**Occurrence of *Listeria monocytogenes* is associated with built environment microbiota in three tree fruit processing facilities**

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**Supplementary Materials**

**Table S1:** Metadata for collected samples

|  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- |
| Bacterial community samplesa | Fungal community sample | Facility | Sample collection section | Sample collection date | Week | Month | *L. monocytogenes* detection result |
| 0105-1s | 0105-1t | F1 | wash | 5-Jan-18 | W5 | January | + |
| 0105-2s | 0105-2t | F1 | dry | 5-Jan-18 | W5 | January | - |
| 0105-3s | 0105-3t | F1 | wax | 5-Jan-18 | W5 | January | - |
| 0105-4s | 0105-4t | F2 | wash | 5-Jan-18 | W5 | January | + |
| 0105-5s | 0105-5t | F2 | dry | 5-Jan-18 | W5 | January | + |
| 0105-6s | 0105-6t | F2 | wax | 5-Jan-18 | W5 | January | + |
| 0105-7s | 0105-7t | F3 | wash | 5-Jan-18 | W5 | January | + |
| 0105-8s | 0105-8t | F3 | dry | 5-Jan-18 | W5 | January | - |
| 0105-9s | 0105-9t | F3 | wax | 5-Jan-18 | W5 | January | + |
| 0119-1s | 0119-1t | F1 | wash | 19-Jan-18 | W6 | January | - |
| 0119-2s | 0119-2t | F1 | dry | 19-Jan-18 | W6 | January | - |
| 0119-3s | 0119-3t | F1 | wax | 19-Jan-18 | W6 | January | + |
| 0119-4s | 0119-4t | F2 | wash | 19-Jan-18 | W6 | January | + |
| 0119-5s | 0119-5t | F2 | dry | 19-Jan-18 | W6 | January | + |
| 0119-6s | 0119-6t | F2 | wax | 19-Jan-18 | W6 | January | + |
| 0119-7s | 0119-7t | F3 | wash | 19-Jan-18 | W6 | January | + |
| 0119-8s | 0119-8t | F3 | dry | 19-Jan-18 | W6 | January | - |
| 0119-9s | 0119-9t | F3 | wax | 19-Jan-18 | W6 | January | - |
| 0202-1s | 0202-1t | F1 | wash | 2-Feb-18 | W7 | February | + |
| 0202-2s | 0202-2t | F1 | dry | 2-Feb-18 | W7 | February | + |
| 0202-3s | 0202-3t | F1 | wax | 2-Feb-18 | W7 | February | - |
| 0202-4s | 0202-4t | F2 | wash | 2-Feb-18 | W7 | February | + |
| 0202-5s | 0202-5t | F2 | dry | 2-Feb-18 | W7 | February | + |
| 0202-6s | 0202-6t | F2 | wax | 2-Feb-18 | W7 | February | + |
| 0202-7s | 0202-7t | F3 | wash | 2-Feb-18 | W7 | February | + |
| 0202-8s | 0202-8t | F3 | dry | 2-Feb-18 | W7 | February | - |
| 0202-9s | 0202-9t | F3 | wax | 2-Feb-18 | W7 | February | - |
| 0216-1s | 0216-1t | F1 | wash | 16-Feb-18 | W8 | February | + |
| 0216-2s | 0216-2t | F1 | dry | 16-Feb-18 | W8 | February | - |
| 0216-3s | 0216-3t | F1 | wax | 16-Feb-18 | W8 | February | - |
| 0216-4s | 0216-4t | F2 | wash | 16-Feb-18 | W8 | February | + |
| 0216-5s | 0216-5t | F2 | dry | 16-Feb-18 | W8 | February | + |
| 0216-6s | 0216-6t | F2 | wax | 16-Feb-18 | W8 | February | + |
| 0216-7s | 0216-7t | F3 | wash | 16-Feb-18 | W8 | February | - |
| 0216-8s | 0216-8t | F3 | dry | 16-Feb-18 | W8 | February | - |
| 0216-9s | 0216-9t | F3 | wax | 16-Feb-18 | W8 | February | - |
| 0302-1s | 0302-1t | F1 | wash | 2-Mar-18 | W9 | March | - |
| 0302-2s | 0302-2t | F1 | dry | 2-Mar-18 | W9 | March | - |
| 0302-3s | 0302-3t | F1 | wax | 2-Mar-18 | W9 | March | - |
| 0302-4s | 0302-4t | F2 | wash | 2-Mar-18 | W9 | March | + |
| 0302-5s | 0302-5t | F2 | dry | 2-Mar-18 | W9 | March | + |
| 0302-6s | 0302-6t | F2 | wax | 2-Mar-18 | W9 | March | + |
| 0302-7s | 0302-7t | F3 | wash | 2-Mar-18 | W9 | March | + |
| 0302-8s | 0302-8t | F3 | dry | 2-Mar-18 | W9 | March | - |
| 0302-9s | 0302-9t | F3 | wax | 2-Mar-18 | W9 | March | + |
| 0316-1s | 0316-1t | F1 | wash | 16-Mar-18 | W10 | March | - |
| 0316-2s | 0316-2t | F1 | dry | 16-Mar-18 | W10 | March | - |
| 0316-3s | 0316-3t | F1 | wax | 16-Mar-18 | W10 | March | - |
| 0316-4s | 0316-4t | F2 | wash | 16-Mar-18 | W10 | March | + |
| 0316-5s | 0316-5t | F2 | dry | 16-Mar-18 | W10 | March | + |
| 0316-6s | 0316-6t | F2 | wax | 16-Mar-18 | W10 | March | + |
| 0316-7s | 0316-7t | F3 | wash | 16-Mar-18 | W10 | March | + |
| 0316-8s | 0316-8t | F3 | dry | 16-Mar-18 | W10 | March | + |
| 0316-9s | 0316-9t | F3 | wax | 16-Mar-18 | W10 | March | - |
| 0404-1s | 0404-1t | F1 | wash | 4-Apr-18 | W11 | April | - |
| 0404-2s | 0404-2t | F1 | dry | 4-Apr-18 | W11 | April | - |
| 0404-3s | 0404-3t | F1 | wax | 4-Apr-18 | W11 | April | - |
| 0404-4s | 0404-4t | F2 | wash | 4-Apr-18 | W11 | April | + |
| 0404-5s | 0404-5t | F2 | dry | 4-Apr-18 | W11 | April | + |
| 0404-6s | 0404-6t | F2 | wax | 4-Apr-18 | W11 | April | + |
| 0404-7s | 0404-7t | F3 | wash | 4-Apr-18 | W11 | April | - |
| 0404-8s | 0404-8t | F3 | dry | 4-Apr-18 | W11 | April | - |
| 0404-9s | 0404-9t | F3 | wax | 4-Apr-18 | W11 | April | - |
| 0416-1s | 0416-1t | F1 | wash | 16-Apr-18 | W12 | April | - |
| 0416-2s | 0416-2t | F1 | dry | 16-Apr-18 | W12 | April | - |
| 0416-3s | 0416-3t | F1 | wax | 16-Apr-18 | W12 | April | - |
| 0416-4s | 0416-4t | F2 | wash | 16-Apr-18 | W12 | April | + |
| 0416-5s | 0416-5t | F2 | dry | 16-Apr-18 | W12 | April | + |
| 0416-6s | 0416-6t | F2 | wax | 16-Apr-18 | W12 | April | + |
| 0416-7s | 0416-7t | F3 | wash | 16-Apr-18 | W12 | April | - |
| 0416-8s | 0416-8t | F3 | dry | 16-Apr-18 | W12 | April | - |
| 0416-9s | 0416-9t | F3 | wax | 16-Apr-18 | W12 | April | - |
| 0427-1s | 0427-1t | F1 | wash | 27-Apr-18 | W13 | April | + |
| 0427-2s | 0427-2t | F1 | dry | 27-Apr-18 | W13 | April | + |
| 0427-3s | 0427-3t | F1 | wax | 27-Apr-18 | W13 | April | - |
| 0427-4s | 0427-4t | F2 | wash | 27-Apr-18 | W13 | April | + |
| 0427-5s | 0427-5t | F2 | dry | 27-Apr-18 | W13 | April | + |
| 0427-6s | 0427-6t | F2 | wax | 27-Apr-18 | W13 | April | + |
| 0427-7s | 0427-7t | F3 | wash | 27-Apr-18 | W13 | April | - |
| 0427-8s | 0427-8t | F3 | dry | 27-Apr-18 | W13 | April | - |
| 0427-9s | 0427-9t | F3 | wax | 27-Apr-18 | W13 | April | - |
| 1103-1s | 1103-1t | F1 | wash | 3-Nov-17 | W1 | November | + |
| 1103-2s | 1103-2t | F1 | dry | 3-Nov-17 | W1 | November | - |
| 1103-3s | 1103-3t | F1 | wax | 3-Nov-17 | W1 | November | - |
| 1103-4s | 1103-4t | F2 | wash | 3-Nov-17 | W1 | November | + |
| 1103-5s | 1103-5t | F2 | dry | 3-Nov-17 | W1 | November | + |
| 1103-6s | 1103-6t | F2 | wax | 3-Nov-17 | W1 | November | + |
| 1103-7s | 1103-7t | F3 | wash | 3-Nov-17 | W1 | November | - |
| 1103-8s | 1103-8t | F3 | dry | 3-Nov-17 | W1 | November | + |
| 1103-9s | 1103-9t | F3 | wax | 3-Nov-17 | W1 | November | - |
| 1121-1s | 1121-1t | F1 | wash | 21-Nov-17 | W2 | November | + |
| 1121-2s | 1121-2t | F1 | dry | 21-Nov-17 | W2 | November | - |
| 1121-3s | 1121-3t | F1 | wax | 21-Nov-17 | W2 | November | - |
| 1121-4s | 1121-4t | F2 | wash | 21-Nov-17 | W2 | November | + |
| 1121-5s | 1121-5t | F2 | dry | 21-Nov-17 | W2 | November | + |
| 1121-6s | 1121-6t | F2 | wax | 21-Nov-17 | W2 | November | + |
| 1121-7s | 1121-7t | F3 | wash | 21-Nov-17 | W2 | November | + |
| 1121-8s | 1121-8t | F3 | dry | 21-Nov-17 | W2 | November | + |
| 1121-9s | 1121-9t | F3 | wax | 21-Nov-17 | W2 | November | + |
| 1208-1s | 1208-1t | F1 | wash | 8-Dec-17 | W3 | December | + |
| 1208-2s | 1208-2t | F1 | dry | 8-Dec-17 | W3 | December | - |
| 1208-3s | 1208-3t | F1 | wax | 8-Dec-17 | W3 | December | - |
| 1208-4s | 1208-4t | F2 | wash | 8-Dec-17 | W3 | December | + |
| 1208-5s | 1208-5t | F2 | dry | 8-Dec-17 | W3 | December | + |
| 1208-6s | 1208-6t | F2 | wax | 8-Dec-17 | W3 | December | + |
| 1208-7s | 1208-7t | F3 | wash | 8-Dec-17 | W3 | December | + |
| 1208-8s | 1208-8t | F3 | dry | 8-Dec-17 | W3 | December | + |
| 1208-9s | 1208-9t | F3 | wax | 8-Dec-17 | W3 | December | + |
| 1218-1s | 1218-1t | F1 | wash | 18-Dec-17 | W4 | December | - |
| 1218-2s | 1218-2t | F1 | dry | 18-Dec-17 | W4 | December | + |
| 1218-3s | 1218-3t | F1 | wax | 18-Dec-17 | W4 | December | - |
| 1218-4s | 1218-4t | F2 | wash | 18-Dec-17 | W4 | December | + |
| 1218-5s | 1218-5t | F2 | dry | 18-Dec-17 | W4 | December | + |
| 1218-6s | 1218-6t | F2 | wax | 18-Dec-17 | W4 | December | + |
| 1218-7s | 1218-7t | F3 | wash | 18-Dec-17 | W4 | December | - |
| 1218-8s | 1218-8t | F3 | dry | 18-Dec-17 | W4 | December | + |
| 1218-9s | 1218-9t | F3 | wax | 18-Dec-17 | W4 | December | - |

aAmplicon sequences corresponding to listed samples are available on NCBI under BioProject accession number PRJNA527988.

