

Diversity Study of Nitrate Reducing Bacteria from Soil Samples – A Metagenomics Approach

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Abstract

The nitrogen cycle is one of the most important nutrient cycles in terrestrial ecosystems. Environmental bacteria maintain the global nitrogen cycle by metabolizing organic as well as inorganic nitrogen compounds. It is thought that most of the microbial taxa cannot be cultured outside of their natural environment, thus, microbial diversity remains poorly described before a decade. But the metagenomic techniques developed recently have therefore greatly extended our knowledge of microbial genetic diversity. The objective of this work was to analyze the taxonomic composition of three metagenome communities from the soil sample mainly rain forest, temperate broadleaf and temperate grassland via MG-RAST. Using the M5NR database, the affinities were tested for the sequences of known metabolic function against both SEED subsystems and KEGG metabolic pathways using a maximum e-value of $1e^{-5}$. Although there are a number of metabolic functions that can be tested but we probed particularly for enzymes related to the components of nitrogen cycle. The results explain the potential taxonomic diversity of nitrate reducing bacteria with the dominance of *Bradyrhizobium japonicum* from soil sample.

Keyword: Metagenome; Metagenomics; MG-RAST; Nitrate reducing bacteria

Introduction

Soils cover almost all of the terrestrial area on the Earth and have indispensable ecological functions in the global carbon cycle, nitrogen cycle and sulfur cycle. Due to their physico-chemical complexity with many micro-niches, they teem with bio-diversity, both phylogenetically and functionally [1,2]. A single gram of soil has been estimated to contain thousands to millions of different bacterial, archaeal and eukaryotic species² interwoven in extremely complex food webs. Communities of soil microbes carry out a multitude of very small-scale processes that underlie many environmentally important functions [3].

As nitrogen is one of the essential elements for all living organisms, the availability of a suitable nitrogen source often limits primary productivity of both natural environments and agriculture. Nitrogen levels in the environment are affected by an interacting web of processes, which are including the oxidation of ammonium and nitrite (nitrification), the dissimilatory reduction of nitrate (NO_3^-) to ammonium (NH_3) (nitrate ammonification), and the dissimilatory reduction of nitrate via nitrite (NO_2^-) and gaseous nitrogen oxides (NO_x) to dinitrogen gas N_2 (denitrification) [4]. Nitrogen (N) can be found in several oxidation states, from +5 in the most oxidized compound (nitrate NO_3^-) to -3 in the most reduced form (ammonium NH_4^+), but in biological compounds it is almost exclusively present in the fully reduced state [5].

The nitrogen cycle is one of the most important nutrient cycles in terrestrial ecosystems. Nitrogen cycling involves four key microbiological processes: nitrogen fixation, mineralization (decay), nitrification and denitrification [1]. Microorganisms play very important roles in the nitrogen cycles of various ecosystems. Research has revealed that a greater diversity of microorganisms is being involved in the nitrogen cycle than previous knowledge [6]. It is becoming clear that denitrifying fungi, anammox bacteria, nitrifying archaea [7] aerobic denitrifying bacteria and heterotrophic nitrifying microorganisms are key players in the nitrogen cycle [1].

Environmental bacteria maintain the global nitrogen cycle by metabolizing organic as well as inorganic nitrogen compounds.

Denitrification is critical for maintenance of the global nitrogen cycle, through which nitrate (NO_3^-) or nitrite (NO_2^-) is reduced to gaseous nitrogen forms such as N_2 and nitrous oxide (N_2O) [5]. It is thought that most of the microbial taxa cannot be cultured outside of their natural environment; thus, microbial diversity remains poorly described. The explicit functional and ecological roles of individual taxa remain unknown because most microbes withstand laboratory cultivation [8]. Therefore the most basic questions in microbial ecology is that about “who” and “what”. While soils seem to harbor [3] for the most complex microbial communities, these considerations apply to many other environments as well, like e.g. oceans and sediments [9]. The metagenomic techniques [10] developed recently have therefore greatly extended our knowledge of microbial genetic diversity [11]. With metagenomic technologies new dimensions in the characterization of complex microbial communities have been reached [12]. A large scale shotgun sequencing approaches will able the discovery of many novel genes found in the environments and which are independent of cultivation techniques [13].

