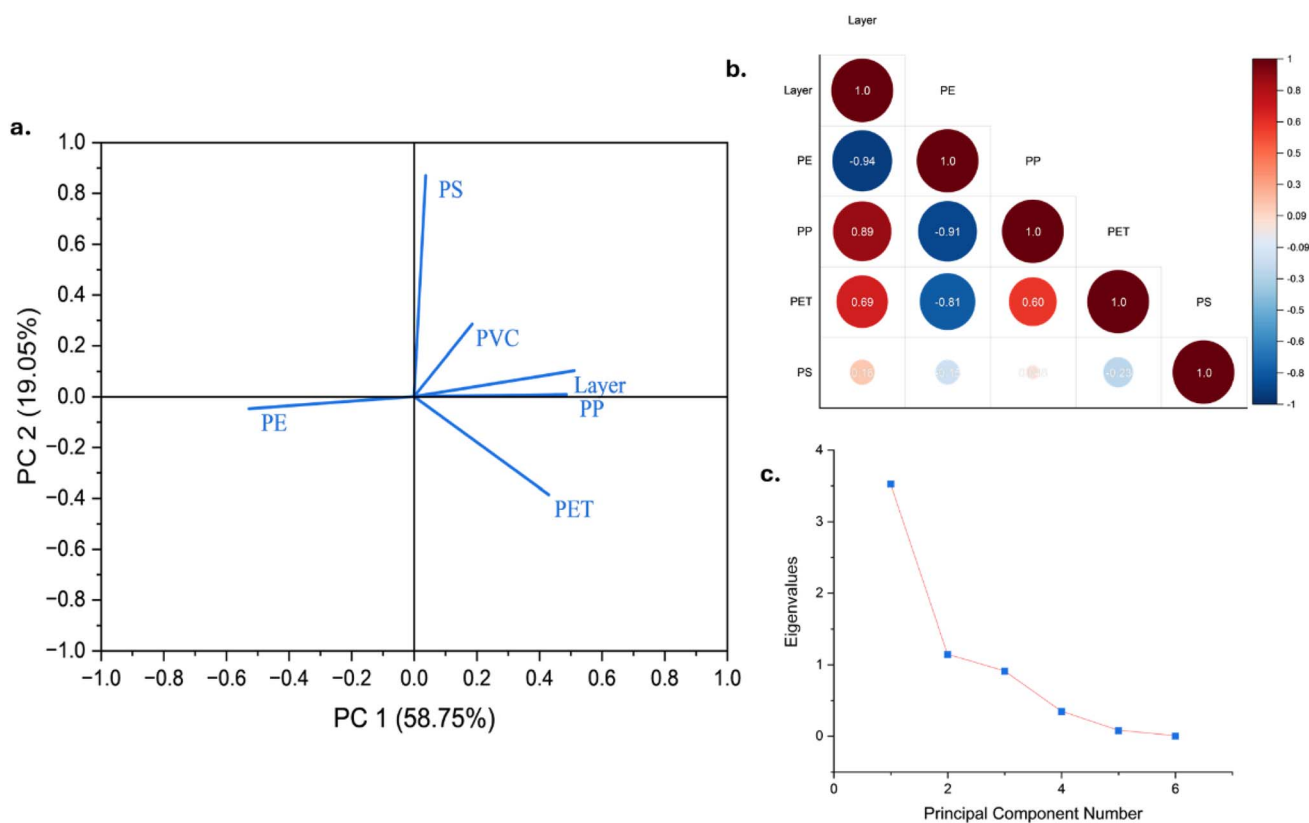


Fig. 3 (a) PCA biplot and (b) correlation matrix for the distribution of plastic types across 3 layers. (c) Scree plot for the PCA.

inferred that the lightweight PE from the topmost layers of the waste mounds might get displaced and get lodged onto the soil surface, thus integrating itself into the soil.

These results highlight how the composition of plastic waste evolves over time within a dumpsite, as heaps of new waste are dumped on top of the older heaps. These insights can be utilised for waste management practices and material persistence under different environmental conditions.