**Table S2:** Chi-square test of *L. monocytogenes* occurrence among processing sections

|  |  |  |  |
| --- | --- | --- | --- |
| *L. monocytogenes* occurrence | Sectiona | | |
|  | Dry | Wash | Wax |
| Absent | 18 | 12 | 21 |
| Present | 21 | 27 | 18 |
|  | Chi-Square | DF | P-value |
| Pearson | 4.38 | 2 | 0.112 |
| Likelihood | 4.454 | 2 | 0.108 |

aDF, degree of freedom.

In Facility F1 there was a significant difference among occurrence of *L. monocytogenes* in different sampled sections (P = 0.029). No significant difference in *L. monocytogenes* occurrence was observed among samples collected from different sections in Facility F3 (P = 0.476). Statistical analysis was not conducted for samples collected in facility F2 given that all samples from Facility F2 were positive for *L. monocytogenes*.

**Table S3:** Chi-square test of *L. monocytogenes* occurrence among facilities

|  |  |  |  |
| --- | --- | --- | --- |
| *L. monocytogenes* occurrence | Facilitya | | |
|  | F1 | F2 | F3 |
| Absent | 28 | 0 | 23 |
| Present | 11 | 39 | 16 |
|  | Chi-Square | DF | P-value |
| Pearson | 46.51 | 2 | < 0.001 |
| Likelihood | 61.07 | 2 | < 0.001 |

aDF, degree of freedom.

**Table S4:** Results of pairwise PERMANOVA analyses for microbial communities

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Factor | DFa | Sums of squares | F model | R2b | P value | P adjusted |
| January vs February | 1 | 0.431 | 1.052 | 0.030 | 0.347 | 1.000 |
| January vs March | 1 | 0.524 | 1.307 | 0.037 | 0.083 | 1.000 |
| January vs April | 1 | 0.559 | 1.397 | 0.031 | 0.040 | 0.600 |
| January vs November | 1 | 0.534 | 1.296 | 0.037 | 0.044 | 0.660 |
| January vs December | 1 | 0.330 | 0.783 | 0.023 | 0.930 | 1.000 |
| February vs March | 1 | 0.310 | 0.785 | 0.023 | 0.861 | 1.000 |
| February vs April | 1 | 0.571 | 1.441 | 0.032 | 0.030 | 0.450 |
| February vs November | 1 | 0.549 | 1.349 | 0.038 | 0.050 | 0.750 |
| February vs December | 1 | 0.447 | 1.074 | 0.031 | 0.299 | 1.000 |
| March vs April | 1 | 0.549 | 1.411 | 0.032 | 0.055 | 0.825 |
| March vs November | 1 | 0.588 | 1.477 | 0.042 | 0.025 | 0.375 |
| March vs December | 1 | 0.553 | 1.359 | 0.038 | 0.043 | 0.645 |
| April vs November | 1 | 0.651 | 1.632 | 0.037 | 0.004 | 0.060 |
| April vs December | 1 | 0.675 | 1.665 | 0.037 | 0.005 | 0.075 |
| November vs December | 1 | 0.442 | 1.056 | 0.030 | 0.330 | 1.000 |
| F1 vs F2 | 1 | 2.788 | 7.664 | 0.092 | 0.001 | 0.003 |
| F1 vs F3 | 1 | 1.525 | 3.904 | 0.049 | 0.001 | 0.003 |
| F2 vs F3 | 1 | 3.259 | 8.983 | 0.106 | 0.001 | 0.003 |
| Wash vs dry | 1 | 0.713 | 1.780 | 0.023 | 0.003 | 0.009 |
| Wash vs wax | 1 | 1.346 | 3.337 | 0.042 | 0.001 | 0.003 |
| Dry vs wax | 1 | 0.716 | 1.801 | 0.023 | 0.003 | 0.009 |

aDF, degree of freedom.  
bR2, R square.

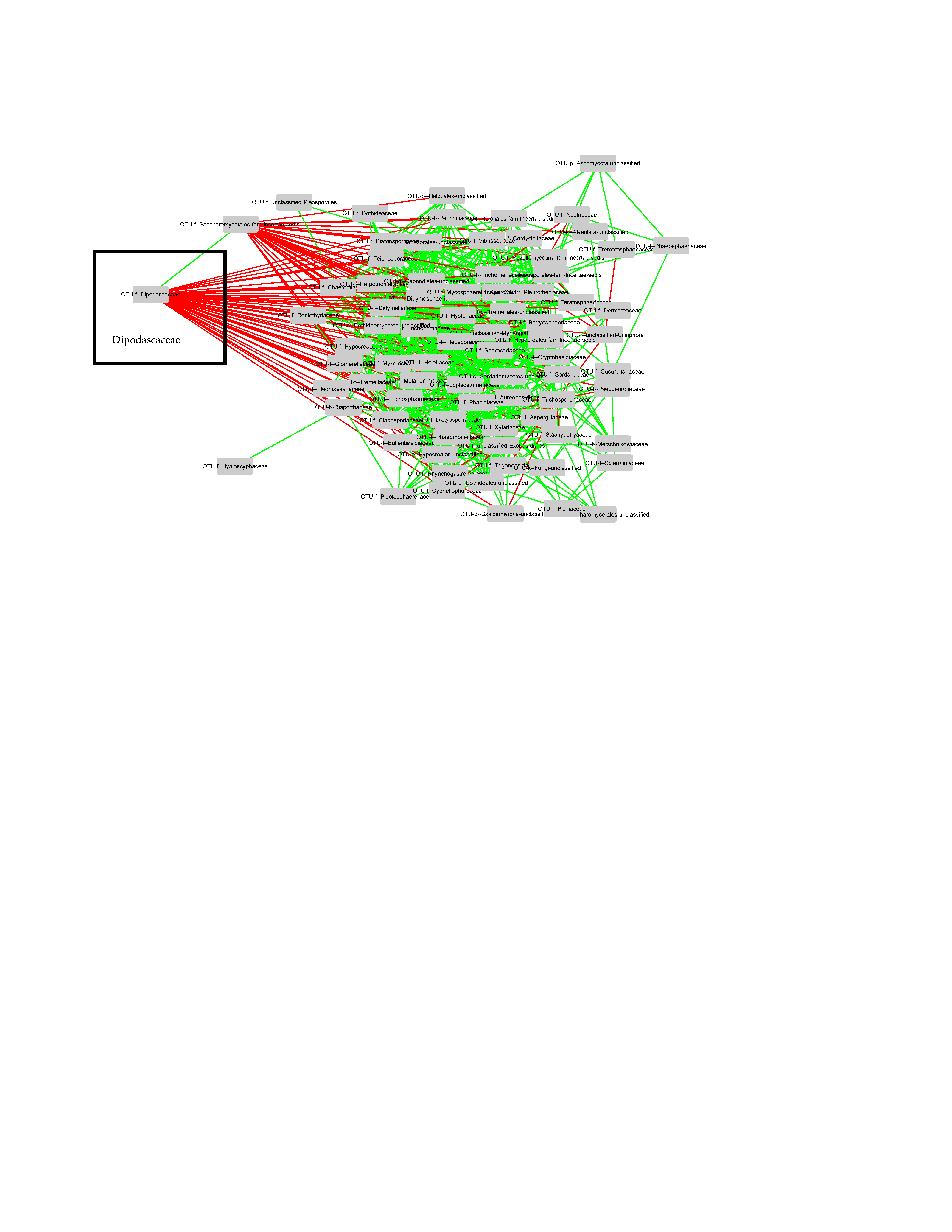
**Table S5:** Results of pairwise PERMANOVA analyses for fungal communities

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Factor | DFa | Sums of squares | F model | R2b | P value | P adjusted |
| January vs February | 1 | 0.152 | 0.688 | 0.020 | 0.588 | 1.000 |
| January vs March | 1 | 0.195 | 0.857 | 0.025 | 0.474 | 1.000 |
| January vs April | 1 | 0.436 | 1.714 | 0.038 | 0.138 | 1.000 |
| January vs November | 1 | 0.303 | 1.261 | 0.036 | 0.215 | 1.000 |
| January vs December | 1 | 0.163 | 0.734 | 0.021 | 0.514 | 1.000 |
| February vs March | 1 | 0.212 | 0.886 | 0.025 | 0.454 | 1.000 |
| February vs April | 1 | 0.234 | 0.892 | 0.020 | 0.454 | 1.000 |
| February vs November | 1 | 0.382 | 1.521 | 0.043 | 0.163 | 1.000 |
| February vs December | 1 | 0.165 | 0.712 | 0.020 | 0.570 | 1.000 |
| March vs April | 1 | 0.247 | 0.919 | 0.021 | 0.462 | 1.000 |
| March vs November | 1 | 0.532 | 2.053 | 0.057 | 0.078 | 1.000 |
| March vs December | 1 | 0.459 | 1.913 | 0.053 | 0.108 | 1.000 |
| April vs November | 1 | 0.522 | 1.871 | 0.042 | 0.100 | 1.000 |
| April vs December | 1 | 0.335 | 1.270 | 0.029 | 0.236 | 1.000 |
| November vs December | 1 | 0.172 | 0.680 | 0.020 | 0.599 | 1.000 |
| F1 vs F2 | 1 | 10.111 | 82.072 | 0.519 | 0.001 | 0.003 |
| F1 vs F3 | 1 | 3.247 | 23.382 | 0.235 | 0.001 | 0.003 |
| F2 vs F3 | 1 | 8.281 | 65.764 | 0.464 | 0.001 | 0.003 |
| Wash vs dry | 1 | 0.748 | 3.175 | 0.040 | 0.019 | 0.057 |
| Wash vs wax | 1 | 1.167 | 4.874 | 0.060 | 0.004 | 0.012 |
| Dry vs wax | 1 | 0.243 | 0.951 | 0.012 | 0.412 | 1.000 |