In this study we used web based server, MG-RAST (Metagenomics RAST) <http://metagenomics.anl.gov/> for metagenomic analysis which is an automated analysis platform for metagenomes providing quantitative insights into microbial populations based on sequence data [14]. Metagenome samples were analyzed for the taxonomic composition with the MG-RAST server using similarity to a large non-redundant protein database; M5NR. Using the same non-redundant database, affinities were also tested for the sequences for known metabolic function against both SEED subsystems and KEGG

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metabolic pathways using a maximum e-value of $1e^{-5}$ [15]. Although there are a number of metabolic functions that can be tested but specific interest was focused on the microbial contributions at the level of nitrate reduction in nitrogen metabolism. Thus, enzymes were selected particularly to this area only.

Materials and Methods

Selection of metagenomes

For this study metagenomes were selected from MG-RAST server (<http://metagenomics.anl.gov/>) public. Mainly three parameters were taken as consideration for the selection of metagenomes respectively environment (material), environment (biome) and sequence type. Mainly the Rain forest, temperate broadleaf and temperate grassland type of soil was chosen. Whole genome shotgun sequence in sequence type was selected to have an idea of microbes present in metagenomes.

Analysis of the taxonomic abundance

Abundance is the measurement of large number of individuals in a given sample. The analysis for the taxonomic abundance on selected metagenomes was carried over by different parameters like database for annotation sequence (M5NR), e-value and percentage identity cut-off and alignment length. The e-value and percentage identity cutoff were set at $1e^{-5}$ and 60%. Alignment length of 45 was set for minimum length of matching aligned sequences in amino acid for proteins and base pair for RNA database.

Characterization of functional attributes

Characterization of functional attributes related to nitrogen metabolism, the MG-RAST server was used with the following parameters as hierarchical classification, subsystems and others. COG, NOG, SEED and etc databases were selected to find out the abundances that support relationship between functions. To compare annotated sequences, subsystem database was chosen. The values for the other

SI No.	Name	MG-RAST ID
1.	Luquillo Experimental Forest Soil, Puerto Rico	4446153.3
2.	002006_Sedgewick_ACTGAT_filtered_merged.fastq	4508941.3
3.	METASOIL J1 Rothamsted 2009 July 0-21cm Direct MPBIO1O1	4453261.3

Table 1: List of metagenomes from different source. The table shows the MG-RAST id for selected three metagenomes along with their names from the server.

METAGENOMES	4446153.3	4508941.3	4453261.3
Upload: bp Count	322,213,082 bp	1,997,174,632 bp	465,558,160 bp
Upload: Sequences Count	782,404	11,066,959	1,130,719
Upload: Mean Sequence Length	411 ± 103 bp	180 ± 11 bp	411 ± 117 bp
Upload: Mean GC percent	59 ± 6 %	61 ± 7 %	61 ± 7 %
Artificial Duplicate Reads: Sequence Count	83,075	82,191	0
Post QC: bp Count	279,379,947 bp	1,895,810,832 bp	465,558,160 bp
Post QC: Sequences Count	642,197	10,805,789	1,130,719
Post QC: Mean Sequence Length	435 ± 73 bp	175 ± 24 bp	411 ± 117 bp
Post QC: Mean GC percent	59 ± 5 %	61 ± 7 %	61 ± 7 %
METAGENOMES	4446153.3	4508941.3	4453261.3
Processed: Predicted Protein Features	677,007	10,166,026	1,159,527
Processed: Predicted rRNA Features	39,548	1,383,491	6
Alignment: Identified Protein Features	341,249	3,272,265	629,343
Alignment: Identified rRNA Features	178	6,021	911
Annotation: Identified Functional Categories	314,106	2,570,983	574,253

Table 2: Analysis Statistics Table. Metagenomic samples included in this study with associated metadata and summary statistics described below.

parameters like e-value, percentage identity cutoff and alignment length were the same as used in taxonomic abundance.

