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1 **Unraveling key environmental drivers and microbial biomarkers in the rhizosphere of mature**
2 **golf course putting greens.**

3 Giol, X^{*1}, M. Viñas¹, J. Romanya², K. L. Dodson³, M. Guivernau¹, Y. Lucas¹, M. Carreras¹

4 ¹ IRTA (Institute of Agrifood Research and Technology), Sustainability in Biosystems Programme. Caldes de
5 Montbui (Barcelona), Spain

6 ² UB, University of Barcelona, Faculty of Pharmacy and Food Sciences. Department of Biology, Healthcare
7 and Environment, Barcelona, Spain

8 ³ Syngenta AG Corporation, Basel, Switzerland

9 ***Corresponding author:** xavier.giol@irta.cat

10 **Keywords:** turfgrass, biomarkers, environmental drivers, soil microbiota, 16S rRNA-based metataxonomy

11 **Abstract:**

12 Unravelling the role of environmental drivers and native microbial communities in sandy soils of golf courses'
13 putting greens is imperative for a more sustainable turf management practices. Harnessing this knowledge may
14 aid in developing targeted strategies that promote beneficial microbial populations, enhance nutrient cycling,
15 suppress plant pathogens, and ultimately contribute to the overall health and resilience of golf course
16 ecosystems. In this project, the soil rhizosphere microbial community of the Golf de Pals (Girona – Spain)
17 putting greens is assessed in a chrono sequence of 14 and 56-years old greens. 16S rRNA and ITS2 paired end
18 amplicon sequencing (16S-metabarcoding) was used to determine both soil bacterial and fungal community,
19 respectively, in a 2-year long trial to determine microbial taxon richness, community composition, and
20 abundances of taxa involved in N and C cycling and other PGPR (Plant growth promoting rhizobacterium)
21 traits.

22 Age of putting greens, location, season, physical and chemical soil parameters were assessed to unravel
23 potential environmental drivers with a critical impact on turfgrass management and the diversity of the
24 rhizosphere microbiota. The analysis of alpha diversity on bacterial and fungal communities showed a seasonal
25 effect, increasing bacterial diversity in autumn, and fungal diversity in spring. No effects on Beta Diversity
26 due to season (Permanova $F=1.575$, $R^2=0.0571$, $p_{\text{value}}=0.112$), but a significant effect by the age (Permanova
27 $F=11.949$, $R^2=0.3149$, $p_{\text{value}}=0.001$) and location was revealed (Permanova Green 1 vs Green 5 $F=1.974$,
28 $R^2=0.1098$, $p_{\text{value}}=0.003$; Green 1 vs Green 14 $F=9.894$, $R^2=0.367$, $p_{\text{value}}=0.0015$; Green 5 vs Green 14 $F=8.409$,
29 $R^2=0.331$, $p_{\text{value}}=0.0015$). Main phyla were *Proteobacteria*, *Actinobacteria*, *Chloroflexi*, *Acidobacteria*,
30 *Planctomycetota*, and *Desulfobacterota*, with significant differences depending on the location of the putting
31 green. Mantel Test revealed that the environmental parameters with higher and significant contribution to soil
32 microbial diversity were solar radiation - PPFD (Photosynthetic photon flux density) and physicochemical
33 parameters such as, in order of importance, soil moisture and temperature, EC, OM, OC, N_{Kjeldahl}, NO₃⁻, P_{Olsen},
34 P_{Total} and Sulphates). Main phyla associated to soil parameters were *Crenarcheota*, *Acidobacteria*,
35 *Desulfobacterota*, *Chloroflexi*, *Actinobacteria*, *Firmicutes* and *Gemmatimonodata*, whereas FAPROTAX
36 assessment revealed that the main potential metabolic pathways associated with the most predominant
37 microbial community were nitrite respiration, nitrous oxide denitrification, nitrite denitrification,
38 denitrification, and dark sulphide oxidation and methanogenesis.

39 The combination of classical physicochemical monitoring tools with the emerging molecular microbial tools
40 allows us to identify the impact of turf management, not only in the microbial diversity and some key
41 biomarkers, but also on the whole metabolic potential traits and the ecosystem services of soil and rhizosphere
42 microbiota.