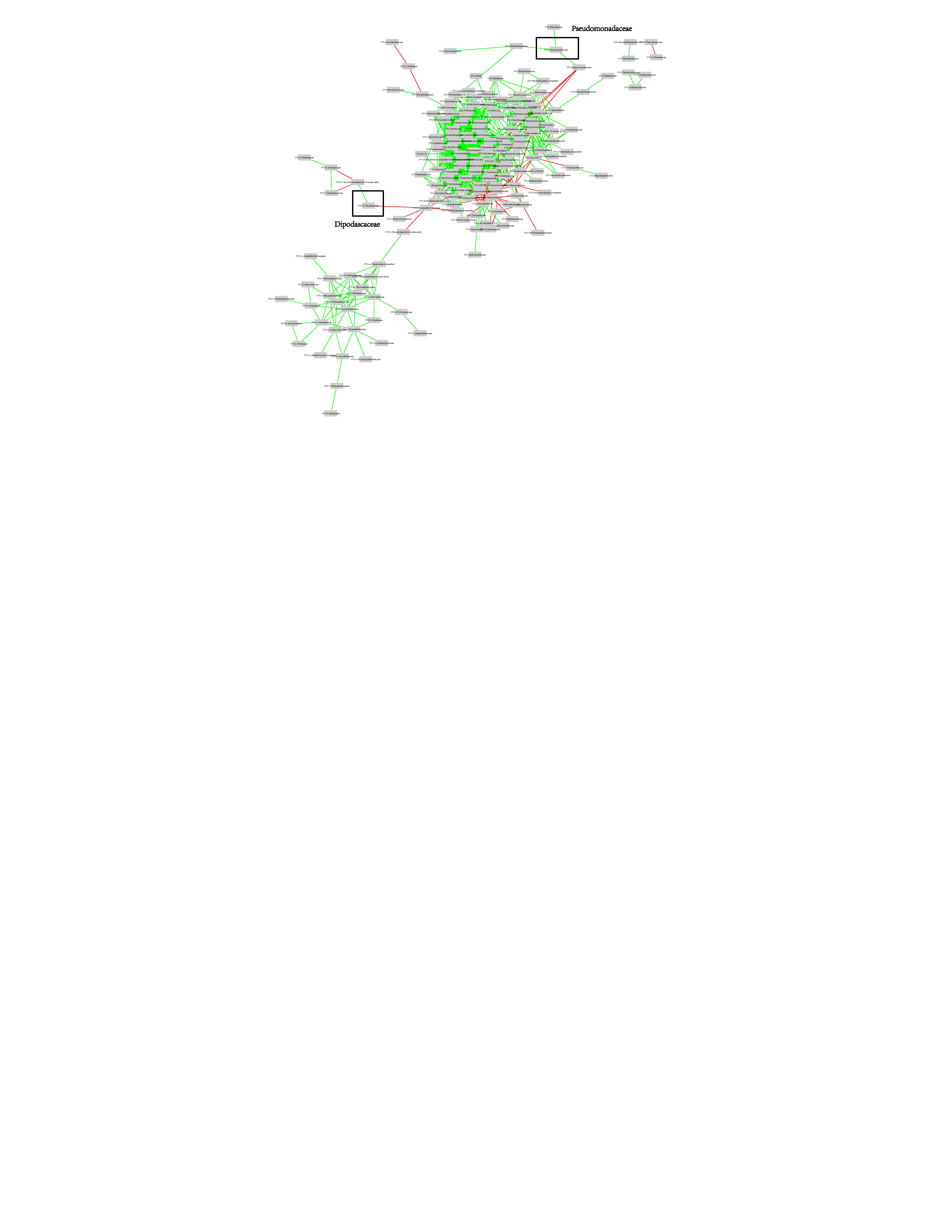
aDF, degree of freedom.  
bR2, R square.

**Table S6**: Comparison of relative abundances of bacterial families identified using Greengenes and SILVA database. Statistical t test indicated insignificant differences in relative abundances of bacterial families obtained using these two databases.