Pathway detection

For the pathway detection related to functional aspect (nitrate metabolism) in a metagenome, the KEGG map tool of the MG-RAST server used and the parameters were selected as follows: Database for annotation sequence comparison (subsystems), Maximum probability of a sequence with higher similarity to target sequence than one provided (e-value set to $1e^{-5}$), Minimum percent identity between selected metagenome with existing sBLAT sequences (percentage identity cutoff 60%), Minimum length of matching sequence in amino-acids for proteins and base pairs for RNA database (alignment length of 45).

Results and Discussion

In continuation to the earlier discussion about parameters, in the material and method section three metagenomes (Table 1) from different soil samples were selected. Soil borne microorganisms are one of the earth's greatest sources of biodiversity [16], with ranging between 3000 and 11 000 microbial genomes per gram of soil [17]. One gram of soil may contain up to 4,000 different species [18] however, current estimates indicate that less than 1% of these organisms are readily cultured with known cultivation techniques [19]. Because of the huge diversity of soil and its history as a source of commercially important molecules in agriculture, chemical, industrial and pharmaceutical industries, it remains the most common target for studies of functional metagenomics [20-22].

Metagenome analysis

In metagenome analysis we find out occurrence of domain, phyla, and etc. All three metagenomes (Table 1) were analyzed in MG-RAST server. This open-source metagenomics RAST service provides a new paradigm for the annotation and analysis of metagenomes [14]. Which has built-in support for multiple data sources and a back end that houses abstract data types, the metagenomics RAST is stable, extensible, and freely available to all researchers. This service has removed one of the primary bottlenecks in metagenome sequence analysis – the availability of high-performance computing for annotating the data. The taxonomic analysis for the bacterial community was accomplished up to the species level via M5NR database in MG-RAST. A suitable reference using similar parameter cut-off values have been cited as well.

In the first metagenome 4446153.3, 140,207 sequences failed quality control (reads more than two standard deviations away from the mean read length are discarded). Of those, dereplication identified 83,075 sequences (10.6% of total) as artificial duplicate reads (ADRs). Of the 642,197 sequences (totaling 279,379,947 bps) that passed quality control, 637,914 (99.3%) produced a total of 677,007 predicted protein coding regions. Of these 677,007 predicted protein features, 341,249 (50.4% of features) have been assigned an annotation using at least one of our protein databases (M5NR) and 335,758 (49.6% of features) have no significant similarities to the protein database (orfans). 314,106 features (92.0% of annotated features) were assigned to functional categories (Table 2). The failure rate seems high for the first metagenome. But if we look at the class distribution of the microbial community, we can observe that the Alphaproteobacteria has been dominant in the metagenome.

In metagenome 4508941.3, 261,170 sequences failed quality control. Of those, de-replication identified 82,191 sequences (0.7% of total) as artificial duplicate reads (ADRs). Of the 10,805,789 sequences (totaling 1,895,810,832 bps) that passed quality control, 10,223,265 (94.6%) produced a total of 10,166,026 predicted protein coding regions. Of these 10,166,026 predicted protein features, 3,272,265 (32.2% of features) have been assigned an annotation using at least one of our protein databases (M5NR) and 6,893,761 (67.8% of features) have no significant similarities to the protein database (orfans). 2,570,983

features (78.6% of annotated features) were assigned to functional categories (Table 2).

Table 2 also describe about the third metagenome 4453261.3, where 0 sequences failed quality control (indicates all reads are of about to mean length). Of those, dereplication identified 0 sequences (0.0% of total) as artificial duplicate reads (ADRs). Of the 1,130,719 sequences (totaling 465,558,160 bps) that passed quality control, 1,103,922 (97.6%) produced a total of 1,159,527 predicted protein coding regions. Of these 1,159,527 predicted protein features, 629,343 (54.3% of features) have been assigned an annotation using at least one of our protein databases (M5NR) and 530,184 (45.7% of features) have no significant similarities to the protein database (orfans). 574,253 features (91.2% of annotated features) were assigned to functional categories.

