

Microbiome-inspired phyllospheric synthetic communities increase microbial diversity and productivity of field-grown plants of *Agave tequilana*

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1 **TITLE**

2 Microbiome-inspired phyllospheric synthetic communities increase microbial diversity and
3 productivity of field-grown plants of *Agave tequilana*

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18 **ABSTRACT**

19 Agaves are plants native to North America that sustain life in arid and semi-arid ecosystems.

20 Previous studies revealed that cultivated plants of *Agave tequilana* had lower microbial
21 diversity and functionality compared to wild species. Here, we tested if microbiome-inspired
22 phyllospheric synthetic communities, based on microbial hubs or enriched phyllospheric
23 microbial functions, could increase microbial diversity and plant fitness in *A. tequilana*.

24 We applied seven functional and three random syncoms on one-year-old plants of *A.*
25 *tequilana* in the field and monitored their development for two years. Phylogenetic analyses
26 revealed that the composition and diversity of the phyllospheric prokaryotic and eukaryotic
27 communities associated with *A. tequilana* were affected by the season, with greatest
28 diversity observed during the dry months. The addition of functional syncoms promoted
29 phyllospheric microbial communities with a higher diversity, greater capacity to consume
30 carbon sources, and more integrated and complex networks. Additionally, plants treated with
31 functional syncoms had an increased content of sugars (°Brix, a measure of productivity in
32 agaves) in the stems, and higher leaf metabolic diversity.

33 Our work demonstrates that inoculation of syncoms formulated based on functional analyses
34 of microbiome data represents an effective biotechnological tool to improve the sustainability
35 and productivity of crops of arid ecosystems.

36 INTRODUCTION

37 Plants of the genus *Agave* originated in the arid and semi-arid regions of North America,
 38 reaching their peak of diversity and endemism in Mexico¹. Agaves have been used by
 39 humans since pre-hispanic times, and are intensively cultivated to produce alcoholic distilled
 40 beverages, such as Tequila produced from plants of *Agave tequilana* Weber var. Azul².

41 In 2020, production of Tequila for the local and international markets required the
 42 fermentation of 1.3 million tons of five-to-seven-years-old mature *A. tequilana* stems, and
 43 annual sales of Tequila reached on average of 51.2 billion MXP in 2019-21
 44 (<https://www.crt.org.mx>, <https://www.statista.com>). Plant productivity, and the consequent
 45 selling price of *A. tequilana*, depends heavily on the amount of sugars (measured as °Brix)
 46 and biomass in the stems. As the Tequila market continues expanding
 47 (<https://acevesspirits.com/>), producers of *Agave* plants require sustainable methods to foster
 48 plant productivity and health.

49 Our research on the *Agave* microbiome has revealed that microbial communities associated
 50 with wild and cultivated plants of this genus are diverse and structured, and that their
 51 composition and diversity is dependent on the plant compartment and the biogeography of
 52 the host plants³. Also, it has shown that the associated microbiome and their emitted volatile
 53 organic compounds can effectively promote growth and development of agaves and other
 54 plants^{4,5}.

55 Remarkably, we identified that cultivated *A. tequilana* showed reduced microbial diversity in
 56 the rhizosphere and phyllosphere compared to wild *A. deserti* and *A. salmiana*³. This lower
 57 diversity was also observed when analyzing microbial functions using shotgun
 58 metagenomics. Particularly, the phyllosphere of *A. tequilana* plants had a reduced number of
 59 genes related to phototrophy, carbon metabolism, secondary metabolites biosynthesis and
 60 xenobiotic degradation⁶. However, it was unclear if the addition of potentially keystone
 61 microorganisms could enhance microbial diversity, and if such an increase could improve
 62 plant fitness in this agriculturally managed *Agave* species.

63 In the present study, we designed seven functional synthetic communities (syncoms)
 64 inspired by two concepts: 1) keystone taxa resulting from network analyses (also called
 65 microbial hubs⁷, V. M. F.-N, Citlali Fonseca-García, Damaris Desgarennés and L. P. P.-M. *in*
 66 *press*, Fig S1a); and 2) taxa with enrichment in specified functional traits in the phyllosphere
 67 (e.g. phototrophy⁶, Fig S1b). These syncoms were inoculated in one-year-old plants of
 68 *Agave tequilana* in the field (Fig. S2, Table S1), and their impact on microbiome composition

69 and plant fitness were compared against three random syncoms (Fig S1c), and non-
70 inoculated plants (MOCK).

71 We employed amplicon sequencing of the 16S rRNA and ITS2 genes, Biolog profiling, and
72 plant developmental and physiological measurements (height, diameter, number of leaves,
73 number of rhizomatic suckers, sugars concentration in the stem and plant metabolomics) to
74 determine microbial composition, alpha and beta diversity, network complexity in the
75 phyllosphere, as well as plant fitness across a span of two years (Fig. S3).

76 Our work shows that the microbial communities associated with the phyllosphere of *A.*
77 *tequilana* respond to the season, being more diverse in the dry months. We demonstrate
78 that a subset of functional syncoms promoted microbial diversity and functionality, as well as
79 the complexity and integration of the phyllospheric microbial networks. These syncoms also
80 positively impacted plant productivity (°Brix) and significantly changed the leaf metabolome
81 of their host plants. In sum, our work shows how microbiome-based knowledge can be
82 successfully used to improve the effectiveness of biofertilizers in crops of drylands.

83 RESULTS

84 The phyllosphere of field-grown *A. tequilana*.

85 We analyzed the prokaryotic and eukaryotic communities associated with the phyllosphere
 86 of *A. tequilana* after the inoculation of syncoms inspired in the microbiome of *Agave* species.
 87 Syncoms were designed based on highly connected and central OTUs in correlation
 88 networks (PFC, PFCF, PFCS, and PFCT), enriched microbial functions in the phyllosphere
 89 (AAP, CC, BC) or by selecting strains randomly (RC1, RC2, RC3). The microbial
 90 suspensions were sprayed on the leaf surface of one-year-old, field-grown *A. tequilana*
 91 plants in November 2019 (t0), and the microbial communities were evaluated at 3-, 8-, and
 92 12-months post inoculation (t3, t8 and t12, respectively, Fig. S3).

93 Across all samples, 19 prokaryotic phyla were detected. Proteobacteria, Actinobacteria,
 94 Bacteroidetes, and Firmicutes comprised on average 94.7% of the relative abundance (Fig.
 95 S4a). Burkholderiales, Micrococcales, Cytophagales, and Bacillales were the most abundant
 96 orders in each phylum, respectively (Fig. 1a). Seven eukaryotic phyla were detected.
 97 Ascomycota and Basidiomycota comprised on average 87.9% of the relative abundance
 98 (Fig. S4b). Pleosporales and Tremellales were the most abundant orders in each phylum,
 99 respectively (Fig. 1b).

100 The order Caulobacterales was enriched in the rainy period (t8, Fig. 1a, Table S2) together
 101 with 26 OTUs, most of them classified as Proteobacteria (16, 61.5%) and Bacteroidetes (5,
 102 19.2%) (Table S3). During the dry period (t3 and t12), Kineosporiales and other low
 103 abundance orders were enriched (Table S2). Also, 222 OTUs including Actinobacteria (46,
 104 20.7%), Bacteroidetes (28, 12.6%), Firmicutes (18, 8.1%) and Proteobacteria
 105 (Burkholderiales 10%, Rhizobiales 12.6%, Sphingomonadales, 10%, Table S3).

106 Agaricales, Melanosporales and Filobasidiales were fungal orders enriched in the rainy
 107 period, while Chaetothyriales, Cystobasidiomycetes, Erysiphales and Trichosphaeriales
 108 were more abundant in the dry period (Fig. 1b, Table S2). Several OTUs were enriched in
 109 both seasons (238 dry, 138 rainy) belonging to diverse orders, but primarily Pleosporales or
 110 unclassified (Table S4).

111 PERMANOVA analyses of the binary Bray-Curtis distances between samples showed that
 112 the sampling season had the most influence on the microbial communities (Table S5). This
 113 factor explained 14.2% and 27.4% of the variance in the prokaryotic and eukaryotic
 114 communities, respectively. The sampling block and treatment had a significant but smaller
 115 influence on these communities. We observed that few OTUs were enriched in the syncom-

116 derived communities (Fig. S5). NMDS ordination analyses showed that most samples tend
 117 to cluster together in the prokaryotic communities, but the rainy season samples were the
 118 most dissimilar (Fig. 2a). In the eukaryotic communities, samples cluster more clearly based
 119 on the sampling season (Fig. 2b).

120 In the rainy season, there was a significant reduction in alpha diversity and an increase in
 121 the variability of the prokaryotic and eukaryotic microbial communities (Fig. 2). The position
 122 in the field did affect the diversity of the eukaryotic communities (Fig. S6), probably as a
 123 result of the different presence of weeds. We observed more weeds in block A than in B
 124 (Fig. S2a).