|  |  |  |
| --- | --- | --- |
|  | Relative abundance | |
| Bacterial family | Greengenes | SILVA |
| Pseudomonadaceae | 21.95321 | 21.79674 |
| Flavobacteriaceae | 19.30667 | 18.83549 |
| Weeksellaceae | 11.53586 | 10.65667 |
| Moraxellaceae | 8.59841 | 8.54560 |
| Burkholderiaceae | 7.12522 | 7.07451 |
| Sphingomonadaceae | 6.04874 | 6.01967 |
| Caulobacteraceae | 5.69444 | 5.64196 |
| Xanthomonadaceae | 5.61480 | 5.58256 |
| Sphingobacteriaceae | 4.02314 | 3.96145 |
| Enterobacteriaceae | 2.49579 | 2.48440 |
| Rhizobiaceae | 2.42454 | 2.39619 |
| Rhodobacteraceae | 2.25574 | 2.25949 |
| Chitinophagaceae | 2.14125 | 2.06358 |
| Arcobacteraceae | 1.38955 | 1.39003 |
| Azospirillaceae | 1.32850 | 1.33916 |
| Bdellovibrionaceae | 1.18188 | 1.17862 |
| Microbacteriaceae | 1.06613 | 1.06179 |
| env.OPS\_17 | 0.85030 | 0.84204 |
| uncultured | 0.82669 | 0.83509 |
| Mycobacteriaceae | 0.80599 | 0.81442 |
| Beijerinckiaceae | 0.72306 | 0.72184 |
| Spirosomaceae | 0.60753 | 0.61613 |
| Aeromonadaceae | 0.55232 | 0.56030 |
| Blastocatellaceae | 0.43648 | 0.42639 |
| Nocardiaceae | 0.36724 | 0.36777 |
| Propionibacteriaceae | 0.36610 | 0.36678 |
| Rhodanobacteraceae | 0.36084 | 0.35287 |
| Bacteriovoracaceae | 0.31219 | 0.31234 |
| Rhodocyclaceae | 0.30830 | 0.31850 |
| Alphaproteobacteria\_unclassified | 0.30386 | 0.30260 |
| Methylophilaceae | 0.29097 | 0.28552 |
| Gammaproteobacteria\_unclassified | 0.28070 | 0.27677 |
| Alteromonadaceae | 0.25693 | 0.25551 |
| Bacteria\_unclassified | 0.25264 | 0.25194 |
| Microscillaceae | 0.24414 | 0.23962 |
| Dysgonomonadaceae | 0.20361 | 0.18259 |
| Acetobacteraceae | 0.19044 | 0.18498 |
| Shewanellaceae | 0.17066 | 0.16014 |
| Rubritaleaceae | 0.14377 | 0.14822 |
| Xanthobacteraceae | 0.13429 | 0.12895 |
| Proteobacteria\_unclassified | 0.13272 | 0.13531 |
| Haliangiaceae | 0.13201 | 0.13690 |
| Paludibacteraceae | 0.13179 | 0.13590 |
| SM2D12 | 0.11850 | 0.11782 |
| Devosiaceae | 0.11832 | 0.11762 |
| Hymenobacteraceae | 0.11331 | 0.11484 |
| Polyangiaceae | 0.10762 | 0.10193 |
| Verrucomicrobiaceae | 0.10101 | 0.09299 |
| Opitutaceae | 0.09664 | 0.09477 |
| Leuconostocaceae | 0.09368 | 0.08742 |
| WPS-2\_fa | 0.09046 | 0.09040 |
| Micrococcaceae | 0.08059 | 0.08424 |
| Intrasporangiaceae | 0.08058 | 0.07391 |
| Bacteroidaceae | 0.07523 | 0.07570 |
| Nocardioidaceae | 0.07380 | 0.07491 |
| Chthoniobacteraceae | 0.06875 | 0.06378 |
| Crocinitomicaceae | 0.06746 | 0.06815 |
| Obscuribacterales\_fa | 0.06604 | 0.06338 |
| Myxococcales\_unclassified | 0.06248 | 0.06855 |
| Bacteroidia\_unclassified | 0.06223 | 0.06020 |
| Dermatophilaceae | 0.05847 | 0.06120 |
| Betaproteobacteriales\_unclassified | 0.05835 | 0.05365 |
| Sericytochromatia\_fa | 0.05239 | 0.05365 |
| Unknown\_Family | 0.04943 | 0.05047 |
| Thiovulaceae | 0.04654 | 0.04451 |
| 0319-6G20 | 0.04482 | 0.04510 |
| Sulfurospirillaceae | 0.04030 | 0.03954 |
| Hyphomicrobiaceae | 0.03759 | 0.03874 |
| Pirellulaceae | 0.03697 | 0.03676 |
| Cellvibrionaceae | 0.03498 | 0.03696 |
| Methylacidiphilaceae | 0.03480 | 0.03318 |
| Rhizobiales\_unclassified | 0.03462 | 0.03219 |
| Xiphinematobacteraceae | 0.03416 | 0.03179 |
| NS11-12\_marine\_group | 0.03365 | 0.03258 |
| Nakamurellaceae | 0.03102 | 0.03298 |
| Micrococcales\_unclassified | 0.03099 | 0.03258 |
| Legionellaceae | 0.03095 | 0.02742 |
| Cytophagaceae | 0.02972 | 0.02821 |
| Oligoflexaceae | 0.02955 | 0.02643 |
| Carnobacteriaceae | 0.02943 | 0.02682 |
| JG30-KF-CM45 | 0.02851 | 0.03060 |
| Deltaproteobacteria\_unclassified | 0.02837 | 0.02941 |
| Absconditabacteriales\_(SR1)\_fa | 0.02832 | 0.02841 |
| A0839 | 0.02808 | 0.02960 |
| Sphingobacteriales\_unclassified | 0.02711 | 0.02484 |
| Rhodopirillaceae | 0.02650 | 0.02086 |
| A4b | 0.02636 | 0.02424 |
| Beutenbergiaceae | 0.02489 | 0.02345 |
| Corynebacteriales\_unclassified | 0.02395 | 0.02424 |
| KD3-10 | 0.02356 | 0.02523 |
| Saprospiraceae | 0.02310 | 0.02325 |
| Prolixibacteraceae | 0.02103 | 0.02066 |
| JGI\_0000069-P22\_fa | 0.02022 | 0.02007 |
| Micavibrionales\_unclassified | 0.02002 | 0.01888 |
| Aquaspirillaceae | 0.01986 | 0.02126 |
| Magnetospirillaceae | 0.01908 | 0.01927 |
| Cytophagales\_unclassified | 0.01892 | 0.01768 |
| Clostridiaceae\_1 | 0.01836 | 0.01828 |
| Neisseriaceae | 0.01834 | 0.01788 |
| Fimbriimonadaceae | 0.01715 | 0.01689 |
| Hyphomonadaceae | 0.01712 | 0.01470 |
| WD2101\_soil\_group | 0.01682 | 0.01570 |
| Gemmataceae | 0.01595 | 0.01431 |
| Terrimicrobiaceae | 0.01591 | 0.01609 |
| Phaselicystidaceae | 0.01568 | 0.01590 |
| Actinobacteria\_unclassified | 0.01552 | 0.01629 |
| Fibrobacteraceae | 0.01437 | 0.01291 |
| Paracaedibacteraceae | 0.01382 | 0.01212 |
| Cellulomonadaceae | 0.01378 | 0.01291 |
| Diplorickettsiaceae | 0.01317 | 0.01252 |
| Rhodospirillaceae | 0.01308 | 0.01192 |
| Tannerellaceae | 0.01289 | 0.01172 |
| Saccharimonadales\_fa | 0.01275 | 0.01192 |
| Trueperaceae | 0.01255 | 0.01212 |
| Kineosporiaceae | 0.01250 | 0.01291 |
| Dermabacteraceae | 0.01106 | 0.01172 |
| Micromonosporaceae | 0.01098 | 0.00874 |
| Geodermatophilaceae | 0.00969 | 0.01073 |
| Bacteroidales\_unclassified | 0.00968 | 0.00815 |
| Methylopilaceae | 0.00953 | 0.00874 |
| Subgroup\_6\_fa | 0.00915 | 0.01113 |
| Tepidisphaeraceae | 0.00914 | 0.00894 |
| Deinococcaceae | 0.00863 | 0.00815 |
| Actinomycetaceae | 0.00856 | 0.00854 |
| Ruminococcaceae | 0.00842 | 0.00755 |
| Kaistiaceae | 0.00738 | 0.00616 |
| Saccharimonadaceae | 0.00737 | 0.00676 |
| Cyclobacteriaceae | 0.00736 | 0.00874 |
| Geminicoccaceae | 0.00730 | 0.00656 |
| Parachlamydiaceae | 0.00701 | 0.00616 |
| Marinilabiliaceae | 0.00700 | 0.00556 |
| Parcubacteria\_unclassified | 0.00689 | 0.00775 |
| Lachnospiraceae | 0.00674 | 0.00636 |
| Aerococcaceae | 0.00662 | 0.00715 |
| P3OB-42 | 0.00655 | 0.00695 |
| Xanthomonadales\_unclassified | 0.00623 | 0.00457 |
| Solirubrobacteraceae | 0.00619 | 0.00556 |
| Rhodospirillales\_unclassified | 0.00611 | 0.00596 |
| Procabacteriaceae | 0.00607 | 0.00556 |
| Desulfovibrionaceae | 0.00571 | 0.00636 |
| Solibacteraceae\_(Subgroup\_3) | 0.00568 | 0.00477 |
| Dietziaceae | 0.00526 | 0.00437 |
| Flavobacteriales\_unclassified | 0.00523 | 0.00397 |
| Oxyphotobacteria\_unclassified | 0.00514 | 0.00576 |
| Rickettsiaceae | 0.00505 | 0.00258 |
| Pleomorphomonadaceae | 0.00495 | 0.00517 |
| Rikenellaceae | 0.00490 | 0.00437 |
| Christensenellaceae | 0.00478 | 0.00338 |
| Geobacteraceae | 0.00471 | 0.00576 |
| Lactobacillales\_unclassified | 0.00452 | 0.00417 |
| Brevibacteriaceae | 0.00437 | 0.00457 |
| Nitrosomonadaceae | 0.00417 | 0.00358 |
| Rubinisphaeraceae | 0.00415 | 0.00616 |
| Synergistaceae | 0.00413 | 0.00457 |
| Micavibrionaceae | 0.00407 | 0.00417 |
| 37-13 | 0.00405 | 0.00417 |
| Acidobacteriaceae\_(Subgroup\_1) | 0.00395 | 0.00238 |
| FBP\_fa | 0.00395 | 0.00298 |
| Caedibacteraceae | 0.00392 | 0.00397 |
| Gracilibacteria\_fa | 0.00382 | 0.00358 |
| Chloroflexaceae | 0.00379 | 0.00397 |
| Caldilineaceae | 0.00372 | 0.00258 |
| Lentimicrobiaceae | 0.00360 | 0.00238 |
| Elsteraceae | 0.00353 | 0.00358 |
| Tepidisphaerales\_unclassified | 0.00348 | 0.00238 |
| Clostridiales\_unclassified | 0.00347 | 0.00179 |
| Saccharimonadales\_unclassified | 0.00347 | 0.00318 |
| OPB56\_fa | 0.00342 | 0.00318 |
| Desulfobulbaceae | 0.00333 | 0.00397 |
| Myxococcaceae | 0.00331 | 0.00397 |
| Gemmatimonadaceae | 0.00329 | 0.00358 |
| Reyranellaceae | 0.00326 | 0.00397 |
| Verrucomicrobiae\_unclassified | 0.00322 | 0.00397 |
| Rickettsiales\_unclassified | 0.00318 | 0.00219 |
| Gastranaerophilales\_fa | 0.00311 | 0.00278 |
| Micrococcales\_Incertae\_Sedis | 0.00304 | 0.00219 |
| Rhodothermaceae | 0.00302 | 0.00219 |
| Solimonadaceae | 0.00290 | 0.00397 |
| KD4-96\_fa | 0.00274 | 0.00397 |
| Pedosphaeraceae | 0.00273 | 0.00199 |
| Ignavibacteria\_unclassified | 0.00271 | 0.00298 |
| 67-14 | 0.00266 | 0.00238 |
| Sanguibacteraceae | 0.00264 | 0.00099 |
| Holophagaceae | 0.00260 | 0.00159 |
| Ardenticatenaceae | 0.00259 | 0.00219 |
| PB19\_fa | 0.00255 | 0.00358 |
| Micropepsaceae | 0.00252 | 0.00199 |
| Chloroflexi\_unclassified | 0.00242 | 0.00179 |
| mle1-27 | 0.00231 | 0.00238 |
| Chlamydiales\_unclassified | 0.00231 | 0.00179 |
| Roseiflexaceae | 0.00230 | 0.00318 |
| Veillonellaceae | 0.00230 | 0.00219 |
| SC-I-84 | 0.00226 | 0.00278 |
| Chitinibacteraceae | 0.00218 | 0.00219 |
| Oligoflexales\_unclassified | 0.00213 | 0.00139 |
| Halomonadaceae | 0.00212 | 0.00219 |
| Sandaracinaceae | 0.00209 | 0.00199 |
| Kallotenuales\_unclassified | 0.00205 | 0.00199 |
| Gaiellaceae | 0.00200 | 0.00159 |
| Demequinaceae | 0.00196 | 0.00199 |
| Deferribacteraceae | 0.00195 | 0.00219 |
| Anaerolineaceae | 0.00178 | 0.00199 |
| SJA-28\_fa | 0.00168 | 0.00199 |
| Bacillales\_unclassified | 0.00163 | 0.00099 |
| SBR1031\_fa | 0.00159 | 0.00119 |
| Herpetosiphonaceae | 0.00159 | 0.00179 |
| Alteromonadales\_unclassified | 0.00159 | 0.00179 |
| WCHB1-41\_fa | 0.00150 | 0.00159 |
| Streptomycetaceae | 0.00141 | 0.00099 |
| Coxiellaceae | 0.00140 | 0.00119 |
| R7C24\_fa | 0.00137 | 0.00219 |
| Acidithiobacillaceae | 0.00135 | 0.00139 |
| Promicromonosporaceae | 0.00134 | 0.00099 |
| Vampirovibrionales\_fa | 0.00132 | 0.00179 |
| Bacillaceae | 0.00127 | 0.00159 |
| Chitinophagales\_unclassified | 0.00124 | 0.00099 |
| Phycisphaerae\_unclassified | 0.00124 | 0.00159 |
| Chroococcidiopsaceae | 0.00122 | 0.00060 |
| 01D2Z36 | 0.00118 | 0.00079 |
| Amoebophilaceae | 0.00113 | 0.00060 |
| CHAB-XI-27\_fa | 0.00110 | 0.00060 |
| vadinHA49\_fa | 0.00109 | 0.00099 |
| Spirochaetaceae | 0.00109 | 0.00159 |
| Phycisphaeraceae | 0.00107 | 0.00060 |
