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Supplemental Information

Impaired host response and the presence

of Acinetobacter baumannii in the serum

microbiome of type-II diabetic patients

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Supplemental Information

Supplemental Figures and Legends

 Figure S1. Mock library composition evaluation. Related to Figure 1. A Zymogen mock community library was used for sequence library construction at 10 and 1 ng in duplicate between both sequencing runs. QIIME2 / DADA2 unfiltered data was given taxonomical assignment via SILVA and assessed via the "q2-quality-control" plugin. (A) Taxon accuracy rate 8 (TAR), taxon detection rate (TDR) and linear regression scores (r-squared) are plotted at each
9 taxonomic level. (B) Expected vs observed abundance of each species in the mock community taxonomic level. (B) Expected vs observed abundance of each species in the mock community is plotted at each taxonomic level. (C) Distance between false positive observations vs nearest expected feature. Misclassifications were unique to environmental contaminant taxa and no false negative features were observed (not shown).

 Figure S2. Alpha-diversity at different rarefaction levels. Related to Figure 1. Total observed OTUs (here ASVs) are plotted (A) vs sequencing depth and each sample that can accommodate that sequencing depth is plotted in (B).

 Figure S3. Run to Run variability quantification. Related to Figure 1. A Run vs Run Beta diversity comparison was performed for unfiltered data by comparing unweighted Unifrac distances as described in the methods. Pairwise PERMANOVA of Run-1 vs Run-2 (group size of 2, sample size of n=134 total indexes) was performed in 999 permutations resulting in a pseudo-F value of 0.567, p-value of 0.999 and corrected q-value of 0.999. No significant differences were observed between each run.

 Figure S4. 16S Alpha and Beta diversity of serum microbiomes changes with subject status. Related to Figure 1. Species richness is displayed based on Faith's phylogenetic diversity (Faith, 1992) (A) and species evenness based on Pielou's evenness index (Pielou, 1966) (B) for healthy, type-II diabetic (T2D) and type-II diabetics with periodontitis (T2DP) 32 samples. Each graph used minimally filtered data for analysis. Significance was determined by
33 Kruskal-Wallis analysis of variance (Kruskal and Wallis, 1952) for each comparison indicated Kruskal-Wallis analysis of variance (Kruskal and Wallis, 1952) for each comparison indicated and Benjamini-Hochberg correction (Benjamini and Hochberg, 1995) was applied to generate adjusted q-values. * indicates q-value < 2e-3 vs healthy, ** indicates q-value < 0.024 vs healthy, (C) Beta-diversity Bray Curtis distances. Pairwise PERMANOVA of each category vs each (group size of 3, n=112) was performed in 999 permutations. *** indicates q-value differs 38 from healthy < 0.001 .

Figure S5. LEfSe Comparisons reveal taxonomical differences between subject groups.

 Related to Figure 3. (A) LEfSe Cladogram plots (Segata *et al.*, 2011) reveal general taxonomic shifts compared to (B) specifically enriched genera or species in all diabetics vs healthy and (C) compared across all 3 subject groups. Data was generated from DADA2 / QIIME2 analyzed data aligned to the SILVA database that was then filtered for contaminating taxa abundant in no-template control indexes. *Pseudomonas* genus / species level results were also removed for clarity.

 Figure S6. Cytokine correlation with specific taxa in healthy subjects. Related to Figure 7. HOMD/RDP assigned taxonomy of MED analyzed 16S data for healthy samples and cytokine concentrations were analyzed via Pearson correlation coefficient in R using the *rcorr* function. Significance was determined using the asymptotic p-values generated by *rcorr* with * = p-value <0.05 ** = p-value <0.01. Data are strictly filtered with taxa present in no-template controls subtracted.

 Figure S7. 16S Alpha and Beta diversity of serum microbiomes changes with subject status. Related to Figure 1. Alpha diversity is displayed based on Shannon's index (Shannon and Weaver, 1975) (A) and species evenness based on Simpson's evenness measure (Simpson, 1949)) (B) for healthy, type-II diabetic (T2D) and type-II diabetics with periodontitis (T2DP) samples. Each graph used minimally filtered data for analysis. Significance was determined by Kruskal-Wallis analysis of variance (Kruskal and Wallis, 1952) for each comparison indicated and Benjamini-Hochberg correction (Benjamini and Hochberg, 1995) was applied to generate adjusted q-values. * indicates q-value < 1e-5, ** indicates q-value < 0.007, (C) Beta-diversity unweighted Unifrac distances. Pairwise PERMANOVA of each category vs each (group size of 3, n=112) was performed in 999 permutations. *** indicates q-value differs 69 from 'healthy' < 0.02 .

73 **Figure S8. Host-microbial correlation resolved type-II diabetic associations with**

74 **inflammatory cytokines.** Related to Figure 7. SILVA assigned taxonomy of DADA2 analyzed

75 16S data for both T2D and T2DP samples and cytokine concentrations were analyzed via
76 Pearson correlation coefficient in R using the rcorr function. Significance was determined u

Pearson correlation coefficient in R using the *rcorr* function. Significance was determined using

77 the asymptotic p-values generated by *rcorr* with $* = p$ -value <0.05 ** = p-value <0.01. Data are 78 strictly filtered with taxa present in no-template controls subtracted.