125 **Temporal variation of syncoms.**

126 The sequenced 16S-rRNA-V4 or ITS2 region of the 31 inoculated strains matched with at
 127 least one OTU (97% and 95% of similarity, respectively, Table S6). Based on the syncom
 128 composition, treatments were grouped in Network (PFC, PFCF, PFCS, PFCT), Metagenome
 129 (AAP, CC, BC), and Random (RC1, RC2, RC3) to evaluate the enrichment of the inoculated
 130 strains at t0. *Belnapia rosea* MJ22, *Methylobacterium* sp RAS18, and *Micrococcus*
 131 *aloeverae* RAT4 were more abundant in plants inoculated with Network syncoms than in the
 132 rest of the plants. These OTUs maintained a low abundance after inoculation (0.57, 0.12,
 133 and 0.02% of mean relative abundance, respectively, Fig. S7).

134 *B. rosea* MJ22, *Kineococcus radiotolerans* AS2.23, *Methylobacterium* sp. RAS18, *M.*
 135 *aloverae* RAT4, *Rhizobium* sp. RAS22, *Sphingomonas aerolata* AS2.4, and *Sphingomonas*
 136 *endophytica* AS3.13 were more abundant at t0 than at t3 only in plants inoculated with
 137 Network and Metagenome syncoms. Several strains had reduced abundance at t8 (rainy
 138 period) and increased again at t12 (Fig. S7).

139 **Syncoms increased the alpha diversity and microbial interactions in the phyllosphere.**

140 A subset of syncoms inoculated on the leaves of *A. tequilana* were observed to affect the
 141 diversity of their phyllospheric prokaryotic communities. The phyllosphere promoted by
 142 PFCS had a higher prokaryotic alpha diversity than control plants (MOCK, Fig. 3a, Fig. S8a).
 143 This effect was maintained until t8, when this phyllosphere showed the greatest increase in
 144 OTU richness and Shannon index (0.89 and 0.66 log2 fold change (FC) compared to MOCK,
 145 Fig. 3a, Fig. S8a). Syncoms PFC and BC also increased OTU richness in February (Fig.
 146 S8a). These syncoms did not influenced the diversity of eukaryotic communities. Only RC2
 147 increased the OTU richness (Fig. S8b), while CC increased alpha diversity at t3 compared to
 148 the control (Fig. 3b, Fig. S8b).

Community metabolic profiling was assessed in the second rainy season (June 2021, t19, Fig. 3c) as a proxy for microbial diversity and functionality. The phyllospheric communities of plants inoculated with PFC, PFCF, PFCS, and RC2 consumed a more diverse array of carbon sources than the MOCK-derived microbial communities. Most of these compounds corresponded to simple sugars and organic acids.

Correlation networks in the treated plants had more OTUs and correlations than non-inoculated controls (Fig. 4, Fig. S9a). Moreover, most syncoms increased prokaryotic-eukaryotic correlations and decreased the eukaryotic-eukaryotic ones (Fig. 4, Fig. S9b). The phyllosphere derived from PFCS had more than two times the number of hub OTUs compared to MOCK plants and the rest of the syncoms (Fig. 4, Table S7). Moreover, the genera present in syncoms were present as microbial hubs. OTU 43 (*Methylobacterium*) was a hub in non-inoculated and inoculated plants. OTUs classified as *Sphingomonas* (OTU 18, 89, 164, and 167), *Methylobacterium* (OTU 113), and *Nocardioides* (OTU 271) were microbial hubs in PFCS-inoculated plants. Finally, *Bacillus* (OTU 14) and *Massilia* (OTU 481) were microbial hubs in plants inoculated with PFCT (Fig. 4, Table S7).

Syncoms influenced the growth and physiology of field-grown *A. tequilana*.

The growth rate of plants increased in the first year after inoculation with syncoms (Fig. 5a), and no signs or symptoms of plant disease were recorded. The changes in height were greater in plants inoculated with PFCT and RC3 (0.86 and 0.54 FC, respectively, compared to MOCK with 1.54 cm/month) while plants inoculated with PFC and RC1 syncoms unfolded more leaves (0.4 and 0.33 FC, respectively, compared to MOCK with 2.14 leaves/month. Only the plants inoculated with the CC syncom grew at a higher rate than the control during the second year.

The plants inoculated with the AAP syncom diminished their growth. The growth rate of these plants was lower in both years (-0.86 and -0.51 FC in height compared to MOCK, Fig. 5a) and produced fewer rhizomatic suckers after 1 year (-0.64 FC, compared to MOCK with 2.8 rhizomatic suckers, Fig. 5b).

The concentration of sugars in the stem sap did not change significantly until two years after the inoculation with syncoms (t24). Plants inoculated with PFC, PFCS and AAP increased the concentration of the °brix in the stem sap by 0.64, 0.38, and 0.42 FC, respectively, compared to the control with 13.16 °Brix (Fig. 5c).

PFCS and PFCT syncoms influenced the plant metabolome of field-grown *A. tequilana*.

182 Untargeted metabolomics in the leaves of field-grown *A. tequilana* plants inoculated with
183 PFCS, PFCT, and MOCK was performed in t12. LC-MS² data extraction by MZmine and
184 filtering by Metaboanalyst retrieved 1,520 potential metabolites with reproducible
185 quantification. As shown in Fig. 6a, the first and second principal components separated
186 PFCS- and PFCT-treated plants on both sides of the graph, hinting at distinct biological
187 effects. Heatmap representation and hierarchical clustering analysis showed specific
188 patterns between datasets using the top-50 most significant features (Fig. 6b). The volcano
189 plots revealed differential effects by PFCS and PFCT on the *Agave* metabolome based on
190 the number of modulated metabolites (Fig. 6c,d). Globally, more metabolites were
191 modulated by PFCS (n=184) compared to PFCT (n=56), and of those, a low overlap among
192 syncom treatments (at the MS1 level) was inferred by UpSet plot analysis (Fig. 6e). As
193 shown in Fig. 6f, a higher repertoire of chemical classes assigned to the up- (36% annotation
194 rate) and down-modulated (30% annotation rate) metabolites were noted in the PFCS-
195 treated *Agave* plants compared to the PFCT group (Fig. 6e). The chemical classes
196 “organooxygen compounds” and “carboxylic acids and derivatives” were predominantly
197 down-modulated and up-modulated by PFCS treatment, respectively. The complete list of
198 putative annotated metabolites (MSI, level 3) aided by CANOPUS and CSI: FingerID tools
199 are described in Table S12 and Table S13. Due to the challenging task of characterizing the
200 entire metabolites (at the molecular structure) modulated by treatments, we further guided
201 our analysis based on molecular networking and spectral matching annotation (MSI, level 2)
202 by the GNPS online platform as previously shown⁸. Our strategy enabled us to confidently
203 annotate (aided by CSI: FingerID and MolDiscovery tools) a group of carbohydrate
204 derivatives belonging to the class of the “organooxygen compounds” that was down-
205 regulated by PFCS treatment, but not by the PFCT (Fig. S10). Further work should reveal in
206 which specific plant metabolic processes these compounds are involved in agaves

207 DISCUSSION

208 Our work shows that phyllospheric microbial communities associated with *A. tequilana* were
 209 affected by seasonal variations and remarkably displayed higher diversity during the dry
 210 months. This pattern contrasts with the decreases in bacterial and fungal diversity observed
 211 in the rhizosphere of *Opuntia ficus-indica* along an aridity gradient^{9,10}, suggesting that below-
 212 and above-ground epiphytic microbial communities respond differently to aridity. Other study
 213 in the Mediterranean semi-arid ecosystem found, similarly to our research, that fungal
 214 communities associated with eight sympatric and perennial plants differentiated based on
 215 the season, being more similar in the hot and dry summers than in the mild and wet
 216 winters¹¹.

217 Importantly, our work demonstrates that phyllospheric microbial communities associated with
 218 *A. tequilana* can be manipulated with the inoculation of microbiome-inspired syncoms. The
 219 syncoms based on networks (PFC, PFCF, PFCS, and PFCT) were formulated with taxa
 220 presenting high centrality and connectivity in correlation networks of the phyllosphere of wild
 221 *A. salmiana* and cultivated *A. tequilana* (Table S1, V. M. F.-N, Citlali Fonseca-García,
 222 Damaris Desgarennnes and L. P. P.-M. *in press*). Therefore, we expected them to act as
 223 keystone microorganisms that will have a major influence on the assembly of microbial
 224 communities^{7,12,13}. *B. rosea* MJ22 and *Methylobacterium* sp. RAS18 are aerobic anoxygenic
 225 phototrophs bacteria⁶ present in all network syncoms, but they alone can not explain the
 226 changes in diversity. The addition of other common keystone bacteria (as in PFC) resulted in
 227 a higher prokaryotic diversity at t3, highlighting the complementarity of diverse taxa, such as
 228 Actinobacteria and Firmicutes, as a force driving this effect. Moreover, the addition of *A.*
 229 *salmiana* strains (PFCS, *Sphingomonas* spp., *Frigoribacterium endophyticum* AS3.20,
 230 *Cellulomonas cellasea* RAS26, and fungi) was relevant for maintaining a high prokaryotic
 231 diversity at t8, when it reached the lowest values. The additive effect was absent when
 232 adding fungal strains alone (PFCF) or *A. tequilana*'s hubs (PFCT). This evidences is in
 233 agreement with the increase in microbial diversity shown in other plants after the inoculation
 234 of biofertilizers^{14,15}, but contrasting results have been reported¹⁶. Together, this evidence
 235 supports that microbial hub's influence on community structure depends on the arising
 236 microbe-microbe interactions and the environment^{7,17}.