43 **Introduction:**

44 The rhizosphere, a dynamic interface between plant roots and soil, harbors a diverse microbial community
45 essential for plant health and ecosystem function. Rhizosphere microbes can enhance biocontrol, promote plant
46 growth, and increase tolerance to biotic and abiotic stresses (Amoaa et al. (2023) ; Conrath et al. (2006) ;
47 Berendsen et al. (2012) ; Lugtenberg & Kamilova (2009)) and biomarkers on turf management practices to
48 optimize soil health and golf course putting green performance.

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52 **Materials and methods:**

53 Location of the case study, putting greens and crop management strategies: As a case study to assess the effects
54 of environmental, physical and chemical soil parameters on the soil quality and microbiological diversity, three
55 putting greens of *Golf de Pals* golf course (41°59'39.95" N, 3°11'32.41" E, Pals – Spain, 10 m above sea level,
56 T_{\min} 4 °C, T_{\max} 28 °C, Rainfall – 562,3mm, Annual Daylight – 4457 h) were selected. All three putting greens
57 were under the same management practices, aged ranging from 14 years (Green #14) to 56 years old (Green
58 #1 and #5). Turf species on the putting greens were a mixing of *Poa annua* and *Agrostis stolonifera*, with an
59 ascending presence of *Agrostis stolonifera* from Green #1 (20%), Green #5 (45%) to Green #14 (90%).

60 Soil collection and physicochemical characterization of soil: Two soil sampling campaigns, in spring and in
61 autumn 2022, with five replicate per putting green, coring to a depth of 10 cm. All cores were homogenized to
62 form composited sample (5 composited samples per green and a season), chilled to 4-8 °C during the sampling
63 time, kept at 20°C in the lab for further microbial analyses. A complete physicochemical assessment was
64 performed by a reference laboratory: Soil Organic Content (SOC) (% of dry matter), Oxidizable organic matter
65 (% of dry matter), Nitrogen Kjeldahl (% of dry matter), N-NO₃⁻, N-NH₄⁺, P (Olsen), P (acid extract), K⁺, pH,
66 Electric Conductivity (EC, dS/m), SO₄²⁻. All statistical analyses were carried out with R Statistical Software
67 version 4.4.2.

68 Microbial community assessment of soil samples: DNA extraction of soil material from each composite sample
69 (n=5), was performed by using PowerSoil™ DNeasy Isolation Kit (Qiagen), according to the manufacturer's
70 instructions. Gene copy numbers of 16S rRNA (total bacteria), ITS1 rRNA (total fungi) were quantified by
71 quantitative real time PCR (qPCR). To assess the bacterial diversity of soil samples, 16S rRNA amplicon (V4-
72 V5 region) gene libraries were sequenced in a MiSeq equipment (Illumina). Primers were removed from the
73 demultiplexed fastq files by using Cutadapt software, and the paired reads were filtered and trimmed, denoised
74 and merged using the R package DADA2. The taxonomic affiliations of the ASVs were assigned by using the
75 SILVA v138.2 database and compiled into each taxonomic level.

76 To assess alpha diversity, Shannon (H), Inverted Simpson (I/D), Chao 1 indexes, were calculated in rarefied
77 samples by using Microeco R package. The dissimilarity in overall community composition among samples
78 (beta diversity) were calculated with Bray-Curtis distance and ordinated in PCoA. The contribution of
79 environmental and physicochemical parameters to change the microbial community structure were assessed
80 by means of non-parametric ANOSIM from total ASVs rarefied distributions, by using Vegan R package.
81 Differential abundance analysis of representative ASVs was performed by LEfSE (Linear discriminant
82 analysis Effect Size) to find biomarkers of groups and sub-groups. The contribution of environmental data to
83 the beta diversity of the rhizosphere microbial community structure (ASV level) were tested by means of
84 Mantel tests (Vegan R package).

85 To assess the most potential metabolic pathways in the rhizosphere, the prediction of functional profiles based
86 on the prokaryotic communities from sequencing results was performed by FAPROTAX (Louca et al. (2016);
87 Sansupa et al. (2021)) with MicroEco R package, matching the taxonomic information of prokaryotes against
88 the database to predict the traits of prokaryotes on biogeochemical roles.

