Temporal Dynamics of Abundance and Composition of Nitrogen-Fixing Communities across Agricultural Soils

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Abstract

**Background:** Despite the fact that the fixation of nitrogen is one of the most significant nutrient processes in the terrestrial ecosystem, a thorough study of the spatial and temporal patterns in the abundance and distribution of N-fixing communities has been missing so far.

**Methodology/Principal Findings:** In order to understand the dynamics of diazotrophic communities and their resilience to external changes, we quantified the abundance and characterized the bacterial community structures based on the *nif* gene, using real-time PCR, PCR-DGGE and 454-pyrosequencing, across four representative Dutch soils during one growing season. In general, higher *nif* gene copy numbers were observed in soils with higher pH than in those with lower pH, but lower numbers were related to increased nitrate and ammonium levels. Results from *nif* gene pyrosequencing confirmed the observed PCR-DGGE patterns, which indicated that the N fixers are highly dynamic across time, shifting around 60%. Forward selection on CCA analysis identified N availability as the main driver of these variations, as well as of the evenness of the communities, leading to very unequal communities. Moreover, deep sequencing of the *nif* gene revealed that sandy soils (B and D) had the lowest percentage of shared OTUs across time, compared with clayey soils (G and K), indicating the presence of a community under constant change. Cosmopolitan *nif* species (present throughout the season) were affiliated with *Bradyrhizobium*, *Azospirillum* and *Methylocistis*, whereas other species increased their abundances progressively over time, when appropriate conditions were met, as was notably the case for *Paenibacillus* and *Burkholderia*.

**Conclusions:** Our study provides the first in-depth pyrosequencing analysis of the N-fixing community at both spatial and temporal scales, providing insights into the cosmopolitan and specific portions of the nitrogen fixing bacterial communities in soil.


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Introduction

The nitrogen cycle is one of the most significant nutrient cycles in the terrestrial ecosystem. Among its processes, biological nitrogen fixation (BNF), the reduction of atmospheric N₂ gas to biologically available ammonium, is a key one, as it replenishes the pool of reduced nitrogen that is lost to the atmosphere via anaerobic ammonium oxidation and denitrification [1]. BNF is widely distributed across bacterial and archaean taxa [2], being performed by a highly diverse group of microorganisms. These diazotrophic organisms harbor the *nif* gene, which encodes part of the nitrogenase enzyme. This gene is considered to be a suitable marker for studying the diversity and composition of N fixers without the need of microbial cultivation [3,4]. In traditional agricultural fields, nitrogen input may rely to some extent on biological N fixation, which occurs primarily (but not exclusively) by diazotrophic bacteria in symbiosis with legumes [5]. It has been suggested that symbiotic N-fixers can fix up to 150 kg N/ha/year in some ecosystems [6]. By contrast, estimates of the contribution of free-living nitrogen fixation are much lower (0 to 60 kg N/ha/ year) [7]. However, these can also be significant and even critical under certain conditions, depending on the environment [8-10].

It is known that the local environment in which soil organisms coexist often changes with time [11], which influences how they
structure themselves and perform their functions. Furthermore, global change is expected to severely affect N turnover in soils [12], impacting processes of the nitrogen cycle to an unknown extent. In this sense, several classes of nitrogen-fixing bacteria might be affected, as these are known to be sensitive to perturbation [13,14]. In general, fluctuations in local conditions affect the soil microbial community structure and composition [15,16] at different temporal and spatial scales. For instance, it has been found that the temporal variation in nitrogen fixation rates within ecosystems may vary up to 50% across seasons [17-19]. These temporal variations are likely driven by fluctuations in precipitation [20] and temperature [21]. Agricultural practices, such as fertilization and ploughing, also play major roles as determinants of bacterial community structure in soil [22-24]. In this context, several environmental factors have been suggested to influence N fixation in soils, including soil moisture, oxygen, pH, carbon quantity and quality, nitrogen availability [3], soil texture and aggregate size [25] and clay content [8]. Changes in nutrient availability also alter the rates of nitrogen fixation through changes in community composition [3,26,27]. For instance, the expression of *nifH* gene was found to be down-regulated under low P conditions [28]. Moreover, Lindsay et al. [27] observed a negative correlation between *nifH* gene abundance and high N concentrations in a managed ecosystem.

Understanding temporal and spatial patterns in the abundance and distribution of communities has been a fundamental goal of ecology. In order to understand the functioning of microbial communities and their resilience to external changes, a key issue is to assess the community composition, quantify individual microbial population sizes and rare micro biosphere in ecological analyses is challenging in the face of the technical noise in pyrosequencing data [31,32], and such information has been so far missing for nitrogen fixing communities. Therefore, the aim of this study was to provide a thorough description of the natural variation of diazotrophic organisms in a wide range of soils. To this end, we characterized the spatio-temporal variability of *nifH*-containing microorganisms in agricultural soils, as measured by changes in PCR-DGGE fingerprints, real-time PCR and *nifH*-gene pyrosequencing, and related this variability to environmental conditions. These analyses were performed for four different soils at three times over one growing season, i.e. April, June and October 2010. Considering the sensitivity of diazotrophic organisms to N fertilization, we hypothesized that their abundance and diversity would be lower at the beginning of the season, when fertilizers are applied, compared to October. Moreover, due to the strong effect of soil type on microbial communities [4,33] we hypothesized that these differences will be more relevant in clayey than in sandy soils.