3.4 Taxonomic composition: bacterial communities thriving on plastic waste

Microbial community composition reveals ecological succession driven by environmental maturation within the waste dump (Table 2). Layer 1 (surface, freshly exposed plastics) is dominated by *Proteobacteria* (99%), characteristic of pioneer communities with low alpha diversity and high turnover rates typical of newly colonized substrates. In contrast, layers 2 and 3

exhibit significantly reduced *Proteobacteria* abundance (63% and 75.6% respectively) and substantially increased microbial diversity, indicating community stabilization during ecological succession. This depth-dependent shift reflects reduced oxygen availability and substrate depletion at depth, creating selective pressures for metabolically specialized communities.

Table 2 Relative abundance of bacterial phyla on plastic waste (%)

Phylum	Layer 1	Layer 2	Layer 3
<i>Proteobacteria</i>	~99	63	75.6
<i>Firmicutes</i>	<0.3	17	12
<i>Actinobacteria</i>	<0.07	10	6
<i>Bacteroidetes</i>	<0.06	5	3
<i>Planctomycetes</i>	—	2.2	1.3
<i>Chloroflexi</i>	—	0.4	0.3
<i>Acidobacteria</i>	—	0.2	0.1



The emergence of *Firmicutes* (17–12% in layers 2–3) and *Actinobacteria* (10–6% in layers 2–3) in deeper layers signals a functional transition toward anaerobic metabolic pathways. Notably, classes such as *Clostridia* and *Bacilli*—obligate and facultative anaerobes capable of fermentative metabolism and polymer hydrolysis, become dominant only in anoxic deeper layers. The exclusive presence of *Planctomycetes*, *Chloroflexi*, and *Acidobacteria* in layers 2–3 indicates niche differentiation and specialization absent in pioneer surface communities. Control samples, dominated by *Gammaproteobacteria* (~99%), lack these depth-specialized taxa, confirming that prolonged plastic-soil contact selects for functionally diverse, depth-adapted microbial consortia. This vertical stratification demonstrates ecological succession, *i.e.*, initial colonization by generalist proteobacteria to niche differentiation with depth-dependent functional specialization.

3.4.1 Bacterial communities thriving in the soil in and around the open dumpsite. Phylum level analysis of the bacterial community showed 19 different phyla. Distribution of these phyla in the control site and dumpsite mostly contained *Proteobacteria*, *Acidobacteria*, *Actinobacteria*, *Planctomycetes*, *Firmicutes*, *Bacteroidetes*, and *Gemmatimonadetes*. The dumpsite soil was abundant with *Proteobacteria* (46%) and *Actinobacteria*

(28%) whereas the control site was abundant with *Proteobacteria* (33%), *Acidobacteria* (28%), *Actinobacteria* (15%) and *Planctomycetes* (11%) (Fig. S4). The presence of phylum *Nitrospirae*, *Candidatus*, *Deinococcus-Thermus*, and *Spirochaetes* was only detected in the dumpsite soil while they were absent in the control soil.

By class level analysis 42 different classes were recognised; the control site was abundant with *Acidobacteria* (29%), *Alphaproteobacteria* (29%), *Actinobacteria* (13%), and *Planctomycetia* (10%) (Fig. S5). The control site had a greater relative abundance of *Acidobacteria* (29%) when compared to the dumpsite soil (0.5%) (Fig. S6). Both soils had almost equal relative abundance of *Alphaproteobacteria*, while the dumpsite soil was rich in *Actinobacteria* (27%) when compared to the control site (13%). Besides, the dumpsite had a moderately equal distribution of *Planctomycetia* (4%), *Betaproteobacteria* (5%) and *Gammaproteobacteria* (5%).

The top enriched genus of both the control site and dumpsite is presented in Fig. S7. These results indicate that the bacterial communities can help facilitate further degradation of plastic particles to tinier microplastic particles. Highest relative abundance was observed for the phylum *Proteobacteria*, class *Alphaproteobacteria*, and order *Rhizobiales* (Fig. S6).