Taxonomic hits distribution

The taxonomic classification of protein-coding genes was assigned to the M5NR (non-redundant protein database) annotation source using the best hit classification of MG-RAST [14]. Figure 1 showed that bacterial sequences dominated in all three samples with 97.0% (4446153.3), 97.0% (4508941.3) and 95.9% (4453261.3) of all annotated sequences. Simultaneously a low number of eukaryotic 1.6% (4446153.3), 1.7% (4508941.3), 2.8% (4453261.3) as well as 1.2% (4446153.3), 1.0% (4508941.3), 1.1% (4453261.3) of Archaea and other sequences were also found (Table 3). Figure 2 illustrates the individual

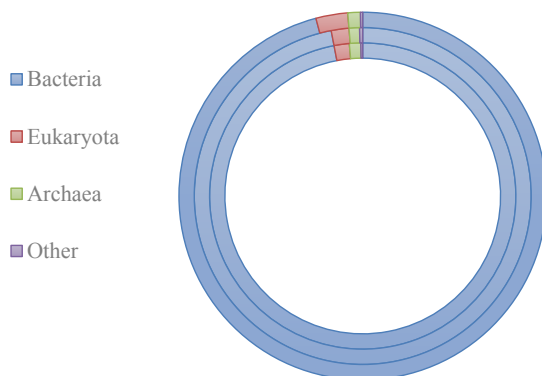


Figure 1: Combined taxonomic domain information of metagenomes. Total DNA sequences were assigned to Bacteria, Eukaryota, Archaea, viruses, and other sequences.

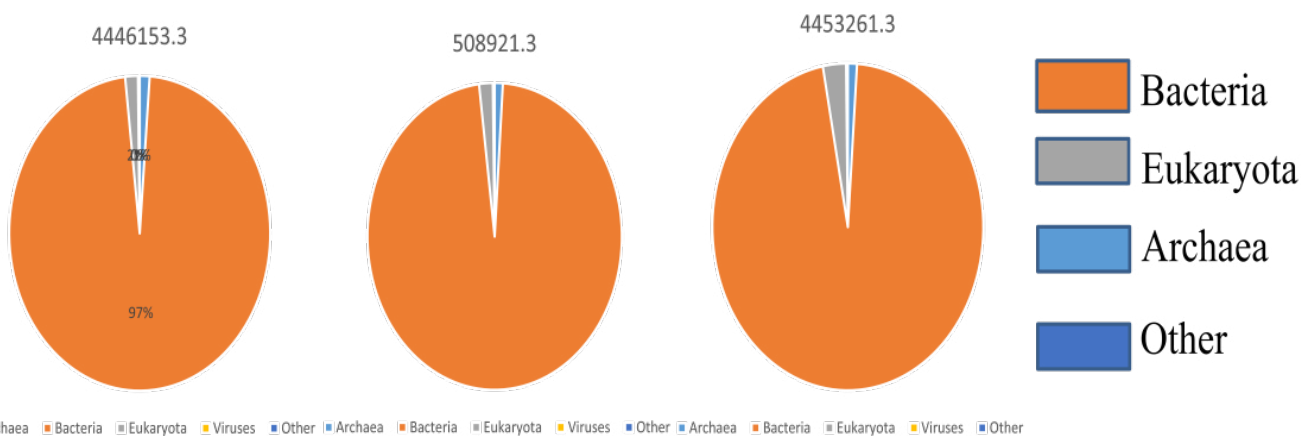


Figure 2: The pie charts below illustrate the distribution of taxonomic domains, phyla, and orders for the annotations. Each slice indicates the percentage of reads with predicted proteins and ribosomal RNA genes annotated to the indicated taxonomic level. This information is based on all the annotation source databases used by MG-RAST.