| Frankiales\_unclassified | 0.00107 | 0.00159 |
| Iamiaceae | 0.00104 | 0.00119 |
| UA11\_fa | 0.00103 | 0.00119 |
| KD1-131 | 0.00098 | 0.00199 |
| AKYH767 | 0.00096 | 0.00079 |
| Dongiaceae | 0.00095 | 0.00119 |
| Subgroup\_6\_unclassified | 0.00095 | 0.00079 |
| Ignavibacteriales\_unclassified | 0.00092 | 0.00099 |
| Bacteroidetes\_vadinHA17 | 0.00090 | 0.00099 |
| Rhizobiales\_Incertae\_Sedis | 0.00089 | 0.00060 |
| RBG-13-54-9\_fa | 0.00089 | 0.00079 |
| Babeliales\_unclassified | 0.00087 | 0.00079 |
| Staphylococcaceae | 0.00086 | 0.00040 |
| TK10\_fa | 0.00086 | 0.00060 |
| Cyanobacteria\_unclassified | 0.00086 | 0.00099 |
| NS9\_marine\_group | 0.00082 | 0.00099 |
| Prevotellaceae | 0.00082 | 0.00079 |
| M2PB4-65\_termite\_group | 0.00080 | 0.00099 |
| AKIW781 | 0.00080 | 0.00040 |
| Thermoleophilia\_unclassified | 0.00079 | 0.00079 |
| uncultured\_fa | 0.00078 | 0.00099 |
| Schlesneriaceae | 0.00076 | 0.00099 |
| Desulfuromonadales\_unclassified | 0.00076 | 0.00040 |
| Elusimicrobiaceae | 0.00075 | 0.00079 |
| Lactobacillaceae | 0.00074 | 0.00060 |
| Archangiaceae | 0.00074 | 0.00040 |
| Victivallaceae | 0.00071 | 0.00159 |
| Propionibacteriales\_unclassified | 0.00071 | 0.00040 |
| Chitinimonadaceae | 0.00070 | 0.00040 |
| SBR1031\_unclassified | 0.00069 | 0.00060 |
| Bifidobacteriaceae | 0.00067 | 0.00000 |
| Chthonomonadaceae | 0.00067 | 0.00060 |
| Puniceicoccaceae | 0.00066 | 0.00139 |
| Steroidobacteraceae | 0.00065 | 0.00079 |
| Pseudomonadales\_unclassified | 0.00065 | 0.00079 |
| Isosphaeraceae | 0.00064 | 0.00099 |
| Verrucomicrobiales\_unclassified | 0.00063 | 0.00099 |
| Nostocaceae | 0.00063 | 0.00060 |
| Dermacoccaceae | 0.00063 | 0.00000 |
| Planctomycetes\_unclassified | 0.00061 | 0.00040 |
| Holosporaceae | 0.00060 | 0.00060 |
| ABY1\_unclassified | 0.00059 | 0.00020 |
| OM190\_fa | 0.00059 | 0.00000 |
| Nannocystaceae | 0.00059 | 0.00079 |
| Midichloriaceae | 0.00058 | 0.00060 |
| Bernardetiaceae | 0.00057 | 0.00060 |
| Stappiaceae | 0.00056 | 0.00119 |
| Syntrophomonadaceae | 0.00056 | 0.00040 |
| D05-2 | 0.00054 | 0.00020 |
| CPR2\_fa | 0.00053 | 0.00020 |
| Parcubacteria\_fa | 0.00053 | 0.00020 |
| Blfdi19 | 0.00051 | 0.00020 |
| Caulobacterales\_unclassified | 0.00051 | 0.00060 |
| Subgroup\_7\_fa | 0.00050 | 0.00060 |
| Melainabacteria\_unclassified | 0.00049 | 0.00040 |
| KD3-93 | 0.00049 | 0.00139 |
| Pseudonocardiaceae | 0.00048 | 0.00079 |
| Campylobacterales\_unclassified | 0.00048 | 0.00020 |
| Thermoanaerobaculaceae | 0.00046 | 0.00079 |
| Gaiellales\_unclassified | 0.00046 | 0.00079 |
| Family\_XII | 0.00044 | 0.00020 |
| Labraceae | 0.00044 | 0.00040 |
| Simkaniaceae | 0.00043 | 0.00079 |
| Orbaceae | 0.00043 | 0.00060 |
| Planococcaceae | 0.00043 | 0.00119 |
| Lineage\_IIb\_fa | 0.00042 | 0.00060 |
| BIrii41 | 0.00040 | 0.00099 |
| Phormidiaceae | 0.00040 | 0.00040 |
| Chthoniobacterales\_unclassified | 0.00039 | 0.00000 |
| Firmicutes\_unclassified | 0.00039 | 0.00020 |
| Candidatus\_Falkowbacteria\_fa | 0.00037 | 0.00040 |
| Eubacteriaceae | 0.00037 | 0.00000 |
| AB1 | 0.00037 | 0.00040 |
| Nitrincolaceae | 0.00034 | 0.00000 |
| Cryptosporangiaceae | 0.00034 | 0.00060 |
| Parvibaculaceae | 0.00034 | 0.00000 |
| Bacteroidetes\_unclassified | 0.00033 | 0.00020 |
| Frankiaceae | 0.00033 | 0.00040 |
| Rokubacteriales\_fa | 0.00032 | 0.00020 |
| EV818SWSAP88\_fa | 0.00030 | 0.00020 |
| Streptosporangiaceae | 0.00030 | 0.00020 |
| T34 | 0.00029 | 0.00020 |
| SB-5 | 0.00029 | 0.00000 |
| Subgroup\_17\_fa | 0.00029 | 0.00040 |
| SAR324\_clade(Marine\_group\_B)\_fa | 0.00028 | 0.00020 |
| Microtrichaceae | 0.00027 | 0.00040 |
| Hydrogenophilaceae | 0.00027 | 0.00000 |
| Ilumatobacteraceae | 0.00026 | 0.00000 |
| Lineage\_IV\_fa | 0.00025 | 0.00000 |
| Candidatus\_Zambryskibacteria\_fa | 0.00025 | 0.00000 |
| Methyloligellaceae | 0.00024 | 0.00040 |
| Leptospiraceae | 0.00024 | 0.00020 |
| Leptotrichiaceae | 0.00024 | 0.00020 |
| MB-A2-108\_fa | 0.00023 | 0.00020 |
| C0119\_fa | 0.00023 | 0.00000 |
| Gitt-GS-136\_fa | 0.00022 | 0.00079 |
| Anaplasmataceae | 0.00022 | 0.00040 |
| Peptococcaceae | 0.00022 | 0.00020 |
| TRA3-20 | 0.00022 | 0.00000 |
| Chloroflexales\_unclassified | 0.00021 | 0.00040 |
| 11-24\_fa | 0.00021 | 0.00020 |
| Thermaceae | 0.00021 | 0.00000 |
| Gracilibacteria\_unclassified | 0.00021 | 0.00040 |
| Microtrichales\_unclassified | 0.00020 | 0.00000 |
| Sneathiellaceae | 0.00020 | 0.00000 |
| Solirubrobacterales\_unclassified | 0.00020 | 0.00040 |
| Armatimonadales\_fa | 0.00019 | 0.00020 |
| Gracilibacteraceae | 0.00019 | 0.00040 |
| Kiritimatiellaceae | 0.00019 | 0.00020 |
| Hydrogenedensaceae | 0.00019 | 0.00000 |
| Enterococcaceae | 0.00019 | 0.00000 |
| Alcanivoracaceae | 0.00018 | 0.00020 |
| OPB41\_fa | 0.00018 | 0.00020 |
| ST-12K33 | 0.00018 | 0.00000 |
| Vermiphilaceae | 0.00017 | 0.00000 |
| DEV007 | 0.00017 | 0.00020 |
| TC1 | 0.00016 | 0.00020 |
| Babeliales\_fa | 0.00016 | 0.00020 |
| V2072-189E03\_fa | 0.00016 | 0.00040 |
| BRC1\_fa | 0.00015 | 0.00000 |
| Planctomycetales\_unclassified | 0.00015 | 0.00000 |
| Acidobacteriales\_unclassified | 0.00015 | 0.00020 |
| Desulfomicrobiaceae | 0.00015 | 0.00000 |
| Candidatus\_Peribacteria\_fa | 0.00015 | 0.00000 |
| Rhodothermia\_unclassified | 0.00014 | 0.00060 |
| Babeliaceae | 0.00014 | 0.00020 |
| Gallionellaceae | 0.00014 | 0.00020 |
| Gaiellales\_fa | 0.00014 | 0.00020 |
| Anaerolineae\_unclassified | 0.00014 | 0.00000 |
| Nitrospiraceae | 0.00014 | 0.00020 |
| PeM15\_fa | 0.00014 | 0.00000 |
| FTLpost3 | 0.00014 | 0.00020 |
| Thermomicrobiales\_unclassified | 0.00013 | 0.00000 |
| Patescibacteria\_unclassified | 0.00012 | 0.00000 |
| Erysipelotrichaceae | 0.00012 | 0.00000 |
| UBA12409 | 0.00012 | 0.00020 |
| Listeriaceae | 0.00012 | 0.00020 |
| SM1A07\_fa | 0.00011 | 0.00000 |
| DS-100\_fa | 0.00011 | 0.00020 |
| Ktedonobacteraceae | 0.00011 | 0.00000 |
| Aminicenantales\_fa | 0.00011 | 0.00040 |
| Rarobacteraceae | 0.00011 | 0.00000 |
| Sporolactobacillaceae | 0.00010 | 0.00000 |
| Tistrellaceae | 0.00010 | 0.00000 |
| Bogoriellaceae | 0.00010 | 0.00000 |
| PLTA13\_fa | 0.00010 | 0.00000 |
| Armatimonadetes\_unclassified | 0.00010 | 0.00020 |
| Chloroflexia\_unclassified | 0.00009 | 0.00000 |
| Longimicrobiaceae | 0.00009 | 0.00000 |
| Planctomycetacia\_unclassified | 0.00009 | 0.00000 |
| Fodinicurvataceae | 0.00009 | 0.00000 |
| Sulfuricellaceae | 0.00009 | 0.00000 |
| Syntrophaceae | 0.00009 | 0.00000 |
| Leptolyngbyaceae | 0.00009 | 0.00000 |
| Cloacimonadales\_unclassified | 0.00009 | 0.00020 |
| Subgroup\_5\_fa | 0.00009 | 0.00020 |
| Acidothermaceae | 0.00008 | 0.00000 |
| cvE6 | 0.00008 | 0.00000 |
| CCD24\_fa | 0.00008 | 0.00000 |
| Cellvibrionales\_unclassified | 0.00008 | 0.00000 |
| Victivallales\_unclassified | 0.00008 | 0.00020 |
| WS4\_fa | 0.00008 | 0.00000 |
| AD3\_fa | 0.00007 | 0.00000 |
| Victivallales\_fa | 0.00007 | 0.00000 |
| Paenibacillaceae | 0.00007 | 0.00000 |
| Alicyclobacillaceae | 0.00007 | 0.00000 |
| Jonesiaceae | 0.00007 | 0.00020 |
| Clostridia\_unclassified | 0.00007 | 0.00000 |
| BRH-c20a\_fa | 0.00007 | 0.00000 |
| Coriobacteriia\_unclassified | 0.00007 | 0.00000 |
| Marinifilaceae | 0.00007 | 0.00020 |
| Acidobacteria\_unclassified | 0.00007 | 0.00020 |
| bac2nit3 | 0.00007 | 0.00000 |
| Armatimonadales\_unclassified | 0.00007 | 0.00020 |
| Ignavibacteriaceae | 0.00006 | 0.00020 |
| vadinBE97 | 0.00006 | 0.00020 |
| Chromobacteriaceae | 0.00006 | 0.00000 |
| Porticoccaceae | 0.00006 | 0.00000 |
| Candidatus\_Magasanikbacteria\_fa | 0.00006 | 0.00000 |
| Candidatus\_Moranbacteria\_fa | 0.00005 | 0.00000 |
| S085\_fa | 0.00005 | 0.00020 |
| Bradymonadales\_fa | 0.00005 | 0.00000 |
| Bacilli\_unclassified | 0.00005 | 0.00020 |
| Candidatus\_Nomurabacteria\_fa | 0.00004 | 0.00000 |
| Latescibacteria\_fa | 0.00004 | 0.00000 |
| Pla4\_lineage\_fa | 0.00004 | 0.00020 |
| FCPU426\_fa | 0.00004 | 0.00000 |
| Barnesiellaceae | 0.00004 | 0.00000 |
| MWH-CFBk5 | 0.00004 | 0.00000 |
| CPla-3\_termite\_group | 0.00004 | 0.00000 |
| Omnitrophaceae | 0.00004 | 0.00000 |
| SM1B06 | 0.00004 | 0.00000 |
| Pyrinomonadaceae | 0.00004 | 0.00000 |
| Thermomonosporaceae | 0.00004 | 0.00000 |
| Sporichthyaceae | 0.00004 | 0.00000 |
| NB1-j\_fa | 0.00003 | 0.00000 |
| Family\_III | 0.00003 | 0.00000 |
| Pseudohongiellaceae | 0.00003 | 0.00000 |
| Segniliparaceae | 0.00003 | 0.00000 |
| LiUU-11-161 | 0.00003 | 0.00000 |
| Muribaculaceae | 0.00003 | 0.00000 |
| Helicobacteraceae | 0.00003 | 0.00000 |
| Azospirillales\_Incertae\_Sedis | 0.00003 | 0.00000 |
| BSV26 | 0.00002 | 0.00000 |
| SRB2 | 0.00002 | 0.00000 |
| Family\_XIII | 0.00002 | 0.00000 |
| Lineage\_IIa\_fa | 0.00002 | 0.00000 |
| Desulfobacteraceae | 0.00002 | 0.00000 |
| Acidimicrobiia\_unclassified | 0.00002 | 0.00000 |
| Candidatus\_Kaiserbacteria\_fa | 0.00002 | 0.00000 |
| SS1-B-06-26 | 0.00002 | 0.00000 |
| Lentisphaerae\_unclassified | 0.00002 | 0.00020 |
| Pasteurellaceae | 0.00002 | 0.00000 |
| Acetobacterales\_unclassified | 0.00002 | 0.00000 |
| Cloacimonadaceae | 0.00002 | 0.00000 |
| Criblamydiaceae | 0.00002 | 0.00000 |
| Gemmatimonadetes\_unclassified | 0.00002 | 0.00000 |
| Nocardiopsaceae | 0.00001 | 0.00000 |
| Acetobacterales\_Incertae\_Sedis | 0.00001 | 0.00000 |
| S-70\_fa | 0.00001 | 0.00000 |
| Azospirillales\_unclassified | 0.00001 | 0.00000 |
| Candidatus\_Amesbacteria\_fa | 0.00001 | 0.00000 |
| Eggerthellaceae | 0.00001 | 0.00000 |
| Elusimicrobia\_unclassified | 0.00001 | 0.00000 |
| Gimesiaceae | 0.00001 | 0.00000 |
| Mollicutes\_RF39\_fa | 0.00001 | 0.00000 |
| MVP-15\_fa | 0.00001 | 0.00000 |
| possible\_family\_01 | 0.00001 | 0.00000 |