237 The inoculation of syncoms resulted in more complex and integrated correlation networks.
 238 This phenomenon was also found in the bulk soil of wheat¹⁸, the phyllosphere of tobacco¹⁹,
 239 and the rhizosphere of tea¹⁵ plants after inoculation with native and non-native bacterial
 240 consortia, suggesting that the addition of native or non-native microorganisms can modify

241 the way that phyllospheric microorganisms interact, even in plants with established
 242 microbiomes. Remarkably, only the phyllosphere derived from PFCS had a greater amount
 243 of hub OTUs, some of them from the same taxa as the inoculated bacteria (*Sphingomonas*,
 244 *Methylobacterium*, *Nocardioides*). *Sphingomonas* spp. were the main component of PFCS,
 245 and *S. endophytica* AS3.13, *Methylobacterium* sp. RAS18, and *F. endophyticum* AS3.20
 246 correlated with the hub OTUs in the PFCS network (Fig S11), suggesting that the inoculated
 247 bacteria could be reinforcing native microbiota to act as hubs similar to the keystone
 248 reinforcement concept¹². This result also highlights the relevance of *Sphingomonas* species
 249 in the plant phyllosphere, a genus that has also been found as keystone taxa in *A.*
 250 *thaliana*²⁰. In contrast, the phyllosphere derived from PFCT had the same inoculated *M.*
 251 *aurea* as microbial hub, suggesting its adaptation to the cultivated plants of *A. tequilana*.

252 Importantly, our work corroborates the hypothesis that the augmentation of hubs from wild
 253 *A. salmiana* (PFCS) in the cultivated species *A. tequilana*, that has lower microbial diversity,
 254 contributes to a more diverse microbiome and a more integrated correlation network with
 255 relevant microbial hubs that resemble the microbiome of wild *Agave* species (V. M. F.-N,
 256 Citlali Fonseca-García, Damaris Desgarenes and L. P. P.-M. *in press*). These wild-like
 257 microbial communities could be more resistant to pathogen invasion and other
 258 environmental stresses and perturbations^{12,21,22}.

259 Other syncoms also showed increments in alpha diversity. BC syncom was formed with taxa
 260 that represented most of the enriched biofilm and quorum sensing (QS) genes in the
 261 phyllosphere⁶. Biofilm formation has been associated with changes in the plant microbiome²³
 262 and the ability of immigrant bacteria to colonize the leaves²⁴. Moreover, the genomic
 263 annotation of Network and Random syncom members have shown that they possess genes
 264 related to biofilm formation and QS (Fig S12), which warrants more research to assess their
 265 influence on community structure. Even RC2 increased the OTU richness of the eukaryotic
 266 phyllosphere, highlighting the possibility of functional redundancy or other microbial
 267 processes that were not targeted in our experimental design.

268 The selection of strains for bio-inoculants has been mainly based on the presence,
 269 frequency and intensity of canonical plant growth promotion traits (e. g. the selection of
 270 biofertilizers for arid lands²⁵) but without knowledge of their influence on the native microbial
 271 community or their persistence. Our work sets the precedent for the selection of functional
 272 strains based on their putative interactions within the microbial communities, and their ability
 273 to influence plant growth in agroecosystems. Both approaches can coexist now that multiple
 274 microbial genomes and amplicon sequencing data sets for several plant species are
 275 available. Here, we show that syncoms inspired in the *Agave* microbiome represent valuable

alternatives for agriculture, as they can persist in the plant tissues in the context of the environment and the indigenous microbiota²⁶. Additionally, the phyllosphere derived from PFC, PFCF, PFCS and RC2 were metabolically more diverse in the second rainy season (t19), suggesting a long-term effect of the inoculated syncoms under natural field conditions.

Microbiome research have postulated different hypothesis about the relationship between microbial diversity and plant fitness¹³. We did not find a significant correlation between plant growth and microbial diversity nor OTU absolute abundance (not shown). Instead, both functional (PFC, PFCT) and random communities (RC1, RC3) increase at least one growth parameter during the first year, suggesting that diverse microorganisms can influence plant growth independently from their network position or known functions. The unfolding of leaves in agaves correlates with the environmental productivity index²⁷, a parameter that was increased by both PFC and RC1 syncoms. Moreover, stem sugars are the primary raw material for Tequila production. Syncoms PFC, PFCS, and AAP increased the number of sugars in the stem after 2 years post-inoculation which is indicative of a higher yield. Contrary to most of the works demonstrating the plant growth promoting effect of *Methylobacterium*^{28,29}, in *A. tequilana* AAP bacteria alone reduced the growth of the plants, but the rest of the PFC bacteria supported their growth, highlighting the relevance of using more diverse microbial inoculants. Inoculated strains could have influenced plant development and growth directly via plant growth promotion traits, including VOCs⁴. Whether inoculated bacteria and fungi influenced growth directly or indirectly via attracting functional microorganisms¹² remains to be tested.

The inoculated syncoms (PFCS, PFCT), formulated with common and specific keystone taxa (*A. salmiana* and *A. tequilana*, respectively), interacted with the plants causing a differential production of leaf metabolites after 12 months of the inoculation. PFCS-inoculation modulated several benzene and steroidal compounds (e. g. saponins). These group of metabolites are common in plants of the genus *Agave*^{30,31}, and the phyllospheric microbiome of wild *A. salmiana* and cacti harbors abundant and diverse genes related to the degradation of these compounds (V. M. F.-N, Citlali Fonseca-García, Damaris Desgarennnes and L. P. P.-M. *in press*). The biggest group of down-regulated metabolites in PFCS plants were organooxygen compounds that include many carbohydrate derivatives. Despite not being able to link them to the biosynthesis of fructans, this decrease inversely correlated with the changes in sugar accumulation in the stem. Moreover, a steroidal saponin, commonly named as yuccalan, was found down-regulated in PFCS plants. This compound was discovered on the leaves of *Yucca* and has antifungal activity³². These results support the notion that syncoms not only influence microbe-microbe interactions, but also the metabolic

311 interactions of the host plant and the microbial community with long lasting effects. Further
312 work is needed to understand the plant metabolic processes that are more prone to be
313 affected by the associated microbiota, and their consequences for plant health and
314 productivity.

315 Altogether, our work offers direct evidence that the selection of microbial inoculants inspired
316 by microbiome knowledge is a promising avenue for improving and sustaining plant health
317 and productivity in keystone desert crop species, such as *Agave tequilana*.

318 ONLINE METHODS

319 Design of the field experiment

320 The field experiment was performed in collaboration with Tequila Real de Pénjamo S.A. de
 321 C.V. For this, rhizomatic suckers of *Agave tequilana* Weber var. Azul were planted in July
 322 2018 in a 1-ha lot located in Pénjamo, Guanajuato, Mexico (Lat. 20.387444, Long. -
 323 101.76425, Altitude 1768 m) (Fig. S2a,b). The distance between plants was ca. 1 m within
 324 rows and ca. 3 m between rows. The rhizomatic suckers were treated with a pre-emergent
 325 herbicide (Diuron) and a foliar pesticide (Malathion) before planting. Watering was
 326 completely dependent on the seasonal rains and no agrochemicals were applied for two
 327 years, until July 2020 when the managers applied an insecticide to prevent the development
 328 of *Scyphophorus acupunctatus* larvae (picudo).

329 The field was divided into two blocks east (A) and west (B). Each block was divided into
 330 squares of 4 x 20 plants = 80 plants. Block A and B harbored 1000 and 1200 plants
 331 approximately and were divided into 12 and 15 squares, respectively. The 10 syncom
 332 treatments plus the non-inoculated control were randomized in each block using semi-
 333 random numbers.

334 Bulk soil samples were taken at the beginning of the experiment and after 12 months, both
 335 at five representative points for each block. Samples were taken 10–15 cm from the surface
 336 using a shovel and stored in clean plastic containers at 4°C. Soil physicochemical
 337 characteristics were determined by the company Fertilab (Celaya, Gto., Mexico).