Metagenome (MG-RAST id)	4446153.3	4508941.3	4453261.3
Bacteria	97%	97%	95.9%
Eukaryota	1.6%	1.7%	2.8%
Archaea	1.2%	1%	1.1%
Other	0.2%	0.3%	0.2%

Table 3: Percentage value (%) of Bacteria, Archaea, and Eukaryota in annotated selected three metagenome sequences.

taxonomic hit distribution of all three selected metagenomes. It was clear from the figure that Bacteria were dominating the others with highest value.

When the taxonomic composition of the samples was compared to each other (with respect to nitrate reduction), they were similar at higher taxonomic organization, but differed slightly in the composition of lower taxonomic groupings Gamma-proteobacteria and Alpha-proteobacteria were dominated all samples. The microbes in all selected metagenomes showed both a taxonomic and functional composition that reflects a nitrate reduction [23]. *Bradyrhizobium japonicum* have the capacity to rapidly denitrify nitrate in soils under anaerobic conditions, *Rhodobacter sphaeroides* are able to completely denitrify nitrate to dinitrogen (N₂). In all 3 metagenomes *Bradyrhizobium japonicum* has been found to be the most abundant followed by *Pseudomonas aeruginosa*, *Pseudomonas putida*, *Rhodobacter sphaeroides* (Table 4). Presence of *Escherichia coli*, *Paracoccus denitrificans*, *Bacillus subtilis*, *Azotobacter vinelandii*, *Klebsiella pneumoniae*, *Pseudomonas stutzeri*, *Roseobacter denitrificans*, *Clostridium perfringens*, *Rhodobacter capsulatus*, *Shewanella denitrificans*, *Pseudomonas denitrificans*, *Achromobacter denitrificans* were also found.

The Gamma proteobacteria were more abundant and were identified with *Azotobacter vinelandii*, *Azotobacter chroococcum*, *Pseudomonas aeruginosa*, *Pseudomonas denitrificans*, *Shewanella denitrificans*, *Klebsiella pneumoniae*, *Escherichia coli* (Table 4) which are nitrate reducers [23]. The Alphaproteobacteria were dominated by *Rhodobacter capsulatus*, *Rhodobacter sphaeroides*, *Roseobacter denitrificans*, *Bradyrhizobium japonicum* [24] which are nitrate reducing species [25]. Out of which *Roseobacter denitrificans* were absent in metagenome 4508941.3. *Achromobacter denitrificans* sp. of the Betaproteobacteria was present in both 4446253.3 and 4508941.3; *Clostridium perfringens* sp. of class Clostridia and *Bacillus subtilis* sp. of class Bacilli were also present in all three metagenome samples.