**Figure S1**: Microbial network indicating co-occurrence of fungal families identified in samples collected from all three facilities. Green edges represent positive relationship (co-occurrence) among families, whereas the red edges represent negative relationship (co-exclusion) between two connected nodes.



**Figure S2**: Microbial network indicating co-occurrence of bacterial and fungal families in samples collected from three facilities F1, F2, and F3 combined. Green edges represent positive relationship (co-occurrence) among families, whereas the red edges represent negative relationship (co-exclusion) between two connected nodes.

**List L1:** Data analyses workflow

Downloadable workflow file is available on the following link: <https://github.com/kovaclab/Apple-packing-house-environmental-microbiomes/blob/master/Tan%20et%20al.%202019%20data%20anaylsis%20workflow.R>

#Data analysis workflow

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#Microbiome sequence data analyses (carried out in Mothur v 1.39.5)

#From the raw sequencing files (.fastq)

#Set the input directory and number of processing units

#Making.files which include list of samples and the sequences associated with those samples (paired R1, R2)

mothur > make.file(inputdir=., type=fastq, prefix=16s)

#Combine the paired-end reads together

#Extract the sequence and quality score data from fastq files, create the contigs

mothur > make.contigs(file=16s.files)

#Generate descriptive statistics of all sequences in .fasta files

mothur > summary.seqs(fasta=16s.trim.contigs.fasta)

#Filter sequences

##Based on the summary statistics, remove any sequences with ambiguous bases ("N"), shorter than 292, longer than 292

mothur > screen.seqs(fasta=16s.trim.contgis.fasta,

group=16s.contigs.groups, summary=16s.trim.contigs.summary,

minlength=292, maxlenth=292, maxambig=0)

#Remove identical (grouped) sequences; representative sequence will be picked and stored in .fasta and corresponding sequences will be saved as sequence names to reduce computational work

mothur > unique.seqs(fasta=16s.trim.contigs.good.fasta)

#Create a count table of current unique sequences

mothur > count.seqs(name=16s.trim.contigs.good.names,

group=16s.contigs.good.groups)

#Analyze the target 16S rRNA V4 region

#Customize database to target V4 region

#Define start and end positions within a 16S rRNA sequence

#Set keepdots=F to false to remove output trailing dots from fragments

mothur > pcr.seqs(fasta=silva.nr\_v132.align, start=11894, end=25319,

keepdots=F, processors=8)

#Rename reference files

mothur > rename.file(input = silva.bacteria.pcr.fasta, new=

silva.v4.fasta)

#Align the target region to SILVA reference database

#Determined k size (ksize=); default is 8

#Allow reverse matching to the reference using flip=T

mothur > align.seqs(fasta=16s.trim.contigs.good.unique.fasta,

reference=silva.v4.fasta, flip=T)

#Generate descriptive statistics for all sequences in .fasta files using summary.seqs

mothur > summary.seqs(fasta=16s.trim.contigs.good.unique.align,

count=16s.trim.contigs.good.count\_table)

#Remove sequences that are before or after the sites of alignment from the previous step

mothur > screen.seqs(fasta=16s.trim.contgis.good.unique.align,

group=16s.contigs.good.count\_table,

summary=16s.trim.contigs.good.unique.summary, minlength=Undecided,

maxlenth=Undecided, maxhomop=8)

#Remove overhangs

#Remove the alignment characters that only consist of "-", using vertical=T

#Remove the sequences containing '.' by using trump=.

mothur > filter.seqs(fasta=16s.trim.contigs.good.unique.good.align,

vertical=T, trump=.)

#Rerun unique.seqs in case new redundant sequences were created by filtering

mothur > unique.seqs(fasta=16s.trim.contigs.good.unique.good.filter.fasta,

count = 16s.trim.contigs.good.good.count\_table)

#De-noise sequences

#Set diffs=2 as a threshold of mismatches in the sequence

mothur > pre.cluster(fasta=16s.trim.contigs.good.unique.good.filter.unique.fast

a, count=16s.trim.contigs.good.unique.good.filter.count\_table,

diffs=2)

#Read .fasta and count file to chimera sequences

mothur > chimera.vsearch(fasta=16s.trim.contigs.good.unique.good.filter.unique.p

recluster.fasta,

count=16s.trim.contigs.good.unique.good.filter.unique.precluster.count

\_table, dereplicate=t)

#Remove chimera

mothur > remove.seqs(fasta=16s.trim.contigs.good.unique.good.filter.unique.prec

luster.fasta,

accnos=16s.trim.contigs.good.unique.good.filter.unique.precluster.deno

vo.vsearch.accnos)

#Assign taxonomy

mothur > classify.seqs(fasta=16s.trim.contigs.good.unique.good.filter.unique.pr

ecluster.pick.fasta,

count=16s.trim.contigs.good.unique.good.filter.unique.precluster.denov

o.uchime.pick.count\_table, reference=silva.nr\_v123.align,

taxonomy=silva.nr\_v123.tax)

#Remove chloroplast and mitochondria sequences

mothur > remove.lineage(fasta=16s.trim.contigs.good.unique.good.filter.unique.p

recluster.pick.fasta,

count=16s.trim.contigs.good.unique.good.filter.unique.precluster.denov

o.uchime.pick.count\_table,

taxonomy=16s.trim.contigs.good.unique.good.filter.unique.precluster.pi

ck.nr\_v132.wang.taxonomy, taxon=Chloroplast-Mitochondria-unknown-

Eukaryota)

#Calculate uncorrected pairwise distances between aligned DNA sequences. By default, a gap is penalized; cutoff value indicates that distances larger than 0.03 (>97% similarity) will not be saved

mothur > dist.seqs(fasta=16s.trim.contigs.good.unique.good.filter.unique.preclu

ster.pick.pick.fasta, cutoff=0.03)

#Assign sequences to OTUs, Mothur provides three different methods of alignment. By default, opticlust method is used.

mothur > cluster(column=16s.trim.contigs.good.unique.good.filter.unique.preclus

ter.pick.pick.dist, count =

16s.trim.contigs.good.unique.good.filter.unique.precluster.denovo.vsea

rch.pick.pick.count\_table)

#Determine how many sequences are assigned to each OTU at the 0.03 cutoff. Distribute OTUs into groups

#The output shared file is used as an OTU table

mothur > make.shared(list=16s.trim.contigs.good.unique.good.filter.unique.precl

uster.pick.pick.opti\_mcc.list,

count=16s.trim.contigs.good.unique.good.filter.unique.precluster.denov

o.vsearch.pick.pick.count\_table, label=0.03)

#Determine taxonomy for all OTUs. Output of this command is a taxonomy file

mothur > classify.otu(list=16s.trim.contigs.good.unique.good.filter.unique.prec

luster.pick.pick.opti\_mcc.list,

count=16s.trim.contigs.good.unique.good.filter.unique.precluster.denov

o.vsearch.pick.pick.count\_table,

taxonomy=16s.trim.contigs.good.unique.good.filter.unique.precluster.pi

ck.pds.wang.pick.taxonomy, label=0.03)

####Downstream analysis###

#Plot L. monocytogenes occurrence in three facilities using a bar plot

#Load the file containing data about facility and L. monocytogenes occurrence

lm <- read.csv(file.choose(), header = T)

lmoccurrence <- ggplot(data = lm, aes(x=Facility, y=Number.of.samples,fill=L..monocytogenes))+

geom\_bar(stat = "identity",width = 0.5)+

geom\_text(aes(label=Number.of.samples, size=3), hjust=0.5, vjust=3) +

theme\_bw(base\_size = 12)+

theme(legend.text=element\_text(size=15), legend.title= element\_text(size=15)) +

theme(axis.text.x = element\_text(size=13), axis.text.y = element\_text(size=13)) +

theme(axis.title = element\_text(size=15)) +

scale\_x\_discrete(name = "Facility") +

scale\_y\_discrete(name = "Number of Samples")

ggsave("lmoccurrence.pdf", plot = lmoccurrence, device="pdf", width=10, height=7, units="in",dpi=600)

#Plot the PCoA

#Obtain required R packages

library(phyloseq)

library(ape)

library(vegan)

library(ggplot2)