**Materials and Methods**

**Sampling sites**

Four soils (B, D, G and K) from different sites in the Netherlands were sampled three times in 2010, April (after seedling), June (before flowering) and October (senescence stage). The fields are used for potato cropping and are under agricultural rotation regime with non-leguminous crops. Information on land use, fertilization scheme and location is available (Table 1). The soils were chosen to represent two agricultural rotation regimes with non-leguminous crops. 

<table>
<thead>
<tr>
<th>Sampling Location</th>
<th>Buinen (B)</th>
<th>Droevendaal (D)</th>
<th>Kollumerwaard (K)</th>
<th>Grebbedijk (G)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Soil type</td>
<td>Sandy loam</td>
<td>Sandy loam</td>
<td>Clayey</td>
<td>Clayey</td>
</tr>
<tr>
<td>Land use</td>
<td>Agricultural</td>
<td>Agricultural</td>
<td>Agricultural</td>
<td>Agricultural</td>
</tr>
<tr>
<td>Sampling coordinates</td>
<td>52°55'386&quot;N 006'49'217&quot;S</td>
<td>51°59'551&quot;N 005'39'608&quot;S</td>
<td>53°19'507&quot;N 006'16'351&quot;S</td>
<td>51°57'349&quot;N 005'38'086&quot;S</td>
</tr>
<tr>
<td>Fertilization scheme/date</td>
<td>KAS (212Kg N/ha April and June)</td>
<td>NPK (200Kg N/ha March and April)</td>
<td>KAS (150Kg N/ha March and April)</td>
<td>KAS (156Kg N/ha March and April)</td>
</tr>
<tr>
<td>pH</td>
<td>4.40±0.02</td>
<td>4.93±0.02</td>
<td>7.40±0.05</td>
<td>7.20±0.02</td>
</tr>
<tr>
<td>N-NO₃ (mg/kg)</td>
<td>47.51±0.37</td>
<td>60.53±1.66</td>
<td>24.60±1.72</td>
<td>29.93±1.15</td>
</tr>
<tr>
<td>N-NH₄⁺ (mg/kg)</td>
<td>9.23±0.74</td>
<td>14.18±1.41</td>
<td>8.40±0.75</td>
<td>15.13±2.35</td>
</tr>
<tr>
<td>OM (%)</td>
<td>3.96±0.72</td>
<td>2.91±0.24</td>
<td>4.24±0.14</td>
<td>5.47±0.22</td>
</tr>
<tr>
<td>Water content (%)</td>
<td>11.20±0.58</td>
<td>11.63±1.62</td>
<td>19.30±1.40</td>
<td>19.60±1.86</td>
</tr>
<tr>
<td>nifH-gene abund (x 10⁵ gdw⁻¹)</td>
<td>0.72±0.28</td>
<td>1.15±0.43</td>
<td>17.0±8.13</td>
<td>3.10±0.94</td>
</tr>
<tr>
<td>Bacterial 16S rRNA gene abund (x 10⁹ gdw⁻¹)</td>
<td>1.74±0.21</td>
<td>1.68±0.13</td>
<td>1.98±0.15</td>
<td>1.98±0.13</td>
</tr>
</tbody>
</table>

Values of environmental and biological data are average of each soil across the three sampling times (mean ± sd), KAS = calcium, ammonium, nitrate; NPK = nitrate, phosphorus, calcium

| Table 1. Sampling locations, environmental and biological data. |
DNA extraction

DNA was extracted from 0.5 g soil using the FastDNA SPIN Kit for Soil (MP Biomedicals) according to manufacturer’s protocol. Extracted DNA was then precipitated and concentrated with cold ethanol to remove impurities. DNA concentrations were determined using the Quant-IT™ PicoGreen® dsDNA Reagent (Invitrogen, Germany) according to the manufacturer’s protocol’s using a spectrophotometer (Spectramax Gemini, Germany). The quality of extracted DNA was estimated by running on agarose gel based on the degree of DNA shearing (average molecular size) as well as the amounts of coextracted compounds.

PCR-DGGE analysis

PCR reactions for DGGE analysis were performed targeting bacterial 16S rRNA gene as well as the nifH gene, as a proxy of nitrogen-fixing community. Bacterial 16S rRNA gene was PCR amplified using the forward primer F968 [35] with a GC-clamp attached to 5′ and the universal primer R1401.1b [36]. PCR of the nifH gene was performed using a nested PCR according to Diallo et al. [37]. PCR reactions and cycling conditions are described in Table S2. DGGE profiles were generated with the Ingeny Phor-U system (Ingeny International, The Netherlands). The PCR products (300 ng/lane) were loaded onto 6% (w/v) polyacrylamide gels, with a denaturant gradient of 45-65% for total bacterial community and 40-65% for nitrogen-fixing community (100% denaturant corresponded to 7 M urea and 40% (v/v) deionized formamide). Electrophoresis was performed at a constant voltage of 100 V for 16 h at 60°C. The gels were stained for 60 min in 0.5x TAE buffer with SYBR Gold (final concentration 0.5 µg/liter; Invitrogen, The Netherlands). Images of the gels were obtained with Imagemaster VDS (Amersham Biosciences, United Kingdom).

Quantitative PCR

The gene abundances in the soil samples were quantified using quantitative PCR (qPCR). For the nifH gene, the primers FPGH19 [38] and PolR [39] were used according to Pereira e Silva et al. [4]. To be able to compare the R1401.1b-R1401.1a PCR reactions of the diazotrophic community, we also quantified the bacterial 16S rRNA gene as well as the nifH gene. The expected sizes of the amplified fragments were checked in 1.5% agarose gel stained with ethidium bromide. Possible inhibitory effect of co-extracted humic compounds was checked by spiking standard concentrations with serial dilutions of soil samples. No severe inhibition was observed at the working dilutions. The qPCR efficiency (E) was calculated according to the equation E = 10^{-1/slope} [41] and was 1.94 (bacterial 16S rRNA gene) and 1.92 (nifH gene). The R² of all these standards was higher than 0.98.