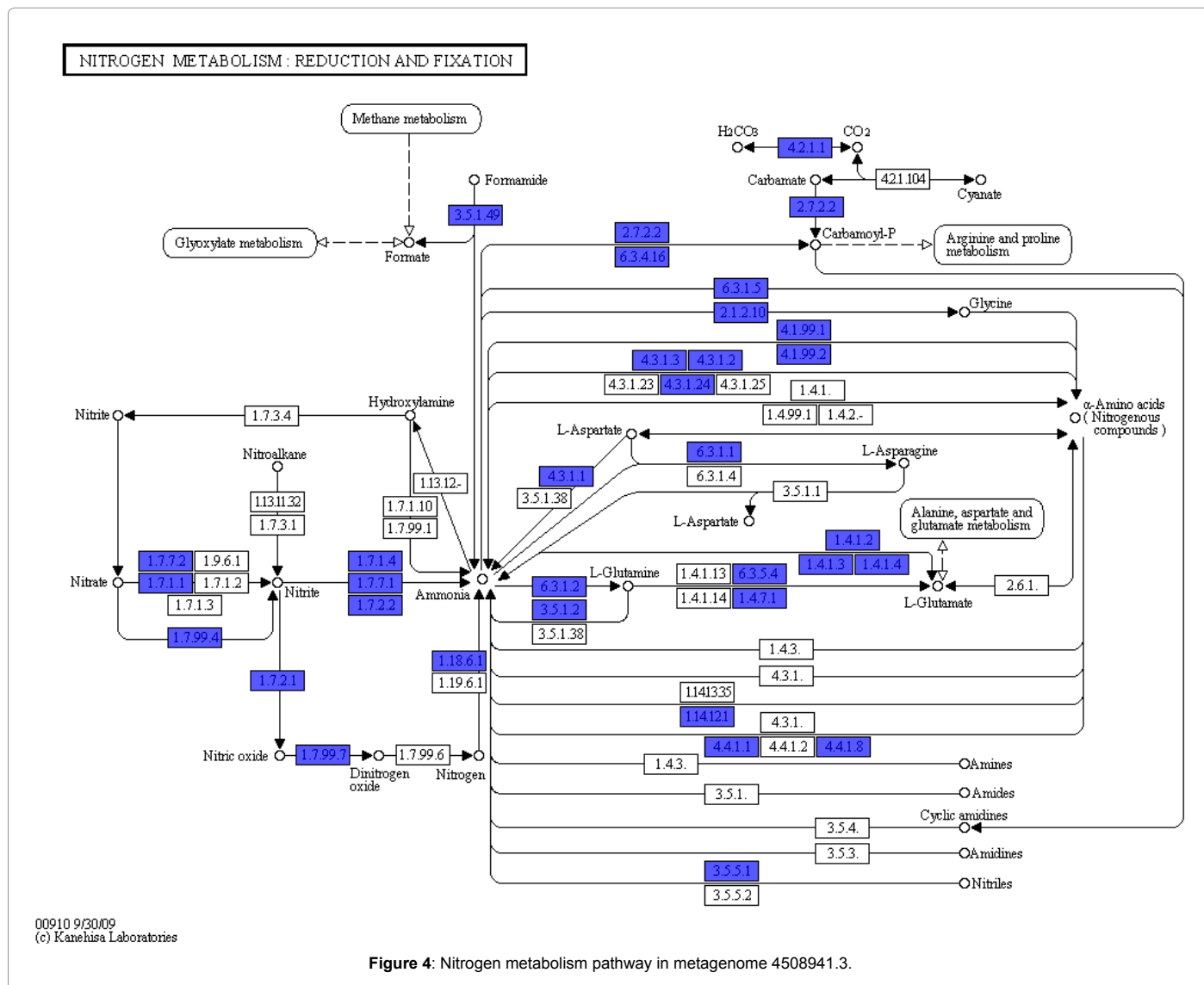
The observation of *B. japonicum* dominance has been found to be similar to the results of VanInsberghe et al. and Ormenno-Orrilli et al. [26,27]. Similarly, Delmont et al. also describes the statistical view of functional distributions of the Rothamsted soil metagenome, which aided the knowledge about soil microbial communities at a metagenomic level [28].

Nitrate is a major nitrogen source for many bacteria. In the general assimilatory pathway, nitrate is converted via nitrite to ammonia, which is then assimilated into nitrogen metabolism [29]. This metabolic route functions aerobically and anaerobically and involves assimilatory nitrate reductases which are repressed by ammonia. Nitrate can also serve as an electron acceptor for anaerobic respiration in the absence of oxygen. In this case nitrate is reduced by respiratory nitrate reductases to nitrite, the end product of nitrate respiration is denitrogen. This denitrification pathway involves [24], in addition to the respiratory nitrate reductase, further respiratory reductases for nitrite, nitric oxide, and nitrous [30]. Assimilatory nitrate reductases were found in bacteria like *Azotobacter chroococcum*, *Clostridium perfringens*, and *Ectothiorhodospira shaposhnikovii* (Ferredoxin-Nas) *Klebsiella pneumoniae* and *Rhodobacter capsulatus* (NADH-dependent).

