

Absolute quantification of microbial taxon abundances

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Keywords: microbial ecology, flow cytometry, microbial community dynamics, amplicon sequencing, 16S rRNA gene, microbial biodiversity, microbiome

CONFLICTS OF INTEREST

The authors declare that there exist no conflicts of interest.

This document is the accepted manuscript version of the following article: Props, R., Kerckhof, F. M., Rubbens, P., De Vrieze, J., Hernandez Sanabria, E., Waegeman, W., ... Boon, N. (2017). Absolute quantification of microbial taxon abundances. *ISME Journal*, 11(587), 584-587. <https://doi.org/10.1038/ismej.2016.117>

20 **ABSTRACT**

21 High-throughput amplicon sequencing has become a well-established approach for microbial
22 community profiling. Correlating shifts in the relative abundances of bacterial taxa with
23 environmental gradients is the goal of many microbiome surveys. As the abundances
24 generated by this technology are semi-quantitative by definition, the observed dynamics may
25 not accurately reflect those of the actual taxon densities. We combined the sequencing
26 approach (16S rRNA gene) with robust single-cell enumeration technologies (flow cytometry)
27 to quantify the absolute taxon abundances. A detailed longitudinal analysis of the absolute
28 abundances resulted in distinct abundance profiles that were less ambiguous and expressed
29 in units that can be directly compared across studies. We further provide evidence that the
30 enrichment of taxa (increase in relative abundance) does not necessarily relate to the
31 outgrowth of taxa (increase in absolute abundance). Our results highlight that both relative
32 and absolute abundances should be considered for a comprehensive biological interpretation
33 of microbiome surveys.

34 Recent advancements in high-throughput sequencing of marker genes, such as the 16S
35 rRNA gene, have provided microbial ecologists the tools to accurately infer the relative
36 composition of microbial communities (Franzosa et al 2015). This resulted in a widespread
37 application of the technology in longitudinal studies where shifts in community structure are
38 related to environmental variables and functional outputs (Faust et al 2015, Wilhelm et al
39 2015). An inherent limitation of the sequencing technology is that the calculated taxon
40 abundances comprise relative values (Widder et al 2016). Hence, caution must be taken with
41 the biological interpretation of these values, since inter-sample differences in cell density are
42 not considered. To our knowledge, there are no descriptive studies that assess the extent to
43 which relative abundances deliver a skewed image of the actual microbial community
44 dynamics. In this study we combined robust cell density measurements from flow cytometry
45 (Prest et al 2013, Van Nevel et al 2013) with the relative abundances derived from 16S rRNA
46 gene amplicon sequencing. We performed two extensive longitudinal surveys on the central
47 water reservoir of a cooling water system. This engineered freshwater ecosystem was
48 subjected to highly controlled operational phases (**Supplementary information (SI) and**
49 **dataset**). We quantified the absolute taxon abundances and assessed whether additional
50 insights could be attained with the combined approach.

51 Based on the sample-specified total cell density, the absolute taxon abundances were
52 calculated for each time point. Individual taxon densities ranged from 0.5 to 1 679 cells μl^{-1} .
53 Several inter-taxon differences became apparent by performing ordinary least squares
54 regression analysis for the three most abundant taxa, which originate from different
55 taxonomic clades (**Fig. 1**) (Newton et al 2011). We identified a significant difference between
56 OTU1 (betl-A clade) and both OTU2 and OTU3 (bacl-A clade; $p=0.044$ and $p=0.046$ for
57 OTU2 and OTU3, respectively), however, no significant difference was observed between
58 OTU2 and OTU3 ($p=0.51$). These findings suggest that identical relative abundances of
59 different taxa require taxon-dependent biological interpretation because they do not
60 necessarily reflect the same absolute abundances. To further verify the limitations of relative

61 abundances, we closely inspected the temporal trajectories of both clades (**Fig. 2**).
62 Throughout the two surveys, the betl-A clade (OTU1) displayed similar variation in relative
63 abundance (coefficient of variation (CV_{rel}) = 54%) and cell density (CV_{dens} = 50%), whereas
64 the bacI-A clade (OTU2 and OTU3) displayed distinct transient behaviour (CV_{rel} = 124%,
65 CV_{dens} = 172%). Overall, both surveys were characterized by dynamic shifts in community
66 composition and density that were interspersed with stable periods (i.e., beginning and end
67 of each survey).

68 When interpreting the temporal trajectories of the relative and absolute abundances, two
69 primary discrepancies could be detected, potentially leading to misinterpretation if
70 conclusions would have been based solely on the relative abundances. First, during dynamic
71 community growth, such as the start-up and early reactor operation in survey 1, there was a
72 well-defined transition in absolute abundance, showing the systematic outgrowth and decay
73 of the bacI-A clade. In contrast, the relative abundance profiles were more ambiguous. They
74 remained relatively constant during the growth and decay event and only conclusively
75 indicated the beginning and the end. Another striking discrepancy could be observed in the
76 second half of reactor operation during survey 2. The relative abundance profiles displayed
77 an increase from $\pm 40\%$ to $>90\%$, potentially indicating a selective outgrowth of the betl-A
78 clade. If instead the absolute abundances were considered, there was no distinct pattern of
79 active outgrowth visible for this clade; its cell densities never considerably exceeded the
80 maximum density that was observed at the end of the start-up phase (relative abundance =
81 57%). This suggests that the decay of other taxa is responsible for this enrichment within the
82 community structure, or that environmental constraints limit its effective outgrowth. Overall,
83 the dynamics of the betl-A and bacI-A clades could be more precisely specified with the
84 absolute abundance profiles. Bulk cell density measurements were, in themselves, a poor
85 descriptive parameter of the microbial community dynamics (**Figure S11**). Only one OTU's
86 relative abundance (OTU2) was strongly correlated to the total cell density (Pearson's
87 correlation: $r_p = 0.60$, $p < 0.01$, $n = 79$), while the mean correlation strength for the entire

88 community composition was -0.08 ± 0.15 ($n = 427$). This shows that only for OTU2, its
89 outgrowth (increase in absolute abundance) frequently corresponds with its enrichment
90 (increase in relative abundance).