Deep-sequencing and bioinformatics analysis of nifH gene

The diversity of nitrogen-fixing bacteria in the four agricultural soils was investigated through barcoded pyrosequencing. The total community DNA was amplified with the nifH gene-specific primers PolF/PolR [39] and RoeschF/ RoeschR [42] in a nested approach using the FastStart High Fidelity PCR system and PCR Nucleotide Mix (Roche Diagnostics GmbH, Germany). PCR conditions and primer sequences are presented in Tables S3 and S4. Triplicate PCR amplifications were performed on each soil DNA template and pooled. Primer dimers were removed by electrophoresis of PCR products on agarose gel, excision, and purification using Qiaquick PCR purification Kit (Qiagen). For pyrosequencing, adapters and sample-specific tags were added to the samples using custom primers in an additional PCR amplification of 20 cycles using the same PCR conditions. Amplicons were further purified with Agencourt AMPure Beads XP (Beckman Coulter, Germany) and pooled in an equimolar ratio as specified by Roche. Sequencing from 5′ (forward) and 3′ (reverse) ends of amplicons was performed. Emulsion PCR, emulsion breaking of DNA-enriched beads, and sequencing runs of the amplicon pools were performed on a second-generation pyrosequencer (454 GS FLX Titanium; Roche) using titanium reagents and titanium procedures as recommended by the manufacturer.

Quality filtering of the pyrosequencing reads was performed using the automatic ampiclon pipeline of the GS Run Processor (Roche) to remove failed and low quality reads from raw data. Amplicon libraries of the nitrogenase gene (nifH) were explored using the FunGene Pipeline of RDP server (http://fungene.cme.msu.edu/FunGenePipeline) using the default settings. Primer sequences were trimmed and reads of low quality and shorter than 320 bp were removed. Filtered nucleotide sequences were translated into amino acid and clipped at 108 bp. All subsequent analyses were done on amino acid sequences. By targeting a protein-encoding gene, frame-shift errors caused by insertions or deletions of bases, can be identified [43]. Sequences were then visually inspected and sequences containing in-frame stop codon(s) were removed. The amino acid sequences were aligned in MUSCLE 3.8 [44].

Operational taxonomic units (OTUs) were then classified and rarefaction curves were constructed with DOTUR [45] using a 90% amino acid sequence similarity cutoff level [46,47]. Richness estimates and diversity indices were calculated for the total number of sequences as well as for subsets normalized to the same number of sequence depth, which were performed using the Perl script daisychopper.pl (available at http://www.genomics.ceb.ac.uk/ GeneSwytch/Tools.html). Phylogenetic analysis was performed for clustered sequences at 90% similarity using CD-HIT [48], and were so called here abundant (core) OTUs. Rare (satellite) species, in contrast to
the abundant (core) ones, are defined here as the OTUs (clustered at 90% similarity cutoff) that were represented by less than 10 sequences in the whole dataset. The nifH representative sequences were blasted against a non-redundant protein sequence database using BLASTp, and used to build neighbor-joining trees in MEGAS [49]. Normalized weighted UniFrac significance [50] was calculated and PCoA analysis was done to evaluate differences between the nifH gene communities based on the phylogenetic trees obtained in MEGAS.

Data analysis

Computer-assisted analysis of DGGE profiles was performed using the GelCompar software (Applied Maths, Belgium), to follow the structural changes on these communities over time. Similarity matrices were constructed based on Pearson’s correlation coefficient, and cluster analyses were done by unweighted pair group method with average linkages (UPGMA). The differences in soil chemical parameters and nifH gene abundances in the different soils over time were estimated with independent t-Tests. To test the influence of soil physicochemical parameters (environmental factors) on the community data, forward selection was used via canonical correspondence analysis (CCA) or redundant correspondence analysis (RDA), depending on the gradient length observed on DCA, using Canoco (version 4.0 for Windows, PRI, Wageningen, The Netherlands). In cases where gradient length was longer than 4.0, a unimodal method (CCA) was used. Otherwise, with gradients shorter than 4.0, RDA was chosen. For this, all parameters except pH were log_{10} transformed, and the community matrices (obtained from DGGE bands as well as from pyrosequencing) were square-root transformed. Furthermore, to test how species richness, evenness, gene abundance and the relative abundance of OTUs varied in relation to environmental variables, parametric Pearson correlation coefficients were calculated in SPSS v20.0 (SPSS Inc., USA). To visualize changes in most abundant and rare genera across samples, we constructed heatmaps and dendrograms with clustering based on Ward’s minimum variance and utilizing Euclidian distance calculation with scaling. The dendrograms are not phylogenetic but clustered based on the relative abundance among samples. Thus, more similar samples are more closely related. These analyses were done in the R package pheatmap (http://cran.r-project.org/web/packages/pheatmap/index.html).

Data Availability

Sequences from this study were submitted to MG-RAST Metagenomic Analysis Server under the ID numbers 4524846.3, 4524847.3, 4524848.3, 4524849.3, 4524850.3, 4524851.3, 4524852.3, 4524853.3, 4524854.3, 4525982.3, 4525983.3 and 4525984.3.