Fig. 4 Comparative analysis of phylum from plastic waste from layers 2 and 3 and soil at the dump site.



3.4.2. Effect of microbial communities in the plastisphere on the soil. Comparative taxonomic profiling of the plastic waste and the dumpsite soil provides proof that the presence of bacteria on the plastisphere leads to the growth or propagation of bacteria in the soil too (Fig. 4). The phylum *Proteobacteria* dominates the microbial community on the plastic waste in layers 2 and 3, while *Firmicutes*, *Actinobacteria*, and *Bacteroidetes* are significantly abundant. These bacterial phyla are known to play some part in the degradation of plastics.²¹ *Proteobacteria*, ranging from 52% in layer 2 to 67% in layer 3, showed a significant increase in dumpsite soil, reaching 32% compared to just 20% in control soil. This substantial enrichment indicates directional bacterial dispersal from plastic waste into the surrounding soil matrices. *Firmicutes* abundance was higher in layers 2 and 3, with dumpsite soil exhibiting 8% *Firmicutes* relative to only 4% in control soil, reflecting successful anaerobic taxa colonization into soil environments. Most strikingly, *Acidobacteria*, which dominated control soil at 30%, declined dramatically to just 10% in dumpsite soil, suggesting that plastic waste suppressed the abundance of the oligotrophic bacteria. The compositional convergence between plastic-waste layers and dumpsite soil, notably absent in control soil comparisons, demonstrates a systematic community assembly driven by plastisphere-derived microbial recruitment into surrounding soil matrices, fundamentally transforming soil ecology adjacent to openly dumped waste.

3.5. Comparative metagenomic insights into bacterial communities on plastic waste

3.5.1. Assessment of variation in bacterial communities: alpha and beta diversity. Metagenomic analysis of plastics along with control plastics (new household consumer product packaging) showed that plastics collected from layers 1, 2, and 3 had high bacterial diversity. The alpha diversity elaborates the species richness and community composition within a sample. The analysis was based on four metrics: Abundance-based Coverage Estimator (ACE), Chao1, the Shannon index, and phylogenetic diversity (Fig. S7).

The ACE graph showed significantly higher species richness for samples of plastics collected from the waste dump (median observation = 3371), while the richness was low (median observation = 174) in the samples considered as control ($p = 0.019$). Similarly, the Chao1 graph also showed elevated species richness within the plastics collected from the dump site (median observation = 3462, $p = 0.019$). Higher values for the bacterial species collected from the plastic waste suggest the presence of a richer microbial community when compared to the control plastics.

The Shannon index and the phylogenetic index were used to explore the diversity within the samples. The Shannon index for plastics from waste had a higher range of value (median = 3.33) while the values for the control samples had a lower range (median = 0.34) with a p -value of 0.011. Similarly, the phylogenetic diversity index also yielded similar results, indicating



Fig. 5 PCoA (Bray–Curtis difference) for the bacterial communities thriving on the plastic waste vs. control plastics.



higher species richness in the plastic waste (median observation = 2448) with a p -value of 0.033. A p -value <0.05 indicates statistically significant data in all cases.

Beta diversity estimation creates a dissimilarity matrix that contains, for each sample (arranged in the rows), dissimilarity to any other sample (arranged in the columns). PCoA is a non-linear dimension reduction method, which is identical to Principal Component Analysis (PCA) (except for scaling). Principal coordinates 1 and 2 are, hence, the percentage of total variations observed. The first principal coordinate suggests 86.22% variance, while the 2nd principal coordinate accounts for 11.39%, capturing secondary differences in the microbial composition (Fig. 5).