338 Isolation of phyllospheric microorganisms

339 *Agave* leaf samples were processed as previously described³³ with some modifications. Two
 340 leaves per individual of three *A. tequilana* F.A.C. Weber and *A. salmiana* Otto ex Salm
 341 subsp. *crassispina* (Trel.) Gentry plants were collected in September 2017. *Agave tequilana*
 342 leaves were sampled from a cultivated population in Pénjamo, Guanajuato, Mexico, (Lat.
 343 20.686, Long. - 101.875, Altitude 1714 m, Tequila Real de Pénjamo) while *A. salmiana*
 344 leaves were sampled from a wild population in San Felipe, Guanajuato, Mexico (Lat. 21.766,
 345 Long. -100.163, Altitude 2089 m)). Leaves were obtained using gloves with a sanitized knife.
 346 Samples were transported in new brown paper envelopes and maintained in refrigeration
 347 until processing. In a laminar flow hood, the leaves were washed with sterile distilled water,
 348 then the leaf surface was brushed constantly with a sterile toothbrush and 50 ml of removal
 349 buffer (6.75 g KH₂PO₄, 8.75 g K₂HPO₄, and 20 µL of Triton X-100 per liter) inside of a 2 L

350 plastic sterile beaker. The resulting suspension was transferred to a conic tube. The pellet
 351 was obtained by centrifugation at 5000 rpm at 18°C for 15 min and resuspended in 5 ml of
 352 removal buffer.

353 Phyllosphere suspensions were ten-fold diluted in saline solution (NaCl 0.85%). A 100 µL
 354 aliquote of each dilution was plated onto TSA (Difco™), Czapek-Dox (Merk™ formulation) ,
 355 ISP3 (HIMEDIA formulation), ISP5 (HIMEDIA formulation) for heterotrophic bacteria, and
 356 PDA (Difco™) for fungi. Plates were incubated for 1-3 weeks at 28°C. Bacterial-like colonies
 357 with different morphologies were sub-cultivated on TSA, while filamentous fungi and yeast-
 358 like colonies with different morphologies were sub-cultivated on PDA. Only microorganisms
 359 with a unique colony and microscopic morphology were selected for further analyses.

360 Axenic and monoclonal cultures of bacteria and yeasts were obtained by dilution and plating
 361 on TSA plates, while axenic cultures of filamentous fungi by subculturing small mycelial
 362 plugs in PDA plates. Strains were stored in glycerol 40% at -80°C and -20°C for long and
 363 medium-term storage, respectively

364 Bacterial and fungal strains were identified by amplifying and sequencing the 16S-rRNA
 365 gene and the ITS, respectively, as previously reported³³ (Table S8).

366 **Selection of synthetic communities (Syncoms)**

367 Synthetic communities were designed based on previous analyses of the *Agave* and cacti
 368 microbiome (Table S1). Highly connected and central taxa (microbial hubs) were identified in
 369 correlation networks of amplicon sequence data (V. M. F.-N, Citlali Fonseca-García,
 370 Damaris Desgarennnes and L. P. P.-M. *in press*) and phyllosphere enriched functions were
 371 identified using comparative metagenomics⁶. Microorganisms were selected from the culture
 372 collection based on their 16S-rRNA or ITS2 sequence identity to the species level (97% and
 373 95%, respectively) or their genera when no exact matches were found.

374 Predicted functional communities (PFC) were composed of taxa that were consistently
 375 highly connected and central in both wild *A. salmiana* and cultivated *A. tequilana* correlation
 376 networks. Other microorganisms were incorporated to test the additional influence of fungal
 377 hubs (PFCF), *A. salmiana* specific hubs (PFCS), and *A. tequilana* specific hubs (PFCT). A
 378 group of aerobic anoxygenic phototrophs (AAP), cyanobacteria (CC), and known biofilm-
 379 producing Proteobacteria (BC) were selected to represent the phyllosphere enriched
 380 functions. Randomly selected syncoms were designed by selecting strains from the
 381 collection using semi-random numbers (RC1 and RC2) or by a personal selection of strains
 382 (RC3) (Table S1).

383 The partial 16S-rRNA sequences of the selected bacteria were submitted under the
 384 accession numbers ON217552-ON217580. The partial ITS sequences of the selected fungi
 385 were submitted under the accession number MT488313 and MT488312 (Table S8).

386 **Cultivation of bacterial and fungal strains for syncoms inoculation**

387 Bacteria and yeast strains were recovered from storage in TSA plates after 1-5 days at
 388 28°C. Pre-inocula were prepared by inoculating one colony in 20 ml or 50 ml of liquid media
 389 (TSB Difco™), for rapid and slow-growing strains, respectively. Liquid cultures were
 390 incubated from 1 to 5 days at 28°C with constant shaking (ca. 120 rpm). The pre-inocula
 391 were inoculated (1% v/v) in the appropriate amount of fresh liquid media and incubated from
 392 1 to 3 days at 28°C with agitation (120rpm). All the cultures were taken out from incubation
 393 in the exponential phase at the same time and the OD was determined at 600 nm using a
 394 Spectrophotometer (Table S9). The purity and axenity of pre-inocula and inocula were
 395 assessed visually with the color of the pellet after centrifugation (10000 rpm 2 min), by sub-
 396 cultivation in TSA, and by gram staining.

397 *Cladosporium*. sp. cutV was recovered from storage in PDA plates after 7–14 days at 28°C.
 398 It was inoculated in fresh PDA plates using a sterile toothpick and incubated for three weeks
 399 at 28°C until the development of conidia. Conidia were harvested with 5 ml of Triton-100X
 400 0.1% per plate, the suspensions of conidia were then filtered through a sterile polyester
 401 mesh to remove big chunks of mycelia and finally counted in a Neubauer chamber. Pre-
 402 inoculum and inoculum axenity were determined by macroscopic and microscopic
 403 morphology.

404 We were unable to accurately quantify the number of colony-forming units of cyanobacterial
 405 strains because they are filamentous and formed biofilms in liquid and solid media.
 406 Filaments of 3-week-old *Leptolyngbya* sp. AT32P and *Nodosilinea* sp. AT21F cultures were
 407 inoculated in 1L of fresh Gromov-21 media each and incubated for 3 weeks in a photoperiod
 408 room of 16/8h at 28°C. The purity of the pre-inocula and inocula was determined by
 409 macroscopic and microscopic morphology.

410 **Inoculation of plants on the field**

411 Syncoms were formulated the day before inoculation. The amount of bacterial culture
 412 necessary to prepare 2.5 L (500 ml x 5 plants) of syncoms suspensions with 1×10^6 CFU/ml
 413 of each strain was calculated based on the OD of each liquid culture and conidia
 414 suspension. These aliquots were mixed to obtain two concentrated suspensions of each
 415 syncom (50 ml, 1.25×10^9 CFU/ml per strain). Concentrated syncoms were centrifuged

416 (10000 rpm, 10 min) to remove remaining culture media, resuspended in NaCl 0.85%, and
 417 kept at 4°C.

418 The presence of the strains was validated with a viable count of the mixtures (Table S10).
 419 Bacterial and fungal colonies were counted based on their morphological features.
 420 Cyanobacteria formed a biofilm in the plates, and we were unable to count them
 421 appropriately.

422 Plants were inoculated in November 2019. On the day of the inoculation, syncoms
 423 suspensions were kept on ice until needed. Each suspension was diluted in 1.2 L of tap
 424 water in clean plastic bottles and mixed to break any clumps. Each plant was sprayed with
 425 50 ml of the diluted syncoms suspensions (1×10^6 CFU/ml each strain) using a hand sprayer.
 426 Approximately, 24 plants were sprayed per treatment (8 plants per row) in each block. Non-
 427 inoculated or control plants were sprayed with tap water (MOCK). The first two plants in
 428 each row and one row between squares were not treated in any way to create a biological
 429 barrier between treatments (Fig. S2c).

430 **Plant measurements and sampling**

431 The number of leaves, as well as the diameter and height of the *A. tequilana* plants was
 432 measured in November 2019 (dry season, fall, t0, beginning of the experiment), February
 433 2020 (dry season, winter, t3), July 2020 (rainy season, summer, t8), November 2020 (t12),
 434 and finally in June 2020 (rainy season, summer, t19). Plant measurements were taken from
 435 the same six plants per treatment to determine the growth rate in each treatment during the
 436 first year. From November 2019 to June 2021 only the height and width of 30 plants per
 437 treatment were recorded. The number of rhizomatic suckers was recorded at 12 and 24
 438 months post-inoculation (Figure S3).

439 The sugar content of the stem sap was estimated *in situ* in November 2020, June, and
 440 November 2021 (t12, t19, and t24, respectively, Figure S3). The stem was drilled with an 8
 441 mm drill, the extracted tissue was pressed in a small garlic press and the sap was collected
 442 on a refractometer. Six plants per square were measured.