Results

Seasonal variations of soil chemical properties

Soil pH, nitrate, ammonium and organic matter levels were determined in triplicate across all soil samples (Table 1). Individual values for each soil at each sampling time can be found in Table S1.

pH and OM levels were significantly higher (P < 0.05) in the clayey soils K and G than in the sandy ones B and D. In contrast, the levels of nitrate were higher in the sandy soils (B and D, 47.17 ± 0.37 and 60.53 ± 1.66, respectively) than in the clayey ones (K and G, 29.93 ± 1.15 and 24.60 ± 1.72, respectively). The levels of ammonium also varied over the whole period: relatively low values were observed for all soils in October (on average 5.47 mg/kg ± 0.66) and higher levels in April and June (on average, 14.63 mg/kg ± 2.66 and 14.01 mg/kg ± 0.61, respectively).

Size of total bacterial and N-fixing community over time as related to soil variables

We quantified the abundance of the bacterial 16S rRNA gene to be able to draw conclusions regarding the fluctuations of the diazotrophic population. Total bacterial abundance showed small fluctuations from April to June but were quite stable from June to October, with copy numbers varying from 5.30 x 10^6 to 4.34 x 10^6 gdw^{-1} over the growing season, being significantly higher (P<0.05) in October (1.69 x 10^6 gene copy numbers gdw^{-1}) compared to April (1.38 x 10^6 gene copy numbers gdw^{-1} and June (9.13 x 10^6 gene copy numbers gdw^{-1}) (Figure 1). Analysis per soil revealed that the clayey soils K and G had significantly higher nifH gene abundance at all sampling times (1.0 x 10^6 gene copy numbers gdw^{-1} on average) compared to the sandy soils B and D (9.33 x 10^6 gene copy numbers gdw^{-1} on average). Pearson’s product-moment correlations were calculated to test the influence of soil variables on bacterial 16S rRNA and nifH gene abundance. None of the soil parameters measured were able to explain the small variations in total bacterial abundance. However, nitrate (r = -0.543, Ps0.0001), ammonium (r = -0.565, Ps0.0001) and pH (0.645, Ps0.0001) were identified as the main drivers of variation in nifH gene abundance.

Structure of the nifH gene carrying communities in relation to soil variables

In order to characterize the dynamics of the N-fixing communities through time, we performed PCR-DGGE analyses based on the bacterial 16S rRNA and nifH genes. Concerning the analysis of the resulting bacterial profiles, we observed similar number of bands in all soils, (approximately 41±1, 46±1, 44±1 and 41±1 for soil B, D, G and K respectively), and tended to decrease from April to June in all soils, increasing again from June to October (Figure S1). Forward selection on canonical correspondence analysis (CCA) (Figure 2A) revealed nitrate (10.3%, P = 0.032), ammonium (2.1%, P = 0.021) and pH
(1.9%, P = 0.021) as main determinants of the structural changes in the bacterial community. DGGE analysis based on the \textit{nifH} gene revealed similar results. Moreover, the number of bands tended to decrease from April to October for the sandy soils, but remained constant in the clay soils. Such number of bands were low and varied to a limited extent over time, with averages of 15±2, 12±1, 11±2 and 11±1 for soils B, D, K and G, respectively (Figure S1). Analysis of the Shannon diversity index values for the whole dataset (data not shown) indicated that the date of sampling had no significant effect on \textit{nifH} gene diversity, except for soil G, which showed a significantly lower Shannon index in June as compared to April and October. Overall, the \textit{nifH} gene-based profiles clustered into five groups at 38% similarity (Figure S2). The clustering reflected an increasing effect of sampling time on the N-fixing community structure. Forward selection on canonical correspondence analysis (CCA) revealed nitrate (28.6%, P = 0.04), clay content

Figure 1. Changes in abundance of bacterial 16S rRNA and \textit{nifH} genes. Soils B, D, K and G were collected in April, June and October of 2010. The copy numbers per gram of dry soil was estimated by real-time PCR. Soils: B, Buinen; D, Droevedaal; K, Kollumerwaard and G, Grebedijk.

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and ammonium (16.1%, P = 0.003) as the main factors affecting this clustering (Figure 2B).

**Pyrosequencing-based assessment of nifH gene diversity**

To further analyze the changes in the composition of the communities of N-fixing microorganisms in the four agricultural soils, the nifH gene was sequenced using pyrosequencing. To assess the effect of any technical biases, the forward and reverse reads were analyzed separately. In total, 91,015 reads were obtained. The total read numbers per sample are described in Table S5. These (transformed into amino acid sequences, average length of 108 amino acids) ranged from 1,921 to 5,892, corresponding to 40-120 OTUs (defined at a 10% sequence difference). The forward reads of the nifH amplicons yielded slightly more OTUs than did the reverse ones; however, the number of sequences was lower, indicating that the resolving power of the forward nifH reads was higher for diversity analysis than that of the reverse reads. In addition, the results obtained from the reverse reads showed similar trends to those obtained from forward ones. Therefore, we only present the results obtained from the forward reads.

Rarefaction analysis of the forward read based libraries resulted in different saturation profiles, suggesting that the nifH gene diversities were different across the soils (Figure S3). In total, the retrieved nifH sequences varied substantially among the soils analyzed and also over time, being distributed among Alpha-, Beta-, Gamma-, Delta- proteobacteria and Firmicutes (Figure 5). The lowest nifH gene diversity was observed in soil B, which was characterized mostly by the dominating Bradyrhizobium, Methylocystis, Methylomonas and Termochromatium species, which together made up 61% of the community. The highest diversity was observed in soil D in October (Figure 3). Soil D was dominated by species of Azospirillum, Azohydromonas, Rhizobium and Herbaspirillum, which represented 46% of the community. In soil G species of Celerinatantimonas, Azoarcus, Burkholderia and Azospirillum comprised up to 91% of the community. Soil K was dominated by Azoarcus, Bradyrhizobium, Celerinatantimonas and Rhodoferax species, making up to 71% of this soil community.