The bacterial communities from the plastic waste collected from layers 1, 2, and 3 cluster separately from the control group (control 1 and 2) indicating distinct bacterial genera between the groups. Within the plastic bacteria group, the bacterial genus from layer 1 has a stark difference from that of layers 2 and 3. Layers 2 and 3 are closely related while layer 1 was not. This dissimilarity indicates a distinct change in the bacterial activity with the change in oxygen and light. The tight clustering of control samples indicates limited variability in the microbiota due to less exposure to the outside environment.

The PERMANOVA analysis statistically confirms differences in the composition of the microbiome with 999 permutations. The results with a pseudo- F value of 6.083 and p -value of 0.011 indicated that the differences between the two sets (plastic waste and control) were statistically significant. The pseudo- F value suggested that the inter-group variability is much greater than the intra-group variability.

Therefore, the results demonstrate that dumpsite plastics harbour distinct bacterial communities compared to control plastics. It can thus be inferred that these differences are likely driven by the environmental exposure at dumpsites, creating unique ecological niches for bacterial colonization.

3.5.2. Microbial communities thriving on the plastic waste – heatmaps. Heatmaps help illustrate the relative abundance of microbial genera across the three layers of the waste dump (layers 1–3) and the control samples (control 1 and 2). The data showcase significant differences in the microbial community structure between the plastics lying in the open dump site and those that have been freshly discarded (Fig. 6). The microbial community structure exhibits clear depth-dependent stratification with functional implications for plastic degradation. Layer 1 is dominated by aerobic bacteria (*Francisella* and *Acinetobacter*), reflecting oxygen availability. Layer 2 shows the



Fig. 6 Heatmap correlating the presence of the top 50 organisms at the genus level in all plastic samples.



critical presence of *Pseudomonas* and *Clostridium*, facultative and strict anaerobes, respectively, representing metabolically versatile taxa known for degrading synthetic polymers. Layer 3 is characterized by obligate anaerobes, including *Lachnoclostridium* and *Desulfallas*, sulphate-reducing bacteria capable of cleaving ester bonds in polyesters through anaerobic metabolism. This depth-stratified distribution aligns with oxygen gradients and reflects distinct plastic degradation niches: aerobic oxidation in surface layers, fermentative degradation in transitional zones (layer 2), and anaerobic ester hydrolysis in deep anaerobic layers (layer 3). The control samples exhibit limited diversity, dominated by *Escherichia* sp., indicative of contamination during waste disposal in households. These findings underscore how environmental gradients influence microbial colonization on plastics. Importantly, the presence of taxa like *Pseudomonas* and *Clostridium* and sulphate-reducing bacteria like *Desulfallas* suggests potential pathways for plastic degradation.⁸ *Pseudomonas* and *Clostridium* are known for their roles in breaking down complex polymers, making them key players in the biodegradation process.

3.5.3. An insight into microbial activity in the open dumpsite – KEGG pathways. PICRUST-based functional predictions yielded gene count data for metabolic pathways across the

three layers of the dumpsite. These data provide a comparative view of predicted microbial metabolic capacities, normalized to the KEGG database reference size. While these results highlight potential metabolic functions and ecological stratification, they should be interpreted as genome-derived predictions rather than direct gene expression. With the change in oxygen availability, light, and moisture, microbial activities are bound to change. The results show a distinctive pattern of metabolic activity in each layer (Fig. 8).

Microbial metabolic capacity demonstrates clear depth-dependent specialization (Fig. 7). Surface layers (layer 1) are characterized by oxidative phosphorylation and carbon fixation, supporting aerobic heterotrophs. Deeper layers (layer 3) show dominant methane metabolism, reflecting anaerobic methanogenesis. Carbohydrate metabolism pathways remain consistently abundant across all depths, indicating continuous organic matter degradation. Amino acid metabolism progressively increases with depth, suggesting enhanced protein utilization under anaerobic conditions. Notably, xenobiotic degradation pathways are present, indicating microbial capacity for plastic-derived compound metabolism. These findings demonstrate oxygen-driven metabolic niches, with implications for bioremediation strategies (Fig. 7).