443 Two leaves per individual plant were sampled to recover the phyllospheric microbial
 444 communities. At the t0, one individual per treatment in each block was sampled, while three
 445 individuals per treatment in each block were sampled at t3, t8, t12, and t19. In total 232
 446 individuals were sampled. Leaves from the second and third row of the *Agave* rosette were
 447 sampled as explained above.

448 **Assessment of phyllospheric communities**

449 **Sample processing**

450 The phyllospheric microorganisms were recovered by brushing the leaf surface with a
 451 toothbrush soaked in the washing buffer as explained above. The microbial pellet was
 452 covered in 5 ml of absolute ethanol and stored at -80°C until the DNA extraction. The
 453 metagenomic DNA was extracted using the DNAeasy Power soil kit (Qiagen) from t0, t3, t8
 454 and t12 sampling points. DNA was recovered in 75 µl of TRIS elution buffer. The quantity
 455 and quality of gDNA were recorded using a nanodrop and the recovered DNA was stored at
 456 -20°C.

457 The 240 samples were randomized in 96 well microplates in ranges of 50 to 600 ng of DNA.
 458 The samples were sent to *AIM – Advanced Identification Methods* in Leipzig, Germany
 459 where the metabarcoding sequencing was performed. The 515F/816R and ITS9F/ITS4R
 460 primers were used for the amplification of the 16S-rRNA-V4 and ITS2 regions, respectively.
 461 Peptide nucleic acid (PNAs) were added in the 16S-rRNA-V4 reactions to reduce the
 462 amplification of plant chloroplast and mitochondrial DNA³⁴. After ligation, the paired-end
 463 2x300 bp sequencing was performed in an Illumina MiSeq v3 instrument.

464 Data was demultiplexed resulting in 240 libraries for the 16S-rRNA-V4 and ITS2, each.
 465 Overall data quality was analyzed using FastQC³⁵ under the Nephele server³⁶. An in-house
 466 command-line pipeline was designed to process the raw reads
 467 (github.com/vicflonun/AgaviSyncoms, Table S11). First, low-quality bases and reads were
 468 trimmed using Trimmomatic³⁷ and the resulting read pairs were assembled using FLASH³⁸.
 469 The VSEARCH³⁹ pipeline was used to process the resulting 21,735,602 and 12,391,310
 470 merged pairs for 16S-rRNA-V4 and ITS2, respectively. Before clustering, all the merged
 471 pairs were renamed, the primers and remaining adapters were cropped from the 5' and 3'
 472 endings, and the sequences were filtered based on their size and quality. The surviving
 473 merged pairs were de-replicated and the resulting unique sequences were clustered in
 474 operational taxonomic units OTUs. Singletons and chimeras were removed resulting in
 475 33,220 and 13,323 for the 16S-rRNA-V4 and ITS2, respectively. The renamed merged pairs
 476 were mapped against the resulting OTUs to generate an OTU table. The OTUs were
 477 annotated using the syntax algorithm with a bootstrap threshold of 0.80 using the RDP
 478 database for 16S-rRNA-V4 and the Unite database for ITS2.

479 Low abundant OTUs were removed based on the thresholds previously determined³. OTUs
 480 present with a high read count in the negative controls, and OTUs that were classified as

481 non-microbial (e. g. plant, chloroplast, mitochondria) were removed. Samples with less than
 482 1000 reads were removed. Finally, we obtained 236 samples with 1,164 OTUs for the 16S-
 483 rRNA-V4 amplicon and 199 samples with 1,416 OTUs for the ITS2 amplicon. Finally, the
 484 16S-rRNA and the ITS2 of the inoculated strains were mapped to the generated OTUs using
 485 VSEARCH(--usearch_global) at 97 and 95% of similarity, respectively (Table S6).

486 The generated raw reads have been deposited under the SRA accessions SRR18063421 to
 487 SRR18063568 (Bioproject PRJNA808160).

488 **Statistical analyses**

489 The phyllospheric microbial communities were analyzed in terms of alpha and beta diversity,
 490 OTU abundance, and network structure. All the data analyses and plots were made using R
 491 software (version 4.1.1)⁴⁰. Plots were made using the package ggplot2⁴¹ and pheatmap⁴². *In-*
 492 *house* R scripts and data used for each step have been deposited at
 493 github.com/vicflonun/AgaviSyncoms.

494 First, the OTU table was rarefied to the minimum read count (794 and 3022 for 16S-rRNA
 495 and ITS2, respectively) and the OTU richness and Shannon indexes were calculated. Beta
 496 diversity was estimated using the binary Bray-Curtis distance between samples, then the
 497 distance matrix was used to ordinate the samples using a non-metric multidimensional
 498 scaling and the distance to their centroid was calculated. PERMANOVA of binary Bray-
 499 Curtis distances was used to identify the factors that influence the microbial community
 500 structure. All diversity indexes (OTU and carbon profiling), distances, and ordinations were
 501 calculated with the package vegan⁴³ and MASS⁴⁴ and their differences between groups of
 502 samples (season, treatments, and block) were tested using a Kruskal-Wallis test with a post
 503 hoc Dunn test.

504 Correlation networks and the differences in the relative abundance of OTUs and taxa
 505 between groups of samples were calculated and analyzed as reported in (V. M. F.-N, Citlali
 506 Fonseca-García, Damaris Desgarennnes and L. P. P.-M. *in press*).

507 To test the influence of syncom treatment on plant growth, the changes per month of height,
 508 diameter, and unfolded leaves were calculated between February 2020 and November 2020
 509 (n=6), and November 2020 and June 2021 (n=30). A t-test was performed to determine if the
 510 growth rate of treated plants, the number of rhizomatic suckers, and the amount of °Brix
 511 were significantly different from non-inoculated plants.

512 **Metabolic profiling of the phyllospheric communities**

513 The metabolic profiling of the phyllospheric microbial communities was performed using Eco-
514 Biolog plates. This assay tests the ability of a microbial community to consume an array of
515 31 carbon sources. Two leaves per individual from each square in June 2021 (t19) were
516 sampled and the phyllospheric microorganisms were obtained as explained before. The
517 microbial pellet was resuspended in 5 ml of sterile phosphate buffer and the OD was
518 adjusted to 0.5. The microbial suspension was inoculated in the plates using 100 µl per well.
519 The plates were incubated in the darkness in a plastic box with wet paper towels to maintain
520 humidity at 28°C for 4 days. The OD in the plates was recorded using a plate reader at 590
521 nm with a circular agitation of 20 seconds.

522 **Leaf metabolomics**

523 **Metabolite extraction**

524 Leaves of *A. tequila* plants were collected and disinfected at t12. Cuttings from the base of
525 two leaves were taken (3 cm approx.) and freeze-dried. The lyophilized samples were
526 ground in a mixer mill (Retsch MM400) for 35 seg at 30 m/seg. Five mg of lyophilized tissue
527 were extracted with 500 µL of a mixture of methanol:acetonitrile:ethyl acetate at a 1:1:1 ratio
528 under sonication for 30 min at room temperature. Samples were then centrifuged at 14,000
529 rpm for 10 min at 4 °C. The supernatant was recovered, transferred to a 1.5 mL Eppendorf
530 tube, and the tube placed into a SpeedVac system to evaporate the solvent at room
531 temperature. The dried extracts were weighed, resuspended at a concentration of 300 ng/µL
532 (solution of water:acetonitrile at a 80:20 ratio), centrifuged at 14,000 rpm for 10 min at 4°C,
533 and the particle-free supernatant was recovered for further analysis. Equal volumes of all
534 particle-free supernatants were mixed in one tube to prepare quality control (QC) samples.