Figure 2. Changes in the structure of total bacterial and N-fixing communities from DGGE fingerprints across different soils and sampling times. Changes in the structure of bacterial (a) and N-fixing (b) communities and the influence of environmental parameters, as revealed by CCA. Only statistically significant environmental variables are shown. The number in each axis shows the percentage of total variation explained. The length of the corresponding arrows indicated the relative importance of the geochemical factor in explaining the variation in microbial profiles. Soil samples were analyzed in four replicates at each sampling time. B, Buinen; D, Droevendaal; K, Kollumerwaard and G, Grebbedijk; Ap, April; Ju, June; Oc, October.

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An analysis at higher taxonomic rank per soil type revealed that soils B and D were dominated by nifH-containing members of the class Alphaproteobacteria, around 69%, followed by 22% of Betaproteobacteria and 1% of Gammaproteobacteria. Soils K and G were dominated by a much lower percentage of Alphaproteobacteria (45%), followed by Beta-, with around 36%, and 17% of Gammaproteobacteria. Members of the Alphaproteobacterial classe were more than 95% similar to e.g. Bradyrhizobium japonium, Rhizobium etli and Azospirillum canadense. Members of Betaproteobacteria were affiliated with Burkholderia vietnamiensis, Herbaspirillum seropedicae and Azohydromonas australica, and members of the Gammaproteobacterial class were closely related to Celerinatantimonas diazotrophica, Thermochromatium tepidum and Azomonas agilii.

Dynamics of abundant and rare nifH gene sequences
To understand which nifH gene types were affected by seasonality, we identified the OTUs which exhibited changes in abundance throughout the year. First, a very clear seasonal effect was observed in all soils. Interestingly, in particular cases a completely different community was found to dominate at each sampling time (Figures S4 and S5). When time was considered alone, only 15 to 37% of the OTUs were shared

Table 2. Richness estimates and diversity indices for forward amplicon libraries at 90% similarity cutoff after random resampling of sequences to the same depth (1921 sequences).

<table>
<thead>
<tr>
<th>Library</th>
<th>OTUs Estimate OTU richness</th>
<th>Shannon</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>B_Ap</td>
<td>44.25 (42.16; 71.48)</td>
<td>53.53 (44.54; 80.31)</td>
</tr>
<tr>
<td>D_Ap</td>
<td>95 (71.10; 175.41)</td>
<td>130.69 (107.37; 108.69)</td>
</tr>
<tr>
<td>K_Ap</td>
<td>73.11 (62.76; 105.96)</td>
<td>123.4 (99.41; 177.03)</td>
</tr>
<tr>
<td>G_Ap</td>
<td>75 (63.49; 110.55)</td>
<td>99.46 (82.48; 141.56)</td>
</tr>
<tr>
<td>B_Ju</td>
<td>71.2 (52.63; 133.53)</td>
<td>59.55 (49.57; 87.45)</td>
</tr>
<tr>
<td>D_Ju</td>
<td>71.43 (58.06; 114.31)</td>
<td>89.17 (74.59; 125.92)</td>
</tr>
<tr>
<td>K_Ju</td>
<td>105.5 (75.89; 212.48)</td>
<td>256.19 (192.21; 375.32)</td>
</tr>
<tr>
<td>G_Ju</td>
<td>85.14 (67.04; 139.51)</td>
<td>174.71 (142.76; 223.78)</td>
</tr>
<tr>
<td>B_Oc</td>
<td>80.2 (66.82; 154.09)</td>
<td>134.79 (95.36; 222.26)</td>
</tr>
<tr>
<td>D_Oc</td>
<td>122.5 (96.20; 188.97)</td>
<td>193.92 (160.06; 256.39)</td>
</tr>
<tr>
<td>K_Oc</td>
<td>138.2 (92.71; 267.43)</td>
<td>173 (141.24; 237.01)</td>
</tr>
<tr>
<td>G_Oc</td>
<td>49 (50.56; 121.68)</td>
<td>106.57 (86.35; 147.91)</td>
</tr>
</tbody>
</table>

a. Calculated with DOTUR at the 10% distance level.

b. Shannon diversity index calculated using DOTUR (10% distance level).

Values in brackets are 95% confidence intervals as calculated using DOTUR.

doi: 10.1371/journal.pone.0074500.g003

Figure 3. Rarefaction analysis of the diversities of nifH gene. Analyses of four soils across three sampling times after resampling of the sequences to the same depth (1921 sequences). The OTUs were classified at 90% similarity cutoff based on amino acid sequences. B, Buinen; D, Droevendaal; K, Kollumerwaard and G, Grebbedijk; Ap, April; Ju, June; Oc, October.
doi: 10.1371/journal.pone.0074500.g003
between sampling times. This percentage was much higher for clayey soils G (37.25%) and K (36.60%) than for sandy soils B (15%) and D (27.5%). For each soil, the stable OTUs consisted of \( nifH \) genes typical for \textit{Bradyrhizobium} and \textit{Paenibacillus} (soil B), \textit{Azospirillum} and \textit{Rhizobium} (soil D), \textit{Azoarcus} and \textit{Celerinatantimonas} (soil G) and \textit{Celerinatantimonas} and \textit{Burkholderia} (soil K) species. Moreover, only low percentages of OTUs were shared between soils at each sampling time, i.e. around 14-27%. In April the \( nifH \) gene based diversity tended to be low for all soils being represented mostly by Alphaproteobacterial members (around 55%), and specific by \textit{Bradyrhizobium} and \textit{Azospirillum} species. Nevertheless, diversity levels increased markedly towards the end of the season. This was very clear for soils B, D and K (Figure S4). In the June samples, similar high percentages of the alpha subdivision were found (51%), represented by \textit{Methylomonas}, \textit{Methylocistis}, \textit{Azospirillum} and \textit{Bradyrhizobium} species. An increase from 3% (in April) to 20% (in October) in abundance was observed for Gammaproteobacterial sequences, which were represented by species of \textit{Azomonas}, \textit{Celerinatantimonas} and \textit{Thermochromatium}.