 Figure S9. Cytokine correlation with specific taxa in healthy subjects. Related to Figure 7. SILVA assigned taxonomy of DADA2 analyzed 16S data for healthy samples and cytokine 82 concentrations were analyzed via Pearson correlation coefficient in R using the *rcorr* function.
83 Significance was determined using the asymptotic p-values generated by *rcorr* with $* = p$ -value Significance was determined using the asymptotic p-values generated by *rcorr* with * = p-value <0.05 ** = p-value <0.01. Data are strictly filtered with taxa present in no-template controls subtracted.

Transparent Methods

Subject recruitment, sampling and storage

 Subject recruitment has been described previously (Freire *et al.*, 2017). Peripheral venous blood (∼60 ml) was collected from patients diagnosed with T2D and from healthy nondiabetic controls. Blood samples were collected and centrifuged at 2300 rpm, and serum was isolated and frozen 93 at -80°C until analysis under IRB protocol #13-07. All subjects gave signed informed consent prior to study evaluations. Clinical periodontal data and peripheral venous blood were collected. The diagnosis of T2D was made by the subject's primary care physician following American Association of Diabetes guidelines (American Diabetes Association, 2015). Information was collected on subject demographics (age, gender, self-reported ethnicity, and self-reported smoking status), body-mass index (BMI; kg/m2), blood total cholesterol, blood glucose (point-of- care), percent hemoglobin A1C (HbA1c), and periodontal condition (Armitage, 1999). HbA1c was used to determine the level of glycemic control for diabetic subjects. One T2D individual lacked HbA1c measurements but fit based on all other diagnostic criteria (blood glucose >200 mg/dl) as well as cytokine profiles in accord with other T2D individuals. Neutrophil and monocyte cell counts were determined by lab assay (described below). Individuals were excluded if they were taking insulin sensitizers, nonsteroidal anti-inflammatory drugs, or antimicrobials within 3 months of study initiation. Smoking status was defined by CDC NHIS terms

(https://www.cdc.gov/nchs/nhis/tobacco/tobacco_glossary.htm#:~:text=Former%20smoker%3A

 %20An%20adult%20who,in%20his%20or%20her%20lifetime). Site collection before blood 109 draw: Samples were drawn in clinical phlebotomy settings. From all serum sampled for this 110 microbiome study, a total of 81 subjects (N=24 healthy, N=57 T2D) were included for analysis, microbiome study, a total of 81 subjects ($N=24$ healthy, $N=57$ T2D) were included for analysis, all of whom were unrelated and over 18 years of age (range: 28-79 years of age).

- 113 Sample preparation for 16S analysis
- *DNA Extraction*
-

 Frozen serum from N=81 samples (N=24 healthy, N=32 T2D and N=25 T2DP) were thawed on ice and aliquots were separated for DNA extraction. DNA was extracted using the Epicentre MasterPure Complete DNA and RNA Purification kit (Lucigen, WI, USA). Extraction was carried out using the manufacturer's instructions, with modifications that enabled bead beating. Briefly, 200µL of Tissue and Cell Lysis Solution (2x), 100µL of nuclease free water and 2µL of Proteinase K were added to Lysing matrix B (LMB) tubes (MP Biomedicals, Santa Ana, CA). 123 Thereafter 100µL of serum was added to the tubes and placed in a Beadbeater (Biospec) for 30 seconds, then placed on ice for 3 minutes and then repeated. Each round of DNA extractions included a no template control, which consisted of 100µL nuclease free water instead of serum. The samples were then incubated and extracted as described in the manufacturer's instructions. Total DNA was precipitated using the manufacturer's instructions, however 300 µL of MPC Protein Precipitation Reagent was used to accommodate the increased volume. DNA was eluted in 50µL of nuclease free water.

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- 16s rDNA primers 27F

(GTCTCGTGGGCTCGGAGATGTGTATAAGAGACAGGWATTACCGCGGCKGCTG)

(Stackebrandt and Goodfellow, 1991; Turner *et al.*, 1999) were used to amplify the V1-V3

regions of 16s rDNA in a 50 µL reaction using 2x Q5 HiFi mastermix (New England Biolabs,

136 Ipswitch, MA) and 23 µL of extracted DNA (35 cycles).

-
- *Illumina MiSeq Library preparation*

139
140 After the first round of PCR was carried out it was cleaned with Ampure XP beads (Beckman Coulter, Pasadena, CA) and then visualized by agarose gel electrophoresis. Full indices and adapters were added using the Illumina Nextera Index Kit (Illumina, San Diego, CA) by running the second round of PCR (50 ng of template DNA, 5 cycles) in 2x Phusion HF Master Mix (New England Biolabs, Ipswitch, MA). PCR were then cleaned with Ampure XP beads and analyzed by agarose gel electrophoresis and using the Agilent BioAnalyzer DNA1000 chip (Agilent Technologies, Santa Clara, CA). Samples were then quantified and normalized prior to pooling using a Qubit fluorimeter (Invitrogen, Carlsbad, CA). The final pooled library was quantified with the KAPA Biosystems Illumina Kit (KAPA Biosystems, Woburn, MA) via qPCR in a Roche LightCycler480. Samples were sequenced (2x250 bp paired-end) on an Illumina MiSeq (Illumina, San Diego, CA) at the Rhode Island Genomics and Sequencing Center (Kingston, RI).