89 **Results and discussion:**

90 Physicochemical characteristics of soil and their influence on the microbial diversity: The environmental
91 characteristics of each green and the age factor showed a significant impact in the species composition and
92 diversity, without season effect on beta diversity (ANOSIM Results – Table 1, NMDS Plots – Figure 1).

93
94 Table 1. ANOSIM. Environmental characteristics of putting greens

	R	P adjusted
Green 1 vs Green 5	0.229	0.001*
Green 1 vs Green 14	0.868	0.001*
Green 5 vs Green 14	0.855	0.001*
14 years vs 56 years	0.937	0.001*
Spring vs Autumn	0.073	0.073

100



101 Figure 1. NMDS (Non-metric Multi-dimensional Scaling ordination plot) - Stress value: 0.0826

102

103 No significant differences on gene copy numbers of 16S rRNA (total bacteria) and ITS rRNA (total fungi)

104 quantified by quantitative real time PCR (qPCR).

105 Soil moisture, electric conductivity, solar radiation – PPFd, soil organic carbon (SOC) and matter (SOM), N

106 Kjeldahl, NO_3^- , Total P, P Olsen (soluble) and sulphates, were identified as the main environmental drivers

107 that more contributed shifting the microbial community structure (Mantel Test results – Table 2).

108 Table 2. Physicochemical data of the main soil drivers influencing the beta microbial diversity in the turf rhizosphere.

	Green #1 mean \pm SE	Green #5 mean \pm SE	Green #14 mean \pm SE	P value	Mantel Test Statistic	Mantel Test (p-values)
Soil Moisture (%)	21.95 \pm 0.017b	18.90 \pm 0.200c	22.60 \pm 0.00a	< 0.001	0.297	0.002*
Soil Temperature ($^{\circ}\text{C}$)	16.48 \pm 0.404a	17.95 \pm 0.683a	17.30 \pm 0.533a	0.066	0.110	0.039*
PPFD ($\mu\text{mol/s.m}^2$)	246.20 \pm 28.93b	352.05 \pm 29.11ab	445.00 \pm 47.73a	0.024	0.241	0.005*
OC (%)	1.05 \pm 0.03b	1.22 \pm 0.055a	0.89 \pm 0.071b	0.001	0.193	0.023*
OM (%)	1.81 \pm 0.055b	2.10 \pm 0.096a	1.54 \pm 0.123b	0.001	0.198	0.018*
EC (dS/m)	0.20 \pm 0.14b	0.19 \pm 0.012b	0.28 \pm 0.012a	< 0.001	0.352	0.003*
N Kjeldahl (%)	0.13 \pm 0.003b	0.15 \pm 0.006a	0.09 \pm 0.006c	< 0.001	0.198	0.019*
NO_3^- (mg/kg)	5.27 \pm 0.469a	4.42 \pm 0.477a	3.42 \pm 0.478a	0.081	0.173	0.026*
P Olsen (mg/kg)	69.13 \pm 2.098a	43.63 \pm 2.23b	27.03 \pm 2.682c	< 0.001	0.334	0.000*
P Total (mg/kg)	793.33 \pm 44.82a	676.10 \pm 39.93ab	340.70 \pm 13.74c	< 0.001	0.565	0.000*
SO_4 (mg/kg)	50.33 \pm 7.063b	44.20 \pm 7.313b	199.90 \pm 34.88a	0.002	0.527	0.000*

109 Statistical tests for the significance of each parameter, and Mantel Test to determine its correlation factors with changes in the

110 microbial diversity (ASVs distribution) are enclosed. Significance: * P < 0.05

111

112 The relationship between environmental factors and microbial functional potentials predicted using

113 FAPROTAX, shows that soil temperature has a positive correlation with methanogenesis and nitrite

114 respiration, EC has a positive strong correlation with sulphur respiration and denitrification, and soil moisture

115 has positive correlations with methanogenesis and nitrate reduction. FAPROTAX assessment showed also a

116 positive correlation between soil organic carbon-organic matter with i) ureolysis, ii) aromatic compound

117 degradation and iii) cellulolysis functions, essential for the breakdown of complex organic molecules. Soil N-

118 NO_3^- and N-NH_4^+ were identified to correlate with i) nitrate reduction potential within the soil, ii) nitrification

119 and iii) denitrification functions, key functions in the nitrogen cycle. P_{Olsen} and P_{Total} have a positive correlation

120 with aerobic ammonia oxidation and nitrification, and soil SO_4^{2-} correlated significantly with the potential of

121 dark sulfide oxidation, and methanogenesis (both hydrogenotrophic and by CO_2 reduction with H_2).