Some \( nifH \) types were spread over all the soils at most or all of the sampling times. For instance, species of the genus \textit{Bradyrhizobium} were observed at least in two out of the three sampling times for all soils. \textit{Azospirillum} exhibited a similar pattern. In contrast, some specific organisms could be identified as being typical for soil type and sampling time. For instance, \( nifH \) gene sequences typical for \textit{Paenibacillus durus}, \textit{Stenotrophomonas maltophilia} and \textit{Dechloromonas aromatica} were observed only in soils B and D (sandy), whereas \textit{Azoarcus} sp. and \textit{Celerinatantimonas diazotrophica} were mostly detected in clayey soils K and G (Figure 5A and Figure S4).

\textbf{Figure 4. Unifrac analysis of \( nifH \)-gene libraries.} PCoA of \( nifH \)-gene sequences using 1921 randomly chosen sequences per sample, from normalized weighted analysis. doi: 10.1371/journal.pone.0074500.g004
From the total number of clusters obtained with CD-HIT (441), 66% represented rare species (288 clusters), which made up to 1.28% of the whole dataset. These rare members of the community also changed dramatically over time (Figure 5B and Figure S5). Intriguingly, most of the rare types were affiliated with Alphaproteobacteria (around 43%), followed by Beta- (33%) and Gammaproteobacteria (15%). Moreover, higher abundances of Firmicutes (3%) and Deltaproteobacterial (4%) members were found, as compared to the abundant types. Some nifH-types remained rare, for instance those affiliated with Acidithiobacillus, Anaeromyxobacter, Halorhodospira and Novosphingobium maintained relative abundances of approximately 0.005 (Figure 5B). Conversely, others became dominant at some point in time. For example, relative abundances of Azonexus increased from 0.009 to 0.067 in soil G from April to June, and those of Zoogloea increased from 0.007 in soil D in April to 0.97 in June (Figure 5B).

Drivers of nifH gene-based diversity

Forward selection on RDA analysis identified ammonium (5.0%; P = 0.032), soil pH (4.2%; P = 0.042) and clay content (4.3%; P = 0.042) as the main drivers of the nifH gene compositional variation of the abundant OTUs (Figure 6A). The same analysis based on the rare OTUs also identified ammonium (5.0%; P = 0.017), pH (5.1%; P = 0.009) and clay content (5.8%; P = 0.005), with the addition of nitrate (5.3%; P = 0.018) (Figure 6B).

We used Pearson’s correlation analysis to gain insights into the drivers of changes in richness, evenness and community dissimilarity (Bray, Curtis dissimilarity). Considering the abundant dataset, we observed significant direct or inverse correlations between soil pH and species richness (r = 0.728, P = 0.007), nitrate and evenness (r = -0.822, P = 0.007) and nitrate and dissimilarity (r = 0.540, P = 0.047) of N-fixing communities. Moreover, the relative abundance of Burkholderia species was significantly correlated to OM levels (r = 0.586, P<0.05), and those of Azooarcus species were driven by clay content (r = 0.747, P<0.01). Despite its ubiquity, no measured
soil parameter could explain the changes in the abundance of \textit{Bradyrhizobium} species.

**Discussion**

Nitrogen fixation is carried out by N-fixing microorganisms belonging to the Archaea and Bacteria phyla. Although some of these organisms might possess nitrogenase genes, they are not necessarily able to fix N, as the \textit{nif}H gene shares conserved regions with paralogous genes not involved in nitrogen-fixation (Clusters IV and V) [52,53]. Nevertheless, the \textit{nif}H gene is the best-known type of nitrogenase; it is the most sequenced and has become the marker gene of choice for phylogenetic, diversity and abundance studies of nitrogen-fixing community in different environments [54]. Taking all this into account, the main idea of this work was to report the temporal and spatial fluctuations of microorganisms harboring the \textit{nif}H gene, and therefore of the microorganisms that have the potential to fix nitrogen. To do so, we made use of degenerate primers to encompass the sequence variability of the \textit{nif}H gene [54], allowing for a greater coverage.

The population sizes of the N-fixing bacterial communities across four soils and three sampling times in the growing season were found to be within the range observed in other soil systems [47,55-57]. We observed significant seasonal fluctuations in these abundances, with a tendency to increase towards October. Temperature is known to be a key parameter that influences soil bacterial communities [58]. Moreover, along with this temperature effect, there is a variable influx of energy throughout the season. As nitrogen fixation is an energy-consuming process, it may tend to increase (in conjunction with a numerical increase of total bacterial communities) with increasing temperature and energy input, supporting the trend of a higher abundance from April (average air temperature 8°C) to June (14°C) and October (11°C). Soil nutrient status (mineral N) also changes seasonally, with generally higher levels in June, after application of fertilizers but before the intake of available nitrogen by the root system of prevailing plants [56]. This enhanced nitrogenous nutrient (ammonium and nitrate) level strongly decreased \textit{nif}H gene abundance, as previously reported in a managed ecosystem [27], and as suggested by field evidence. For instance, free-living nitrogen fixers in the soil will preferentially use available reduced N instead of fixing it [59,60] due to the high energy demand of the latter process. However, these effects are still controversial and some may reflect other processes, such as denitrification, as some studies have reported raised \textit{nif}H gene abundances in soils with high nitrate levels [21].