535 **Untargeted LC-MS² data acquisition**

536 The metabolites were separated (2 µL sample injection) using an Agilent nanoLC 1260
537 Infinity system (Agilent Technologies, Inc., Santa Clara, CA) and a column packed with
538 ZORBAX 80 SB-C18 (75 µm x 43 mm, 5 µm C18 with a 40 nL enrichment or trap column).
539 Water and acetonitrile with 0.1% formic acid were used as mobile phases A and B at a flow
540 rate of 300 nL/min and 2 µL for the nano (to separation column) and micro pumps (to
541 enrichment column), respectively. The gradient started at 5% B, increased linearly to 20% B
542 in 20 min, maintained at 20% B for 5 min, increased linearly to 100%, maintained at 100% B
543 for 5 min, returned to 5% B in 1 min and then maintained at 5% B for 9 min before the next
544 sample (to ensure column re-equilibration). Two blank samples (3 µL of mobile phases A
545 and B at a 95:5 ratio) were run between experimental sample injections to minimize potential

carryover. The eluate from the column was delivered to a 6530 Accurate-Mass Q-TOF mass spectrometer (Agilent Technologies, Inc., Santa Clara, CA) via a HPLC-Chip Cube MS interface. Nanospray ionization under positive mode was employed. MS data was acquired using the following conditions: capillary voltage, 1850 V; gas temperature, 350 °C; drying gas flow, 5 L/min; skimmer voltage, 65 V; octapole RF, 750 V; fragmentor voltage, 175 V; spectra acquisition rate, 4 spectra/s over a mass range of 110 to 2000 m/z. MS² data conditions were: isolation window, narrow (1.3 m/z); spectra acquisition rate, 3 spectra/s; max precursors per cycle, 5 over a mass range of 50-2000 m/z. The active exclusion option was enabled, set to 2 spectra and release after 0.25 min. The ramped collision energy (CE) option was used with slope and offset values of 6 and 4, respectively. The instrument was externally calibrated before sample acquisition using ESI-L low mix concentration tuning mix solution (Agilent Technologies, Inc., Santa Clara, CA) to ensure a mass accuracy of <5 ppm for both MS and MS² data. Instrument performance was monitored during data acquisition by including QC samples every 3 experimental samples and evaluating the blank samples signals. Samples were randomly allocated for data acquisition.

LC-MS² data processing and analysis

We analyzed the LC-MS² datasets using a workflow comprised of open-access software packages and online platforms and following three main steps: 1) features were extracted and aligned among datasets using MZmine version 2.53⁴⁵; 2) univariate and multivariate statistical analyses and hierarchical clustering visualization was done using Metaboanalyst 4.0 (<https://www.metaboanalyst.ca>)⁴⁶; 3) automatic metabolite annotation or identification at the structure level⁴⁷ was performed using the Global Natural Products Social Molecular Networking web platform (GNPS, <https://gnps.ucsd.edu>) (Metabolomics Standards Initiative [MSI] classification level 2)⁴⁸ and *in-silico* tools (MolDiscovery, CSI:FingerID) (MSI, level 3)^{49,50}. The chemical classes of the modulated metabolites (FC >1.5/1 or <1/1.5, *p-value* <0.05) (MSI, level 3) were retrieved by the CANOPUS tool^{51,52} integrated within SIRIUS software version 4.9.12. The detailed processing parameters for all the pipelines are found in supplemental experimental methods in supporting information.

Data availability

The raw datasets have been deposited on the GNPS/MassIVE public repository⁵³ under the accession number MSV000089335. The parameters for classical molecular networking and spectral matching using all datasets are available on the following link: <https://gnps.ucsd.edu/ProteoSAFe/status.jsp?task=fef5e3d881d049dd9a05cfa05ad8ac8e>.

579 The MolDiscovery results are available on the following link:
580 <https://gnps.ucsd.edu/ProteoSAFe/status.jsp?task=186cab767a064f5b871914c6f7545aec>.

581 **Metabolomics statistical analysis**

582 For the metabolomics data, features with reproducible quantification (based on QC
583 coefficient of variation <30%) and fold change ≥ 1.5 or $\leq 1/1.5$ and a p-value <0.05 (t-test)
584 were considered as differentially abundant when comparing two groups. The abundance of
585 selected metabolites was plotted as Box Plots depicting median and 10-90 percentile. For
586 multivariate statistical analysis and heatmap visualization, Metaboanalyst 4.0
587 (<https://www.metaboanalyst.ca>) was utilized. Principal component analysis (PCA) was used
588 to assess for sample clustering behavior and inter-group variation. Data was Log
589 transformed (without scaling) for PCA and heatmap analysis.

590

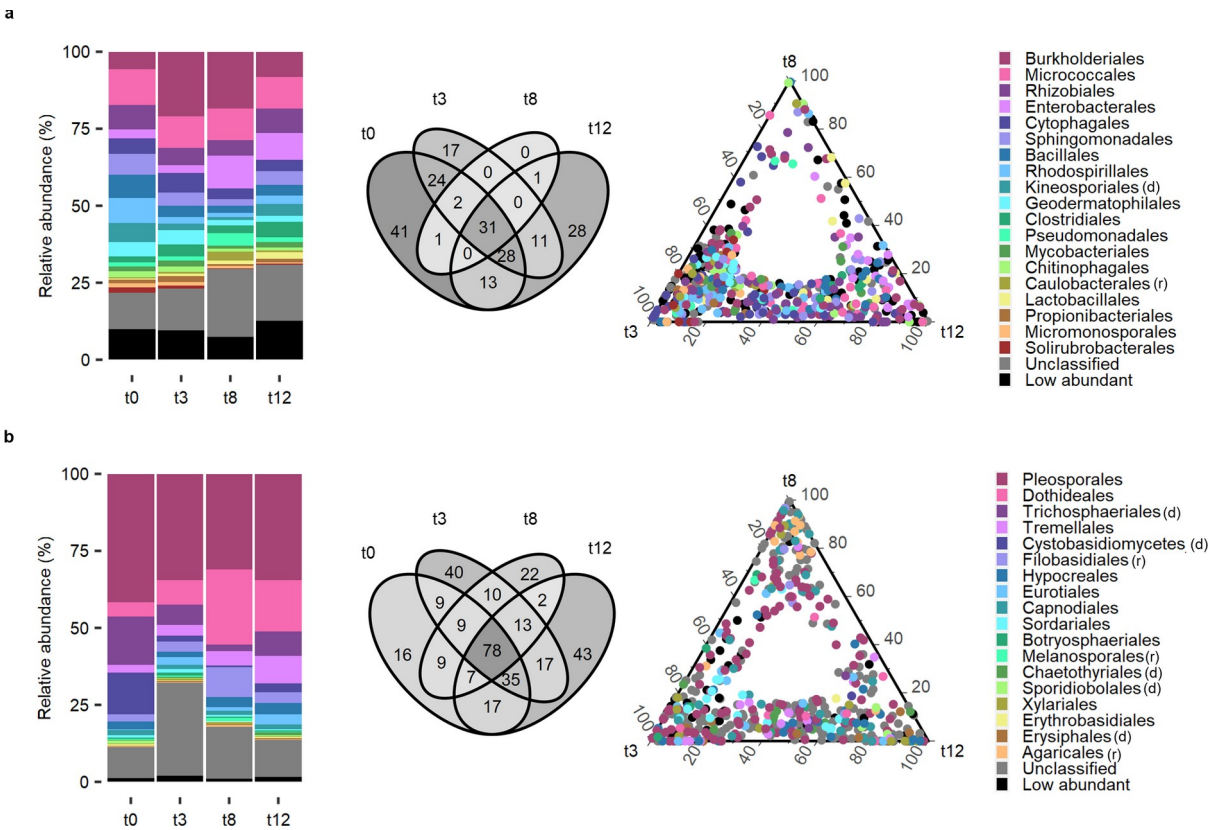
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602 151007.

603 **Author contributions**

604 VMFN and LPPM planned and designed research; VMFN designed syncoms and conducted
605 most research and analyses; JACP, JDChG, RAV, MNVS helped perform experiments and
606 the sampling of plants and microbial communities; AGHM, JXT, AMU analyzed plant
607 metabolic data; VMFN and LPPM wrote paper; all authors revised and approved the final
608 manuscript; and LPPM secured funding.

609 MAIN FIGURES



611 **Fig. 1. Taxonomic classification, core, and enrichment analysis of prokaryotic (a) and**
612 **eukaryotic (b) phyllospheric communities across sampling seasons.** Bars represent the
613 order-level relative abundance of OTUs. Venn diagrams show the number of shared OTUs
614 (50% prevalence). Ternary plots show the ratio of the mean relative abundance of differential
615 OTUs (Kruskal-Wallis test/Dunn test: $FC \geq |1.5|$ and adjusted $p\text{-value} \leq 0.05$). Order-level
616 enrichments are indicated with letters. d: enriched in dry season. r: enriched in rainy season
617 (Wilcox test: $FC \geq |1.5|$ and adjusted $p\text{-value} \leq 0.05$). Sampling seasons. t0:November
618 2019. t3:February 2020. t8:July 2020. t12:November 2020.