#Import data

set.seed(336)

otus <- import\_mothur(mothur\_shared\_file= file.choose())

otus2 <- as.data.frame(otus)

otus.t <- t(otus)

min(rowSums(otus.t))

otus.r <- rrarefy(otus.t,4501)

OTU <- otu\_table(otus.r , taxa\_are\_rows=FALSE)

taxon <- import\_mothur(mothur\_constaxonomy\_file = file.choose())

taxon <- as.data.frame(taxon)

colnames(taxon) <- c("Domain", "Phylum", "Class", "Order", "Family", "Genus")

TAX = tax\_table(as.matrix(taxon))

metadat <- read.table(file.choose(), sep=",", header=T, row.names=1)

META = sample\_data(metadat)

#Plot PCoA for rarefied samples (16S rRNA data)

phyloseq = phyloseq(OTU, TAX, META)

TREE = rtree(ntaxa(phyloseq), rooted=TRUE, tip.label = taxa\_names(phyloseq))

phyloseq = phyloseq(OTU,TAX,META,TREE)

phyloseq

ord = ordinate (phyloseq, "PCoA", "unifrac", weighted = TRUE)

po = plot\_ordination(phyloseq, ord, color="Facility",shape="L..monocytogenes")

PCOA16S <- po +

geom\_point(size=4)+theme\_classic() +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.text.x = element\_text(size=13), axis.text.y = element\_text(size=13)) +

theme(axis.title = element\_text(size=15)) +

scale\_x\_continuous(name = "PC1 (10.7%)") +

scale\_y\_continuous(name = "PC2 (6.6%)")

PCOA16S

#Import data for ITS

set.seed(336)

otus\_ITS <- import\_mothur(mothur\_shared\_file= file.choose())

otus2\_ITS <- as.data.frame(otus\_ITS)

otus.t\_ITS <- t(otus\_ITS)

min(rowSums(otus.t\_ITS))

otus.r\_ITS <- rrarefy(otus.t\_ITS,5323)

OTU\_ITS <- otu\_table(otus.r\_ITS , taxa\_are\_rows=FALSE)

taxon\_ITS <- import\_mothur(mothur\_constaxonomy\_file = file.choose())

taxon\_ITS <- as.data.frame(taxon\_ITS)

colnames(taxon\_ITS) <- c("Domain", "Phylum", "Class", "Order", "Family", "Genus")

TAX\_ITS = tax\_table(as.matrix(taxon\_ITS))

metadat\_ITS <- read.table(file.choose(), sep=",", header=T, row.names=1)

META\_ITS = sample\_data(metadat\_ITS)

#Plot PCoA for rarefied samples (ITS data)

phyloseq\_ITS = phyloseq(OTU, TAX, META)

TREE\_ITS = rtree(ntaxa(phyloseq\_ITS), rooted=TRUE, tip.label = taxa\_names(phyloseq\_ITS))

phyloseq\_ITS = phyloseq(OTU,TAX,META,TREE\_ITS)

phyloseq\_ITS

ord\_ITS = ordinate (phyloseq\_ITS, "PCoA", "unifrac", weighted = TRUE)

po\_ITS = plot\_ordination(phyloseq\_ITS, ord\_ITS, color="Facility",shape="L.monocytogenes")

PCOAITS <- po\_ITS +

geom\_point(size=4)+theme\_classic() +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.text.x = element\_text(size=13), axis.text.y = element\_text(size=13)) +

theme(axis.title = element\_text(size=15)) +

scale\_x\_continuous(name = "PC1 (43.1%)") +

scale\_y\_continuous(name = "PC2 (20.5%)")

PCOA combined plot

#Combine the 16S rRNA and ITS PCoA plots

a = plot\_grid(PCOA16S + theme(legend.position= "none") , PCOAITS + theme(legend.position = "none") ,

ncol=1, nrow=2, labels=c("A", "B"), label\_size = 20)

b = get\_legend(PCOAITS)

c = plot\_grid(a, b, ncol=2, rel\_widths = c(3,1))

c

ggsave("PCOAcombined.pdf", plot =c, device="pdf", width=10, height=10, units="in",dpi=600)

ggsave("PCOAcombined.png", plot =c, device="png", width=10, height=10, units="in",dpi=600)

#Stack barplot for bacterial and fungal communities at a family level

#Use phyoseq objects for 16S rRNA and ITS

#Melt to long format (for ggploting)

family\_16 <- phyloseq %>%

tax\_glom(taxrank = "Family") %>% #agglomerate at a family level

transform\_sample\_counts(function(x) {x/sum(x)} ) %>% #Transform to relative abundance

psmelt() %>% #Melt to long format

arrange(Family) #Sort data frame alphabetically by phylum

family\_ITS <- phyloseq\_ITS %>%

tax\_glom(taxrank = "Rank5") %>%

transform\_sample\_counts(function(x) {x/sum(x)} ) %>%

psmelt() %>%

arrange(Family)

write.csv(family\_16, "combined\_family\_16.csv")

write.csv(family\_ITS, "combined\_family\_ITS.csv") #Write the filtered file into csv format

#Filter the 'Abundance' column to 'less than' 0.10 (in Excel)

#All of the outcome rows are representative families with abundance lower than 0.10

#Change all column labels under 'Family' to "Other"

#Save as .csv file

#Open file in R

#Figure 1: Facility vs. L. monocytogenes

combined\_family\_16s <- read.csv(file.choose(), sep=",", header=T, row.names=1)

Family\_colors <- c("#FFF5F0","#525252","#CB181D","#99000D","#EF3B2C","#FB6A4A","#FC9272","#FEE0D2","#4292C6","#084594","#FFF5F0","#EF3B2C","#9ECAE1","#C6DBEF","#DEEBF7","#6BAED6","#2171B5","#737373")

#Plot Figure 1

fig1 <- ggplot(combined\_family\_16s, aes(x = SampleOrder, y = Abundance , fill = Family)) +

facet\_grid(Facility~Lmono)+

geom\_bar(stat = "identity") +

scale\_fill\_manual(values = Family\_colors) +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_blank(), axis.text.x=element\_blank(), axis.ticks.x=element\_blank(),

axis.title.y=element\_text(size=15)) +

guides(fill = guide\_legend(reverse = FALSE, keywidth = 1, keyheight = 1, ncol=1)) +

ylab("Relative Abundance") + theme(panel.background = element\_rect(fill="transparent", color =NA),

plot.background = element\_rect(fill="transparent", color =NA)) +

theme(strip.background= element\_blank(), strip.text = element\_text(size=15),

panel.border = element\_rect(color="black", fill=NA))

fig1

ggsave("figure1.pdf", plot =fig1, device="pdf", width=8, height=5, units="in",dpi=600)

#Figure 2: Facility vs. section

combined\_family\_16s <- read.csv(file.choose(), sep=",", header=T, row.names=1)

Family\_colors <- c("#FFF5F0","#525252","#CB181D","#99000D","#EF3B2C","#FB6A4A","#FC9272","#FEE0D2","#4292C6","#084594","#FFF5F0","#EF3B2C","#9ECAE1","#C6DBEF","#DEEBF7","#6BAED6","#2171B5","#737373")

#Plot Figure 2

fig2 <- ggplot(combined\_family\_16s, aes(x = SampleOrder, y = Abundance , fill = Family)) +

facet\_grid(Facility~Section)+

geom\_bar(stat = "identity") +

scale\_fill\_manual(values = Family\_colors) +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_blank(), axis.text.x=element\_blank(), axis.ticks.x=element\_blank(),

axis.title.y=element\_text(size=15)) +

guides(fill = guide\_legend(reverse = FALSE, keywidth = 1, keyheight = 1, ncol=1)) +

ylab("Relative Abundance") + theme(panel.background = element\_rect(fill="transparent", color =NA),

plot.background = element\_rect(fill="transparent", color =NA)) +

theme(strip.background= element\_blank(), strip.text = element\_text(size=15),

panel.border = element\_rect(color="black",fill=NA))

fig2

ggsave("figure2.pdf", plot =fig1, device="pdf", width=8, height=5, units="in",dpi=600)

ggsave("figure2.png", plot =fig1, device="png", width=8, height=5, units="in",dpi=600)

library(cowplot)

a = plot\_grid(fig1 + theme(legend.position= "none") , fig2 + theme(legend.position = "none") ,

ncol=1, nrow=2, labels=c("A", "B"), label\_size = 20)

b = get\_legend(fig2)

c = plot\_grid(a, b, ncol=3, rel\_widths = c(10,1))

c

ggsave("figre\_combine.pdf", plot=c, device="pdf", width=10, height=7, units="in", dpi=600)

ggsave("figre\_combine.png", plot=c, device="png", width=10, height=7, units="in", dpi=600)

#Plot a stack bar plot based on ITS data

combined\_family\_ITS <- read.csv(file.choose(), sep=",", header=T, row.names=1)

Family\_colors <- c("#084594", "#2171B5", "#4292C6","#9ECAE1","#FFF5F0","#C6DBEF", "#DEEBF7","#6BAED6" ,"#99000D","#EF3B2C","#FC9272","#F7FBFF","#FB6A4A", "#FCBBA1", "#FEE0D2","#525252","#737373","#CB181D","#FFDAB9","#E6E6FA")

#Plot facility vs. L. monocytogenes occurence

ITSfacilitystack <- ggplot(combined\_family\_ITS, aes(x = SampleOrder, y = Abundance, fill = Family)) + facet\_grid(Facility~L.monocytogenes)+

geom\_bar(stat = "identity") +

geom\_bar(stat = "identity") +

scale\_fill\_manual(values = Family\_colors) +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_blank(), axis.text.x=element\_blank(), axis.ticks.x=element\_blank(),

axis.title.y=element\_text(size=15)) +

guides(fill = guide\_legend(reverse = FALSE, keywidth = 1, keyheight = 1, ncol=1)) +

ylab("Relative Abundance") + theme(panel.background = element\_rect(fill="transparent", color =NA),

plot.background = element\_rect(fill="transparent", color =NA)) +

theme(strip.background= element\_blank(), strip.text = element\_text(size=15),

panel.border = element\_rect(color="black", fill=NA))