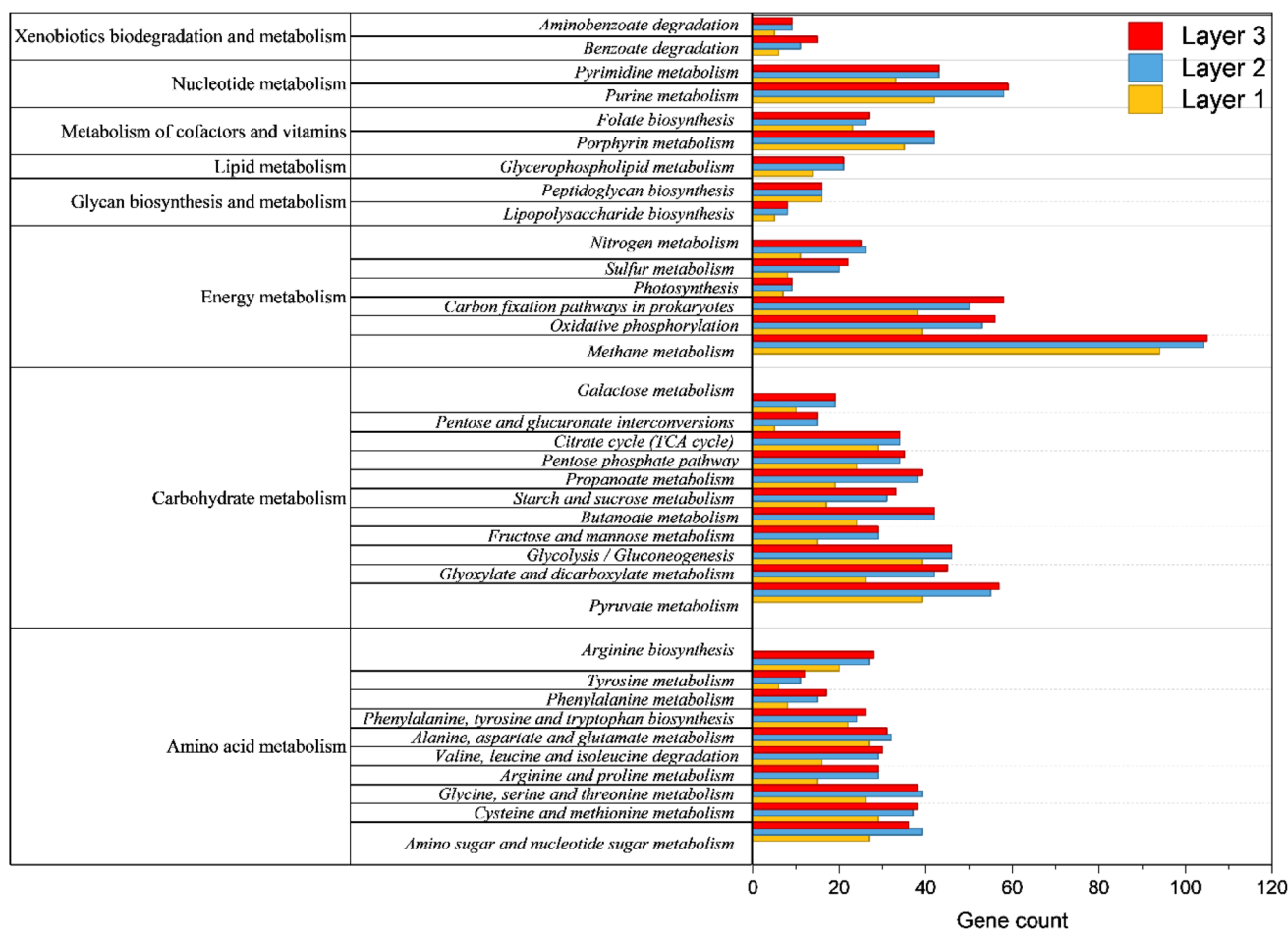


Fig. 7 KEGG pathway classification chart.