Soil pH was also identified as an important parameter controlling \textit{nif}H gene abundances. However, soil pH might be intrinsically linked to soil type, as clayey soils tend to have higher pH than sandy soils. In a microcosm experiment in which the effects of soil pH and clay content were analyzed separately, it was observed that an increase of clay content...
from 60% to 80% is stressful to nitrogen fixers, which showed a drastically decreased abundance [61]. This was not observed when soil pH was raised from 4.5 to 7.5, suggesting that the higher abundances of nifH in clayey soils are likely due to an effect of pH rather than soil texture. Moreover, an effect of soil pH on the abundance of diazotrophs has previously been observed [4]; it co-varied with season, with stronger correlations at the beginning than at the end of the season.

Seasonal shifts were also observed in the nifH gene based community structure, as previously reported [4]. In fact, we found -- through nifH gene-based PCR-DGGE analyses—that on average, around 60% of the diazotrophic communities shifted from April to October. This stands in contrast to studies based on PCR-DGGE which have shown substantial stability of the diazotrophic assemblage over an annual cycle [62,63]. In conjunction with a previous analysis of the same communities at a different sampling time [4], our results indicate an internal restructuring of the diazotrophic community. This was apparently not stochastic, and it occurred to a greater extent in the sandy soils than in the clayey ones. Moreover, variation partitioning showed that nitrate, ammonium and clay content were related to the structural changes. The effects of N availability on the diazotrophic community structures are as-yet not clear. Although some studies reported no correlation between community structure and N fertilization [63,64], we found a significant percentage of the community variation being explained by nitrate (28.6%) and ammonium levels (16.1%), which is in accordance with the physiology of nitrogen fixation (common response to N availability). Besides N, soil texture also represents a critical variable, as clay fractions in soils are important in forming micro aggregates [10]. Up to 70% of the free-living nitrogen-fixing bacteria are often found in soil clay fractions [65]. Occurrence in soil clay fractions is possibly favorable, as microaerophilic or anaerobic conditions prevail in such sites, providing more adequate environments for the oxygen-sensitive nitrogen fixation process.

In order to identify and characterize the members of the N-fixing communities in our samples, we developed a pyrosequencing approach based on the nifH gene. In a recent study, several primer sets designed to amplify nifH gene—including the ones used in this study—were explored in an in silico analysis of specificity and coverage was performed [54]. Here, we observed that the primers used in this study for the pyrosequencing of nifH gene did not amplify alternative nif genes, indicating that they are suitable for diversity studies of potential nitrogen-fixing microorganisms. Overall, our results confirmed the prediction that the nitrogen-fixing microorganisms would be highly diverse [47,66] and dynamic [4], consistent with the results from the PCR-DGGE analyses. The diversity and evenness values were strongly driven by local soil conditions. Furthermore, the evenness and dissimilarities of the communities were affected by nitrate levels, which led to very unequal communities. In a study of the effect of N inputs on soil bacterial communities, it was observed that as N inputs increased, communities became more distinct from those not receiving N fertilizer, and the response of bacterial diversity was inconsistent [67]. Furthermore, a clear influence of soil type and pH on the composition of the N-fixing communities was indicated by UniFrac analysis, confirming the results obtained by CCA analysis of the PCR-DGGE patterns. In this sense we observed that, whereas the presence or absence of different types was driven by soil pH or texture [61], other factors such as nitrate and ammonium drive the differential abundance of these types in each treatment (seen from weighted UniFrac).

Our pyrosequencing dataset revealed strong seasonal variations in the composition of the nifH gene community sequences, confirming previous results from PCR-DGGE data [4]. Temporal variability in the N fixers is often reported in marine environments [68-70], but it has been rarely studied in terrestrial ecosystems [4,47]. The sequences recovered were affiliated with several major groups of bacteria, which changed with sampling time. The sandy soils B and D showed the lowest percentage of shared species between sampling times and hence large fractions of the nitrogen-fixing communities in such soils may be under constant replacement, with some populations disappearing and reappearing during the time period analyzed. Similar results have recently been provided by Gobet et al. [70] who studied the effects of seasonality and sediment depth on the fluctuations of total bacterial communities of marine sands. In this study, when time was considered alone, only 18-37% of unique OTUs were shared between any two sampling times. In our data set, however, some types revealed clear trends in their seasonal dynamics. For instance, nifH genes allocated in the class Gammaproteobacteria (mostly Methyllococcales, Pseudomonadales and Chromatiales) increased significantly in the course of the season. This increase could be explained by an increase in soil nutrients in June and October when compared to April, due to the presence of plants and root exudation. As previously suggested, this response might reflect the copiotrophic behavior of members of this group, which tend to be favored under nutrient-rich conditions [71,72].