91 From our results we are able to show that absolute quantification of taxon dynamics is
92 essential and can shed additional light on many outstanding questions within microbial
93 ecology. Next to flow cytometry, quantitative PCR (qPCR) and fluorescence in-situ
94 hybridization (FISH) may represent alternative approaches for estimating absolute cell
95 densities. The tandem of qPCR and sequencing may be appealing because qPCR and
96 amplicon sequencing analyses start from the same DNA-extract and thus incorporate similar
97 laboratory-induced bias. However, for environmental samples qPCR is only sensitive enough
98 to separate twofold changes in gene concentration (proxy for cell abundance) (Smith and
99 Osborn 2009). qPCR also suffers from specific limitations such as amplification efficiency
100 and primer specificity, which makes it unadvised to compare the results between studies and
101 even assays on the same device (Brankatschk et al 2012, Smith and Osborn 2009). FISH
102 provides a PCR-independent approach for calculating relative taxa abundances or, in case of
103 a standardized methodological approach, even estimates of absolute abundances (Daims et
104 al 2001). Unfortunately, FISH analyses enumerate only the active fraction of the community
105 since the analysis is based on the hybridization of fluorescent probes with the 16S rRNA
106 (Amann and Fuchs 2008). These analyses are also more laborious, and generally provide
107 limited sample sizes (i.e., hundreds of cells). In contrast, our flow cytometric enumeration
108 approach of absolute cell densities is robust and high-throughput, but it requires supervised
109 denoising strategies to account for instrument and (in)organic noise, as well as cell
110 aggregates (materials and methods section, SI).

111 Methodological limitations become particularly crucial when the central hypothesis pertains to
112 the “*rare microbiome*”. This fraction is currently defined at arbitrary thresholds of 0.1 to 0.01%
113 (Lynch and Neufeld 2015). Although the definition of rare taxa will always remain partially
114 based on *ad-hoc* assumptions (Haegeman et al 2013), several bioinformatics-based tools

115 have shown the substantial impact of varying rarity thresholds on community analyses
116 (Gobet et al 2010). By also taking into account the absolute taxon densities, which are
117 comparable between studies, the development of a more consistent framework may now
118 prove possible. In this study, we did not take into account the number of 16S rRNA gene
119 copies during the calculation of the absolute abundances because there were insufficient
120 closely related reference genomes available (SI and figure SI2). For more characterized
121 environments this additional normalization may improve the resolution of absolute
122 abundance calculations (Langille et al 2013, Stoddard et al 2015). Overall, our results
123 demonstrate that when united, robust cell density measurements and phylogenetic marker
124 gene data are able to project a more comprehensive image of the compositional dynamics
125 occurring in microbial ecosystems.

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DATA AVAILABILITY

Flow cytometry data (.fcs format) are available on the FlowRepository archive under repository ID FR-FCM-ZZNA and the Dryad Digital Repository (<http://dx.doi.org/10.5061/dryad.m1c04>). Sequences are available on the NCBI Sequence Read Archive (SRA) under accession number SRP066190.

ACKNOWLEDGEMENTS

This work was supported through the Inter-University Attraction Pole (IUAP) “μ-manager” funded by the Belgian Science Policy (BELSPO, P7/25) and Geconcerteerde Onderzoeksactie (GOA) from Ghent University (BOF15/GOA/006). RP is supported by Ghent University (BOFDOC2015000601) and the Belgian Nuclear Research Centre (SCK•CEN). JDV and EH-S are supported as postdoctoral fellows from the Research Foundation Flanders (FWO- Vlaanderen). F-MK is supported by the Ghent University Geconcerteerde onderzoeksactie (BOF15/GOA/006). PR is supported by Ghent University (BOFSTA2015000501).

199 **AUTHOR CONTRIBUTIONS**

200 RP performed all the laboratory work and data analysis. PR performed data analysis. All
201 authors interpreted the data and wrote the manuscript.

FIGURE LEGENDS

Figure 1: Scatter plot of the absolute and relative abundance of the three most abundant OTUs registered at 79 time points and throughout two time-separated 40-day surveys of a secondary cooling water circuit that operates on a nuclear test reactor. The variance in the relation between absolute and relative abundances increases at elevated values (Breusch-Pagan test, $p < 0.0001$). OTU1 belongs to the betl-A clade. OTU2 and OTU3 belong to the bacI-A clade. Colored dashed lines depict ordinary least squares regression lines for each OTU. These regressions were used solely for statistical inference and do not necessarily represent the optimal predictive models for these data.

Figure 2: Temporal dynamics for taxa of the two most abundant freshwater clades (i.e., bacI-A (OTU2, red; OTU3, orange) and betl-A (OTU1, blue)) during two time-separated 40-day surveys of a secondary cooling water circuit that operates on a nuclear test reactor. The top panel displays the relative abundances (in %) inferred from the 16S rRNA gene amplicon sequencing data. The bottom panel displays the absolute OTU abundances (in cells μl^{-1}) and the circle labels represent the total cell density of the microbial community (in cells $\mu\text{l}^{-1} \pm$ standard deviation). Horizontal stacked bars highlight different phases of the system during surveillance. Gray zones indicate time periods where the cooling water system was not in operation (control phases), green zones indicate the startup and blue zones indicate steady-state operation.

Absolute abundance (cells μl^{-1})

- OTU 1
- OTU 2
- OTU 3