Methane metabolism activity is significantly higher in layer 3 and this activity is directly related to anaerobic conditions that enable the growth of methanogenic bacteria. In the absence of oxygen, the available CO₂ and acetate can be utilized as terminal electron acceptors. Many methanotrophs belong to the phylum *Proteobacteria*, specifically with the classes *Alphaproteobacteria* and *Gammaproteobacteria*.²² Most methanotrophs in an area with low oxygen and high methane availability belong to *Gammaproteobacteria*, while in areas with low methane and high oxygen availability, methanotrophs belonging to the *Alphaproteobacter* class are common.²³ In the taxonomic context, an increase in *Methylobacterium* sp. and *Methylocystis* sp. was significantly evident in layers 2 and 3. Such bacteria are capable of growing on single-carbon compounds that can be derived from different types of polymers.^{24,25}

Oxidative phosphorylation was higher in layers 2 and 3 when compared to layer 1; this suggests crucial adaptation of microbial communities to depth-stratified conditions. The increased oxidative phosphorylation activity in the lower layers (layers 2 and 3) implicates the presence of microbial communities capable of carrying out anaerobic respiration using alternative electron acceptors like nitrate, sulphate, or carbon dioxide.²⁶ The presence of sulphate-reducing bacteria like *Desulfovibrio* sp. and *Desulfovibrio vulgaris* in layers 2 and 3 only indicates the utilisation of sulphate as a terminal electron acceptor through modified oxidative phosphorylation pathways under semi-anaerobic and anaerobic conditions.^{27,28} Layer 1 has been exposed to atmospheric conditions, supporting predominantly aerobic microorganisms; however, the higher oxidative

phosphorylation in the bottom layers reflects a vertical stratification of microbial activity. The stratification is evident where anaerobic organisms like *Clostridium* sp. display significantly higher abundance in layers 2 and 3 when compared to layer 1.^{29–31} Similarly, *Faecalibacterium prausnitzii* shows higher abundance in layer 3, followed by layer 2, while it is absent in layer 1, suggesting adaptation to an anaerobic environment.³² Carbon fixation pathways convert inorganic carbon into organic compounds. The highest expression of this pathway was observed in layer 3, indicating the probable dominance of anaerobic autotrophic bacteria. Layer 2 supports moderate activity with limited oxygen availability thus supporting facultative autotrophs. In contrast, layer 1 with higher oxygen abundance facilitates heterotrophs over autotrophs; consequently, carbon fixation pathways are less active compared to the deeper layers.

Pathways like amino sugar and nucleotide sugar metabolism, glycine–serine–threonine metabolism, alanine–aspartate–glutamate metabolism, and nitrogen metabolism had higher expression in layer 2 than in layer 3. This represents a complex biochemical adaptation to a transitional oxygen environment. Nucleotide sugars act as critical donors for sugar residues in glycosylation reactions and act as substrates for glycosyltransferases.^{33,34} The activated sugars then link glucose metabolism with cellular functions through protein glycosylation.³⁵ Layer 2 represents a transitional zone between aerobic and anaerobic conditions, supporting optimal conditions for facultative anaerobes. Adaptation to such a metabolic strategy is advantageous for rapidly proliferating microbes that require



Fig. 8 Significant pathways identified in the three different layers (layers 1, 2, and 3) with the help of KEGG pathways.