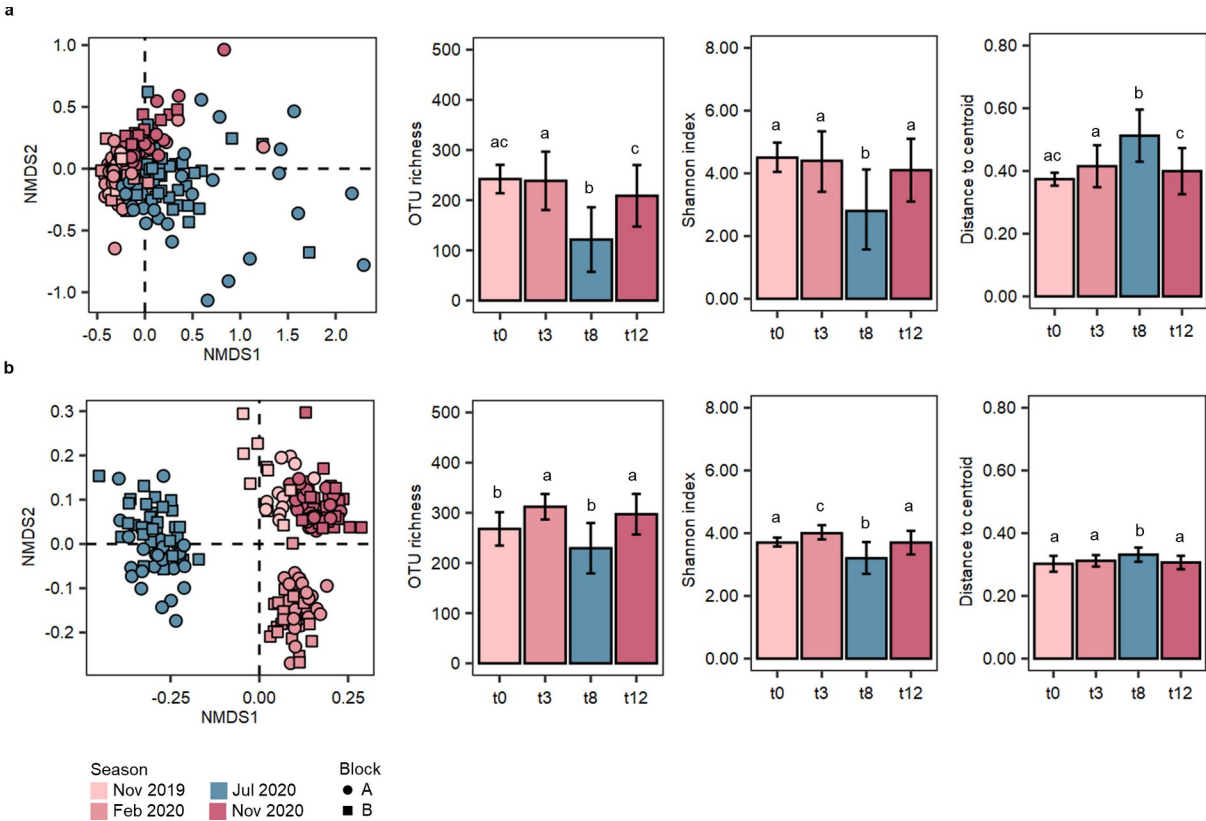


Fig. 2. Diversity of the prokaryotic (a) and eukaryotic (b) phyllospheric communities across sampling season. Non-metric multidimensional scaling (NMDS) was calculated using binary Bray-Curtis dissimilarities between samples. Bars represent the mean $\pm$ sd of the observed OTU richness, Shannon index, and distance to the centroid of the dissimilarities. Letters indicate significant changes between sampling seasons (Kruskal-Wallis test/Dunn test: adjusted p-value $\leq$ 0.05). Sampling seasons: t0:November 2019. t3:February 2020. t8:July 2020. t12:November 2020.

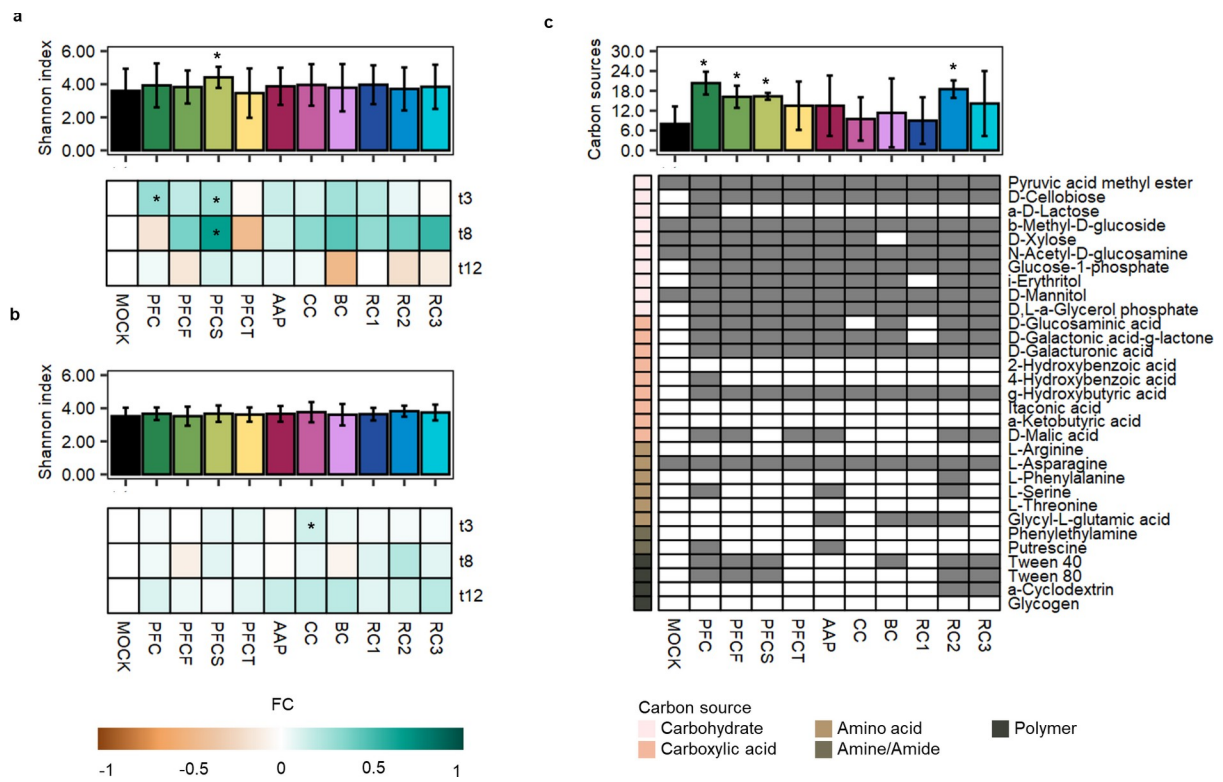


Fig. 3. Taxonomic diversity (a,b) and carbon consumption profile (c) of phyllospheric communities by syncom treatment. Bars represent the mean $\pm$ sd of the Shannon index of prokaryotic (a) and eukaryotic (b) phyllospheric communities by syncom treatment and the heatmaps below show the fold change of the Shannon index between the syncom treatments and the control (MOCK) in each sampling season (t3, t8, t12). The carbon consumption profile (c) is represented as the mean $\pm$ sd of the number of carbon sources consumed by the phyllospheric communities in June 2021 (t19) and the heatmap below represents the unique carbon sources consumed by more than 50% of the samples in each treatment. In all cases, an asterisk indicates a significant change between treatment and control (Wilcox-test: p -value ≤ 0.05). Sampling seasons: t3:February 2020. t8:July 2020. t12:November 2020.