Mainly three different types of nitrate-reducing systems have been described in bacteria [31-33]. The first type is a cytoplasmic assimilatory nitrate reductase, which enables the utilization of nitrate as the nitrogen source for biosynthesis. This enzyme is repressed by ammonium, but is not affected by oxygen [33]. The second type is a membrane-bound respiratory nitrate reductase, which catalyses nitrate respiration and the first step of denitrification to allow ATP synthesis by using nitrate as an alternative electron acceptor under anaerobic conditions. This enzyme is repressed by oxygen, but is insensitive to ammonium [31]. Membrane-bound nitrate reductases are associated with denitrification and anaerobic nitrate respiration in *Escherichia coli* and *Paracoccus denitrificans*, (NO). The membrane-bound dissimilatory nitrate reductase been shown to be involved in anaerobic nitrate reduction in *Paracoccus denitrificans*, *Pseudomonas aeruginosa*, *Pseudomonas denitrificans*, and *Pseudomonas stutzeri*. The third nitrate-reducing system is a periplasmic nitrate reductase found in some Gram-negative bacteria. This enzyme is repressed by neither ammonium nor oxygen and probably participates in redox balance and/or aerobic nitrate respiration [25]. Nitrate reductases located in the periplasmic compartment have also been described in *Rhodobacter capsulatus*, *Rhodobacter sphaeroides*, *Alcaligenes eutrophus*, *Paracoccus*

METAGENOMES	4446153.3	4508941.3	4453261.3
CLASS	SPECIES		
Alphaproteobacteria	<i>Rhodobacter capsulatus</i>	<i>Rhodobacter capsulatus</i>	<i>Rhodobacter capsulatus</i>
	<i>Rhodobacter sphaeroides</i>	<i>Rhodobacter sphaeroides</i>	<i>Rhodobacter sphaeroides</i>
	<i>Roseobacter denitrificans</i>	-	<i>Roseobacter denitrificans</i>
	<i>Bradyrhizobium japonicum</i>	<i>Bradyrhizobium japonicum</i>	<i>Bradyrhizobium japonicum</i>
Beta proteobacteria	<i>Achromobacter denitrificans</i>	<i>Achromobacter denitrificans</i>	
Gamma proteobacteria	<i>Azotobacter vinelandii</i>	<i>Azotobacter vinelandii</i>	<i>Azotobacter vinelandii</i>
	<i>Azotobacter chroococcum</i>	<i>Azotobacter chroococcum</i>	<i>Azotobacter chroococcum</i>
	<i>Pseudomonas aeruginosa</i>	<i>Pseudomonas aeruginosa</i>	<i>Pseudomonas aeruginosa</i>
	<i>Pseudomonas denitrificans</i>	<i>Pseudomonas denitrificans</i>	<i>Pseudomonas denitrificans</i>
	<i>Shewanella denitrificans</i>	<i>Ectothiorhodospira shaposhnikovii</i>	<i>Shewanella denitrificans</i>
	<i>Klebsiella pneumoniae</i>	<i>Klebsiella pneumoniae</i>	<i>Klebsiella pneumoniae</i>
	<i>Escherichia coli</i>	<i>Escherichia coli</i>	<i>Escherichia coli</i>
Clostridia	<i>Clostridium perfringens</i>	<i>Clostridium perfringens</i>	<i>Clostridium perfringens</i>
Bacilli	<i>Bacillus subtilis</i>	<i>Bacillus subtilis</i>	<i>Bacillus subtilis</i>

Table 4: The taxonomic diversity of microbes (nitrate reducers) from the selected metagenomes.



with their probable activity were studied. The results explain the potential taxonomic diversity of nitrate reducing bacteria with the dominance of *Bradyrhizobium japonicum* from soil sample. The nitrate reducing metabolic pathway were studied and it is found that in given metagenomes all three pathways of nitrate reduction i.e. assimilatory nitrate reduction, respiratory (dissimilatory) nitrate reduction as well as dinitrification are present.

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