higher amounts of anabolic precursors. The middle layer likely has an abundance of partially degraded carbohydrates from the upper layer, providing sufficient substrates for nucleotide sugar synthesis. However, in layer 3, such carbohydrates may have been depleted. Glycine–serine–threonine metabolism highlights a crucial point in amino acid metabolism. Glycine is derived from serine by the enzyme serine hydroxymethyltransferase (SHMT), while serine is derived from an intermediate in the glycolysis pathway (3-phospho-D-glycerate).³⁶ Some enzymes in this pathway require oxygen to function optimally. The glycine cleavage system releases ammonia and requires NAD⁺ (nicotinamide adenine dinucleotide) regeneration – such processes function efficiently under semi-aerobic conditions that are available in layer 2.³⁶ Alanine (ALT) and aspartate (AST) aminotransferase play important roles in nitrogen metabolism through transamination reactions.³⁷ ALT catalyses reversible transamination between alanine and 2-oxoglutarate for pyruvate and glutamate, while AST performs similar reactions with aspartate.³⁷ Hence, semi-aerobic conditions in layer 2 provide an optimal environment for transamination reactions that shuttle nitrogen between amino acids and keto acids which is absent in layer 3. The availability of some oxygen in layer 2 facilitates the participation of ALT in gluconeogenesis. Nitrogen metabolism involves the processing of ammonia which is toxic to cells and requires regular detoxification.³⁸ Layer 2 most likely experiences significant ammonia release from protein degradation and amino acid metabolism as discussed earlier. Under semi-aerobic conditions in landfill sites, microbial activity enhances decomposition, releasing nitrogen compounds.^{39,40} Also, the limited oxygen in this layer may create decentralized pockets in the environment where both nitrification and denitrification can occur simultaneously, leading to more active nitrogen cycling.

The metabolic pathways across the three layers of waste dump reflect distinct microbial adaptations to environmental gradients like oxygen availability, nutrient accessibility, and depth. Layer 3 or the bottom layer consistently shows the highest expression of key pathways, including methane metabolism, carbon fixation, oxidative phosphorylation, and glycolysis/gluconeogenesis. This is representative of specialised microbial communities capable of utilising alternative electron acceptors under strict anaerobic conditions. Layer 2 or the middle layer shows moderate expression of most pathways, reflecting the transitional state between aerobic and anaerobic zones. Facultative anaerobes dominate this layer, contributing to enhanced nucleotide metabolism (purine and pyrimidine), amino acid metabolism (alanine–aspartate–glutamate and serine–glycine–threonine metabolic pathways) and nitrogen metabolism. Layer 1, the topmost layer displays lowest pathway expression overall due to its predominantly aerobic environment. Oxygen availability in this layer favours heterotrophic activity over autotrophic and anaerobic bacterial communities. These findings affirm that stratification of metabolic pathways across layers emphasises the dynamic interplay between microbial communities and environmental conditions within the dumpsite. The dominance of plastics also has an effect on the availability of carbon as a source of energy. The results

therefore, underscore the significance of depth-specific microbial adaptation in driving key biogeochemical processes such as methane cycling, carbon fixation and energy metabolism (Fig. 8).

4 Conclusion

This study reveals the vertical and spatial patterns of plastic debris and microbial communities at a major open dumpsite in West Bengal, India. The key findings demonstrate (1) density-driven stratification of plastics, with lighter polymers accumulating at the surface and denser fractions at depth, (2) depth-specialized microbial communities showing functional shifts that may inform bioremediation and gas capture strategies, and (3) off-site PE microplastic dispersal significantly alters soil microbial ecology within a 200 m radius. Limitations of this work lie in the inability to directly measure bulk density, local wind, or hydrodynamic data; hence, results from the literature were incorporated. These novel insights underscore the inadequacy of current waste management practices and highlight the urgent need for revised segregation, site engineering, and monitoring policies. Integrating depth-resolved microbial assessment into waste impact assessments can directly inform remediation strategies, supporting India's solid waste management rules. Region-specific solutions based on these findings offer a practical roadmap toward achieving the United Nations SDGs for sustainable cities and climate action.

Author contributions

Anuja Joseph: conceptualisation, methodology, formal analysis, writing – original draft, writing – review and editing, visualisation
Sudha Goel: resources, formal analysis, writing – review & editing, supervision.

Conflicts of interest

The authors declare that they have no known competing financial interests or personal relationship that could have appeared to influence the work reported in this paper.

Data availability

The data supporting this article have been included as part of the supplementary information (SI). Supplementary information is available. See DOI: <https://doi.org/10.1039/d5va00288e>.

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