Cosmopolitan nifH species were identified as species that were present across time and soil type. In this class, we invariably found Bradyrhizobium, Azospirillum and Methylcystis to be in high abundance, which is consistent with data presented earlier [4,73,74]. The dominance of nifH gene types affiliated with Alpha proteobacteria across all soils and sampling times was consistent with the presumed dominance of this group, observed for terrestrial [75], and marine environments [70]. The high abundance of sequences affiliated with Bradyrhizobium is noteworthy, as it is known as a symbiotic N fixer. In addition, there is an increasing understanding that this organism may also play key metabolic roles as a soil saprophyte, being able to colonize soil and rhizosphere in the presence or absence of plants [76], under which condition they may even denitrify [77]. The fields we investigated are used for potato cropping and were subjected to crop rotations including non-leguminous plants. It is highly likely that the Bradyrhizobium types found are excellent survivors across the diverse conditions applied in these farming systems. Azospirillum, on the other hand, is a genus encompassing non-symbiotic (yet plant-associated) N-fixers known to be associated with roots of grasses, cereals, food crops and soils [78]. Our results corroborate previous evidence
based on cultivation techniques which highlight the cosmopolitan distribution of this diazotrophs [79,80].

We also detected particular (“core”) species in each soil, that is, species that were observed only in a particular soil, soil type and/or at a particular sampling time. Such core species could represent types that are unique to a particular soil or condition, but are sensitive to environmental variations (e.g. *Paenibacillus, Burkholderia* and *Celerinatantimonas* species). The low percentage of shared OTUs among the soils suggested that each soil, with its own structure and characteristics, harbored a particular N-fixing community. Indeed, in previous work, soil type has been reported to be the primary driver of soil bacterial composition [81]. Here, we posit that the community make-up of soil N fixers is determined by the (long) history of the soil under study. The community will consist of free-living and plant-associated nitrogen fixers, as niches for such functional groups will commonly exist in soils. In line with our data, fluctuations in the communities occur regularly, and groups that dominate at one point in time may have dwindled away in a next time point. Overall, however, there will be selection for the function at times, i.e. when reduced nitrogen becomes limiting for the soil biota. This incidental selection for function will tend to favor organisms that thrive under temporally defined conditions of temperature, pH and OM levels, and hence we may at the basis of the strongly differing N-fixing communities observe such differential selections for the very same function. In other words, function is maintained in the community, phylogeny of the underlying players is not.

Thus our results indicate an interaction between sampling time and environment with respect to effects on the evenness of the community, causing an increase in abundance of dominant OTUs in detriment of others at the beginning of the season. In the present study, three main environmental variables were identified as main drivers of the nitrifying communities, i.e. ammonium, pH and clay content. Indeed, although the *nif*H gene based diversity was high in the clayey soils, a lower turnover rate through time was found. An effect of ammonium on nitrogen fixation has been recognized for a long time, as nitrogenase is commonly inhibited by the presence of NH₄⁺ ions [82,83]. The abundant *nif*H-carrying bacterial groups are thought to be well-adapted to their environment [84]; however, rare diazotrophic species also responded to the same environmental parameters, indicating their functional importance. Thus the physicochemical properties of the soils might select both the dominant and the rare types, corroborating Hsu and Buckley [3]. Notably, some rare members increased their abundance at some point in time, when appropriate conditions were met, thus acting as source populations.

This is the first in-depth sequencing analysis of the *nif*H gene in agricultural soils along a temporal scale (one growing season), providing a wealth of new data and insights into the extent diazotrophic communities. The dominant and rare *nif*H gene types showed great spatial and temporal variations, mainly due to different N availabilities and soil pH levels. While some *nif*H gene types were always present (e.g. *Bradyrhizobium*), others only became apparent when conditions presumably provided an adequate environment (e.g. *Paenibacillus*). Furthermore, the broad data set obtained in this study provides a basis for directed studies on individual performance, e.g. of the cosmopolitan versus specific *nif*H containing species.

**Ethic statement**

No specific permits were required for the described field studies. The locations are not protected. The field studies did not involve endangered or protected species.

**Supporting Information**

Figure S1. Richness of bacterial and N-fixing communities estimated as the number of DGGE bands. B, Buinen; D, Droevendaal; K, Kollumerwaard and G, Grebbedijk; Ap, April; Ju, June; Oc, October. (EPS)

Figure S2. Unweighted pair group method with average linkages dendograms of diazotrophic community associated with four soils obtained from analysis of DGGE fingerprints. (EPS)

Figure S3. Rarefaction analysis of the diversities of *nif*H gene in the four soils and three sampling. The OTUs were classified at 90% similarity cutoff based on amino acid sequences. B, Buinen; D, Droevendaal; K, Kollumerwaard and G, Grebbedijk; Ap, April; Ju, June; Oc, October. (EPS)

Figure S4. Taxonomic classification of the most abundant (“core”) *nif*H-gene sequences associated with the two sandy agricultural soils (B and D) at three sampling times, April, June and October. Multi-colored charts at the legend are shown for each sample correspondingly. (PNG)

Figure S5. Taxonomic classification of the less abundant (“rare”) *nif*H-gene sequences associated with the two clayey agricultural soils (G and K) at three sampling times, April, June and October. Multi-colored charts at the legend are shown for each sample correspondingly. (PNG)

**Table S1.** Soil characteristics measured in this study. (DOCX)

**Table S2.** PCR and cycling conditions for PCR-DGGE analysis and real time quantification of *nif*H gene. (DOCX)

**Table S3.** PCR and cycling conditions for pyrosequencing of *nif*H gene. (DOCX)
Table S4. Forward and reverse barcoded primers used for pyrosequencing of nifH gene.

(DOCX)

Table S5. Richness estimates and diversity indices for forward and reverse amplicon libraries at 90% similarity cutoff.

(DOCX)

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Author Contributions

Conceived and designed the experiments: MCPS JDV JFS. Performed the experiments: MCPS. Analyzed the data: MCPS BS-H MS. Contributed reagents/materials/analysis tools: JDV EJS MS. Wrote the manuscript: MCPS JDV EJS.

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