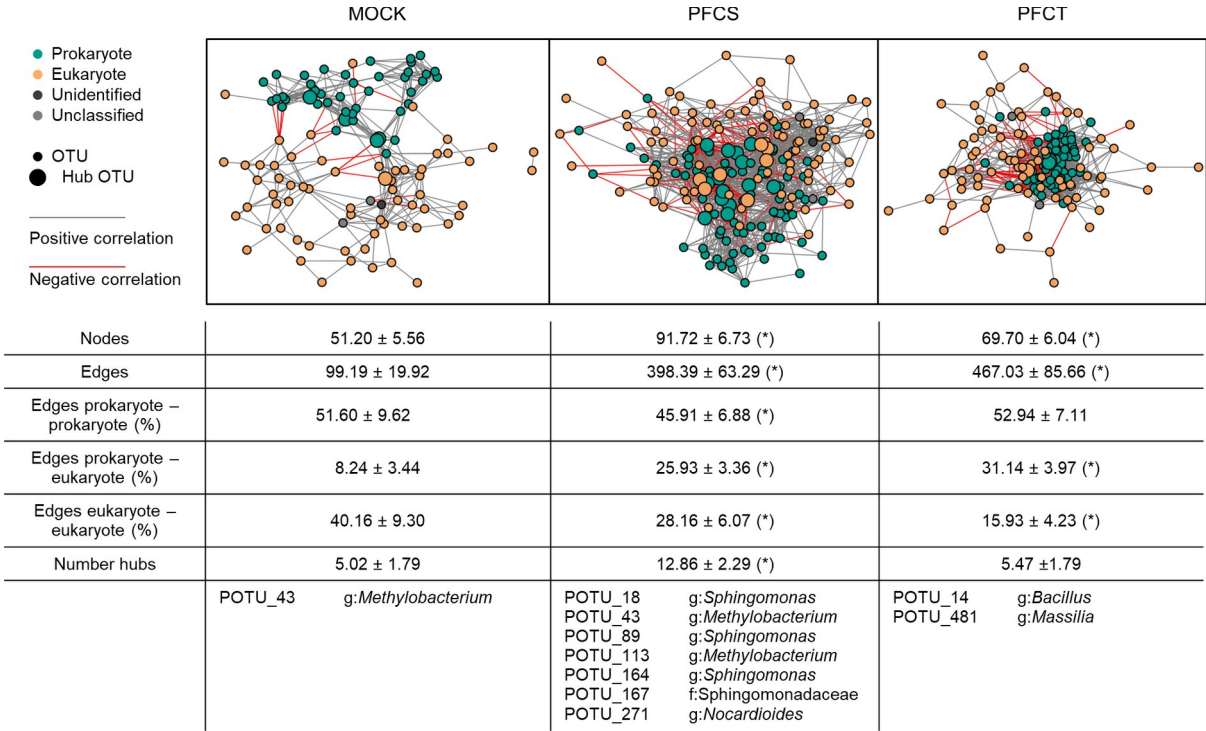


Fig. 4. Network analysis of the phyllospheric microbial communities by PFCS and PFCT syncom treatment. Correlation networks of the prokaryotic and fungal OTU counts in the non-inoculated plants (MOCK) and plants inoculated with the PFCS and PFCT syncoms. Vertices represent OTUs and edges represent significant correlation (Spearman cor. $p \geq |0.6|$) and p -value ≤ 0.01). Vertices are colored based on the OTU kingdom and edges based on their weight. Red was used for negative correlations and black for positive correlations. Bigger OTUs represent microbial hubs (above). Vertex and edge metrics (mean $\pm$ sd) of 100 random correlation networks. Only hubs that matched the same genera or OTU id of inoculated strains are shown (below). An asterisk indicates a significant change between treatment and control (Wilcoxon-test: p -value ≤ 0.05)

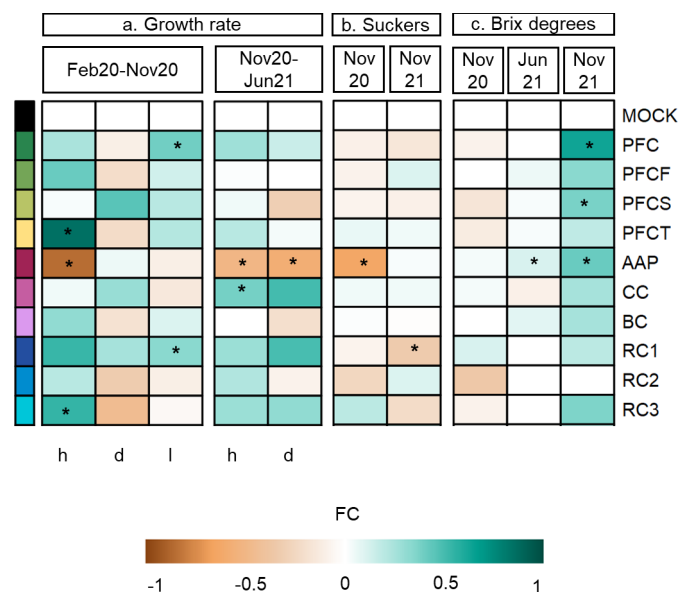
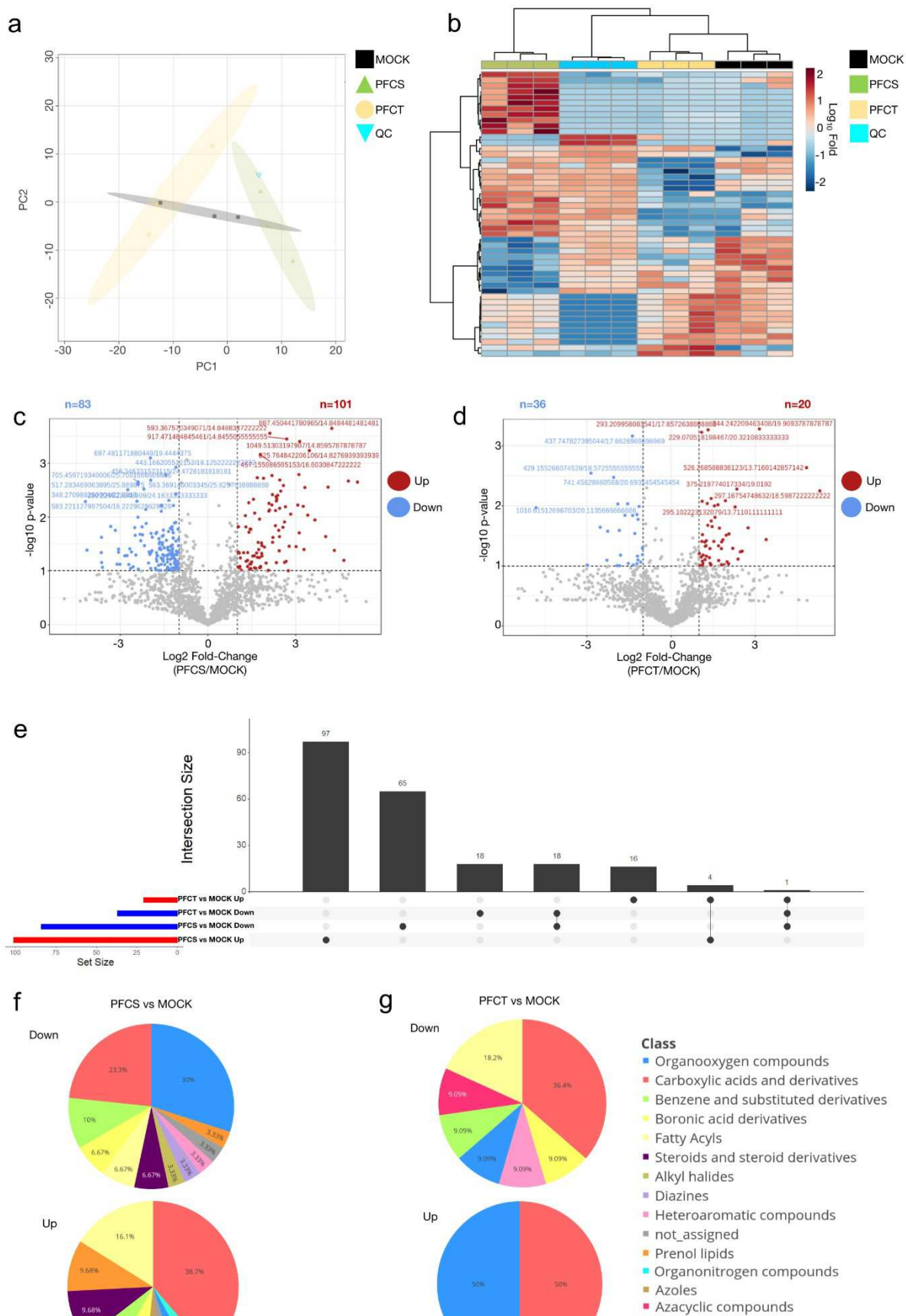


Fig. 5. Growth and developmental changes in *A. tequilana* plants inoculated with syncoms. Heatmap represents the fold changes between the syncom treatments and the control (MOCK) for the plant parameters measured. The changes in height (h), diameter (d), and the number of leaves per month (l) were recorded from February 2020 (t3) to November 2020 (t12, n=6), while only height (h) and diameter (d) from November 2020 (t12) to June 2021 (t19, n=30, a). Rhizomatic suckers (b) were counted in November 2020 (t12) and 2021 (t24) while the °Brix (c) in the stem sap were measured in November 2020 (t12), June 2021 (t19), and November 2021 (t24). Asterix indicates a significant change between treatment and control (Wilcox-test: p-value ≤ 0.05).



666 **Fig. 6. Mass spectrometry-based untargeted metabolomic and chemoinformatic**
667 **analyses of *A. tequilana* plants inoculated with PFCS and PFCT syncoms.** (a) Principal
668 Component Analysis (PCA) and (b) heatmap visualization (of the top 50 metabolites ranked
669 by ANOVA test) of the aligned features among treatments with reproducible quantification.
670 Data was Log-transformed, but not scaled. Shade areas depict the 95% confidence
671 intervals. Quantified features among (c) PFCS and MOCK and (d) PFCT and MOCK
672 treatments. Differential abundant features were determined when reaching a fold change $\geq$
673 1.5 or $\leq 1/1.5$ and p-value < 0.05 (t-test). (e) Overlapping pattern among treatments using the
674 features modulated (based on fold change and t-test criteria) by syncoms. Chemical class
675 (retrieved by CANOPUS tool) profile assigned to the down- and up-modulated features by (f)
676 PFCS and (g) PFCT syncoms.

677

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