122 Identification of Biomarkers: microbial abundance testing of representative ASVs was performed by "LDA

123 Effect Size (LEfSe)", identifying microbial biomarkers associated with specific conditions. LEfSe analysis

124 reveals that the genera *Nitrospira*, *Gaiella*, *Nordella*, and *Rhodococcus* are significantly enriched in the
125 GREEN 1, suggesting their importance in the environmental and management conditions of this group. In
126 contrast, the genera *Pedomicrobium*, *Ludemannella*, *Entotheonella*, *Gemmata*, and *Solirubrobacter* are
127 significantly enriched in the GREEN 5 group, while interestingly, anaerobic-related phylotypes were revealed
128 in Green 14 (*Sulfurifustis*, *Anaerolinea*, *Methanobacterium*) and aerobic methanotroph *Methilocystis*. In
129 addition, correlation assessment (Spearman) of environmental soil data and most predominant bacterial genus
130 revealed a higher impact of sulfate levels in soil, with the abundance of the genera *Anaerolinea*, *Luteolibacter*,
131 *Methanobacterium*, and *Sulfurifustis*. This suggests that sulfate availability plays a critical role in shaping soil
132 microbial communities and influencing functional traits such as methanogenesis, hydrogenotrophic
133 methanogenesis in case of hypoxia (edible organic matter and higher water content in soil), and dark hydrogen
134 oxidation.

135

136 **Conclusion:**

137 The combination of metataxonomic and ecological data analysis provides a powerful approach for
138 understanding the complex relationships between microbial communities and their environments. By using
139 bioinformatics tools like Mantel Test, LEfSe, and FAPROTAX, data-driven insights into environmental
140 drivers, potential biomarkers, and sustainable environmental management strategies can be derived to improve
141 turf and soil quality.

142

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144

145 **References:**

146

147 Eunice, A., Oluwaseyi A., Olanrewaju, S., Babalola, O., Ajilogba, C., Chukwuneme, Omena, C., Ojuederie,
148 Olawale, B. & Omomowa, I. (2023). The functionality of plant-microbe interactions in disease suppression.
149 Journal of King Saud University-Science. Volume 35, Issue 8. <https://doi.org/10.1016/j.jksus.2023.102893>

150

151 Sansupa, C., Wahdan, S., Hossen, S., Disayathanoowat, T., Wubet, T. & Purahong, W. (2021). Can We Use Functional
152 Annotation of Prokaryotic Taxa (FAPROTAX) to Assign the Ecological Functions of Soil Bacteria?
153 MDPI Appl. Sci. 2021 11(2), 688. <https://doi.org/10.3390/app11020688>

154

155 Lugtenberg, B., & Kamilova, F. (2009). Plant-growth-promoting rhizobacteria. Annu. Rev. Microbiol., 63, 541–556.
156 <https://doi.org/10.1146/annurev.micro.62.081307.162918>

157

158 Berendsen, R. L., Vismans, G., Yang Song, K. Y., de Jonge, R., Burgman, W., Burmølle, M., Herschend, J., Bakker, P.
159 & Pieterse, C. (2012). Disease-induced assemblage of a plant-beneficial bacterial consortium. ISME
160 Journal, 6(3), 490-499. <https://doi.org/10.1038/s41396-018-0093-1>

161

162 Louca, S., Parfrey, L.W. & Doebeli, M (2016). Decoupling function and taxonomy in the global ocean microbiome.
163 Science. 2016. 353(6305):1272-7. <https://doi.org/10.1126/science.aaf4507>

164

165 Conrath, U., Beckers, G.J.M, Flors, V., García-Agustín, P., Jakab, G., Mauch, F., Newman, M.A., Pieterse, C.M.J.,
166 Poinssot, B., Pozo, M.J., Pugin, A., Schaffrath, U., Ton, J., Wendehenne, D., Zimmerli, L. & Mauch-Mani, B.
167 (2006). Priming : Getting Ready for Battle. MPMI Vol. 19, No. 10, 2006, pp. 1062–1071.