#Plot Facility vs. Sections

ITSsectionstack <- ggplot(combined\_family\_ITS, aes(x = SampleOrder, y = Abundance, fill = Family)) + facet\_grid(Facility~Section) +

geom\_bar(stat = "identity") +

scale\_fill\_manual(values = Family\_colors) +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_blank(), axis.text.x=element\_blank(), axis.ticks.x=element\_blank(),

axis.title.y=element\_text(size=15)) +

guides(fill = guide\_legend(reverse = FALSE, keywidth = 1, keyheight = 1, ncol=1)) +

ylab("Relative Abundance") + theme(panel.background = element\_rect(fill="transparent", color =NA),

plot.background = element\_rect(fill="transparent", color =NA)) +

theme(strip.background= element\_blank(), strip.text = element\_text(size=15),

panel.border = element\_rect(color="black", fill=NA))

a\_ITS = plot\_grid(ITSfacilitystack + theme(legend.position= "none") , ITSsectionstack + theme(legend.position = "none") ,

ncol=1, nrow=2, labels=c("A", "B"), label\_size = 20)

b\_ITS = get\_legend(ITSfacilitystack)

c\_ITS = plot\_grid(a\_ITS, b\_ITS, ncol=3, rel\_widths = c(5,1))

c\_ITS

#Save and export the figure

ggsave("ITSstackcombined.pdf", plot=c, device="pdf", width=12, height=10, units="in", dpi=600)

#Making phyloseq object for rarefaction curve before normalization

phyloseq\_rare\_16s = phyloseq(otu\_table(otus.t, taxa\_are\_rows=FALSE), TAX, META)

phyloseq\_rare\_ITS = phyloseq(otu\_table(otus.t\_ITS, taxa\_are\_rows=FALSE), TAX\_ITS, META\_ITS)

#Rarefaction curves

rare\_16s\_apple\_plot <- ggrare(phyloseq\_rare\_16s, step = 100, se= TRUE, color="Facility")

rare\_16s\_byfacility\_plot <- rare\_16s\_apple\_plot + facet\_grid(Facility~.) +

theme(strip.text.y=element\_blank()) +xlab("Number of OTUs") + ylab("Number of unique OTUs") +

scale\_x\_continuous(breaks= seq(0,180000, 10000)) + theme(axis.text.x = element\_text(size=10, angle=90)) +

annotate("segment", x=-Inf, xend=Inf, y=-Inf, yend= -Inf) +

annotate("segment", x=-Inf, xend=Inf, y=-Inf, yend= -Inf) +

annotate("segment", x=-Inf, xend=Inf, y=-Inf, yend= -Inf)

rare\_ITS\_apple\_plot <- ggrare(phyloseq\_rare\_ITS, step = 100, se= TRUE, color="Facility")

rare\_ITS\_byfacility\_plot <- rare\_ITS\_applot\_plot + facet\_grid(Facility ~ .) +

theme(strip.text.y=element\_blank()) +xlab("Number of OTUs") + ylab("Number of unique OTUs") +

scale\_x\_continuous(breaks= seq(0,400000, 20000)) + theme(axis.text.x = element\_text(size=10, angle=90)) +

annotate("segment", x=-Inf, xend=Inf, y=-Inf, yend= -Inf) +

annotate("segment", x=-Inf, xend=Inf, y=-Inf, yend= -Inf) +

annotate("segment", x=-Inf, xend=Inf, y=-Inf, yend= -Inf)

rarefig <- plot\_grid(rare\_16s\_byfacility\_plot, rare\_ITS\_byfacility\_plot, nrow=1, ncol=2, labels=c("A", "B"), label\_size = 20)

ggsave("rarefig.pdf", plot=rarefig, device="pdf", width=11, height=6, units="in", dpi=600)

ggsave("rarefig.png", plot=rarefig, device="png", width=11, height=6, units="in", dpi=600)

#Alpha diversity

alpha <-estimate\_richness(phyloseq, measures=c("Shannon", "InvSimpson", "Chao1"))

estimate\_richness(phyloseq, split= TRUE, measures=c("Chao1", "Shannon", "InvSimpson"))

#Import 16S rRNA data

alpha\_16s <- read.csv(file.choose(), sep = ",", header = T, row.names = 1)

alpha\_ITS <- read.csv(file.choose(), sep = ",", header = T, row.names = 1)

#Pairwise.t.test for alpha diversity using Shannon and Inverse Simpson indices

pairwise.t.test(alpha\_16s$Shannon, alpha\_16s$Facility, p.adjust.method = "bonferroni")

pairwise.t.test(alpha\_16s$InvSimpson, alpha\_16s$Facility, p.adjust.method = "bonferroni")

#Import ITS data

alpha\_ITS <- read.csv(file.choose(), sep = ",", header = T, row.names = 1)

#Pairwise.t.test for alpha diversity using Shannon and Inverse Simpson indices

pairwise.t.test(alpha\_ITS$Shannon, alpha\_ITS$Facility, p.adjust.method = "bonferroni")

pairwise.t.test(alpha\_ITS$InvSimpson, alpha\_ITS$Facility, p.adjust.method = "bonferroni")

#Violin plots for alpha diversity

library(ggpubr)

#16S rRNA alpha diversity violin plots

alpha\_16s <- read.csv(file.choose(), sep = ",", header = T, row.names = 1)

alpha16s1 <- ggviolin(alpha\_16s, x = "Facility", y = "Shannon", add = "boxplot",

fill= "Facility" ) +

theme(axis.text.x = element\_text(size=13), axis.text.y = element\_text(size=13)) +

theme(axis.title = element\_text(size=15))

alpha16s2 <- ggviolin(alpha\_16s, x = "Facility", y = "InvSimpson", add = "boxplot",

fill = "Facility" ) +

theme(axis.text.x = element\_text(size=13), axis.text.y = element\_text(size=13)) +

theme(axis.title = element\_text(size=15))

#ITS alpha diversity violin plots

alpha\_ITS <- read.csv(file.choose(), sep = ",", header = T, row.names = 1)

alphaITS1 <- ggviolin(alpha\_ITS, x = "Facility", y = "Shannon", add = "boxplot",

fill= "Facility" ) +

theme(axis.text.x = element\_text(size=13), axis.text.y = element\_text(size=13)) +

theme(axis.title = element\_text(size=15))

alphaITS2 <- ggviolin(alpha\_ITS, x = "Facility", y = "InvSimpson", add = "boxplot",

fill = "Facility" ) +

theme(axis.text.x = element\_text(size=13), axis.text.y = element\_text(size=13)) +

theme(axis.title = element\_text(size=15))

#Combined 16S and ITS alpha diversity plots

a = plot\_grid(alpha16s1 + theme(legend.position= "none") ,

alpha16s2 + theme(legend.position = "none") ,

alphaITS1 + theme(legend.position = "none"),

alphaITS2 + theme(legend.position = "none"),

ncol=2, nrow=2, labels=c("A", "B","C","D"), label\_size = 20)

ggsave("alphadiversity.pdf", plot =a, device="pdf", width=12, height=10, units="in",dpi=600)

ggsave("alphadiversity.png", plot =a, device="png", width=12, height=10, units="in",dpi=600)

#Pairwise PERMANOVA

library(devtools)

install\_github("pmartinezarbizu/pairwiseAdonis/pairwiseAdonis")

library(pairwiseAdonis)

#Run pairwise PERMANOVA for 16S rRNA data

permanova\_data\_16s <- data.frame(sample\_data(phyloseq))

pairwise\_perm\_16s\_f <- pairwise.adonis(otu\_table(phyloseq), permanova\_data\_16s$Facility)

pairwise\_perm\_16s\_s <- pairwise.adonis(otu\_table(phyloseq), permanova\_data\_16s$Section)

#Export the file in .csv

write.csv(pairwise\_perm\_16s\_f, "microbiome\_pairwise\_Facility.csv")

write.csv(pairwise\_perm\_16s\_s, "microbiome\_pairwise\_Section.csv")

#Run pairwise PERMANOVA based on ITS data

permanova\_data\_ITS <- data.frame(sample\_data(phyloseq))

pairwise\_perm\_ITS\_f <- pairwise.adonis(otu\_table(phyloseq), permanova\_data\_ITS$Facility)

pairwise\_perm\_ITS\_s <- pairwise.adonis(otu\_table(phyloseq), permanova\_data\_ITS$Section)

#Export the file in .csv

write.csv(pairwise\_perm\_ITS\_f, "mycobiome\_pairwise\_Facility.csv")

write.csv(pairwise\_perm\_ITS\_s, "mycobiome\_pairwise\_Section.csv")

#PICRUSt analysis plot

#Import csv file for picrust, actural abundance

picrustfuntion <- read.csv(file.choose(), sep = ",", header = T)

#Make boxplots based on PICRUSt data

allfunctionabun <- ggplot(picrustfuntion, aes(x=Facility, y=Abundance, fill=Pathway.category)) +

theme\_bw() +geom\_boxplot()+

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_text(size=15),

axis.title.y=element\_text(size=15))

#Make a plot based on PICRUSt function abundance

functionabundance <- ggplot(picrustfuntion, aes(x=Facility, y=Abundance, fill=Pathway.category)) +theme\_bw() +geom\_boxplot()

#Import .csv file for PICRUSt, relative abundance, all combined

picrustrelabun <- read.csv(file.choose(), sep = ",", header = T)

refunctionabun <- ggplot(picrustrelabun, aes(x=Facility, y=relative.abundance, fill=Category)) +theme\_bw() +

geom\_boxplot() +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_text(size=15),

axis.title.y=element\_text(size=15))

#Create a plot for functional categories based on relative abundance

re\_all\_plot <-ggplot(picrustrelabun, aes(x=Facility, y=relative.abundance, fill=Category)) +theme\_bw() + geom\_boxplot()

#Import csv file for PICRUSt, relative abundance, by category

picrustfuntioncate <- read.csv(file.choose(), sep = ",", header = T)

#Pairwise.t.test for significant difference between categories

pairwise.t.test(picrustfuntioncate$Metabolism, picrustfuntioncate$Facility, p.adjust.method = "bonferroni")

ggplot(picrustfuntioncate, aes(x=Facility, y=Cellular.Processes)) + theme\_bw() + geom\_col()

#Plot metabolism and environment functional categories for each facility

metabolism <- ggplot(picrustfuntioncate, aes(x=Facility, y=Metabolism)) + theme\_bw() +

geom\_boxplot() +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_text(size=15),

axis.title.y=element\_text(size=15))

Environment <- ggplot(picrustfuntioncate, aes(x=Facility, y=Environmental.Information.Processing)) +

theme\_bw() + geom\_boxplot() +

theme(legend.text=element\_text(size=13), legend.title= element\_text(size=15)) +

theme(axis.title.x = element\_text(size=15),

axis.title.y=element\_text(size=15))

a = plot\_grid(allfunctionabun, refunctionabun, nrow=2, labels=c("A", "B"), label\_size = 20)

b =plot\_grid(metabolism, Environment,nrow = 2, labels = c("C","D"), label\_size = 20)

c = plot\_grid(a, b, ncol=2, rel\_widths = c(6,3))

c

ggsave("picrust.pdf", plot=c, device="pdf", width=12, height=10, units="in", dpi=600)

ggsave("picrust.png", plot=c, device="png", width=12, height=10, units="in", dpi=600)