Mock Community Standards Preparation

 In order to quantify potential bias due to low template concentrations, PCR amplification, and contamination, we performed mock library assemblies with commercially available bacterial genomic DNA templates (Zymo Research #D6305). We amplified 2 libraries using 10 and 1 ng

total starting templates in both sequencing runs for a total of 4 mock libraries synthesized.Total

DNA species composition is provided in the manufacturer's instructions.

16S clone library resequencing

 Primers were made to amplify regions of 16s rDNA that enable differentiation of the Acinetobacter species. Primers used were

oMR328 (TAGCGGCGGACGGGTGAGTAATGCTTA) and

 oMR329 (TTCCGACTTCATGGAGTCGAGTTGCAGAC). 50 µL reactions in 2x Q5 mastermix (New England Biolabs, Ipswitch, MA) were run for 35 cycles using 23 µL of DNA extracted from serum as the template. The products were then purified using the Qiagen PCR purification kit according to the manufacturer's instructions. Gotaq 2x mastermix (Promega, Madison, WI) was used to generate 'A' overhangs via PCR for 5 cycles using 10 µL of purified DNA as the template. 4µL of the resulting PCR products were then used to generate clones using the TOPO TA Cloning Kit (Invitrogen, Carlsbad, CA) according to the manufacturer's instructions which was modified to use 0.5µL of TOPO vector. Clones were then transformed into NEB5α cells using the manufacturer's instructions and plated on LB plates supplemented with 0.1mM IPTG and 40µg/mL Kanamycin. Additionally, 40µl of 40mg/mL of X-gal was spread onto each plate 173 prior to plating to enable blue-white screening. The plates were incubated for 24 hours at 37°C. 174 White colonies were inoculated into liquid cultures of LB supplemented with 40µg/mL kanamycin. Plasmids were extracted using QIAprep Spin Miniprep kit (Qiagen, Venlo, Netherlands). Inserts were sequenced using the M13 Forward (GTAAAACGACGGCCAGTG) and M13 Reverse (CACAGGAAACAGCTATGACC) primers on an Applied Biosystems 3500xl using the "BigDye" Terminator v3.1 Cycle sequencing kit. Sequences were analyzed using 179 NCBI BLAST (NCBI Resource Coordinators, 2018). A maximum of 10 colonies were screened
180 for each sample or until a sequence matched a species belonging to the Acinetobacter for each sample or until a sequence matched a species belonging to the *Acinetobacter baumanii- calcoaceticus* complex.

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- Cytokine Quantification
-

 Frozen (-80°C) subject serum was brought up to room temperature and assayed using the Invitrogen human inflammation 20-plex ProcartaPlex cytokine panel (Thermo Fisher Scientific, Waltham, MA) on a Luminex 200 instrument (Luminex, Austin, TX) in universal assay buffer. Assayed cytokines included: MIP-1α, IL-1β, IL-4, IP-10, IL-6, IL-8, IL-10, IL-12p70, IL-13, IL- 17A, IFN-γ, GM-CSF, TNF-α, MIP-1β, IFN-α, MCP-1, P-Selectin, IL-1α, sICAM-1, and E- Selectin. Following manufacturer protocols, all samples were run on a plate with 7 standards (diluted 1:4) and a control (universal assay buffer only), with all samples, standards, and controls run in duplicate similar to our methods previously (Pessoa *et al.*, 2019; Kleinstein *et al.*, 2020) .

 Quality control (QC) steps were conducted according to manufacturer recommendations by xPONENT 4.2 software (Affymetrix eBioscience, San Diego, USA). Any standards with <70 or >130 % recovery of beads were invalidated. Samples were required to have a bead count of >30 beads recovered (all samples had >100 beads recovered and none were excluded at this step). Following QC, results were reported as average pg/mL for all measured cytokines. The lower limit of quantification (LLOQ) was determined based on the standard curve (after QC) as the average value of the lowest validated standards. Values at or below the LLOQ for each cytokine were reported at the LLOQ. As samples were run in two batches (with similar LLOQs for each batch), LLOQ for the cytokines (in pg/mL) were averaged across both runs: MIP- 1α=1.79, IL-1β=5.58, IL-4=23.99, IP-10= 1.17, IL-6=18.89, IL-8=2.48, IL-10=2.01, IL- 12p70=11.72, IL-13=5.63, IL-17A=4.63, IFN-γ=11.93, GM-CSF=14.57, TNF-α=9.66, MIP- 1β=7.66, IFN-α=1.32, MCP-1=3.73, P-Selectin=1077.15, IL-1α=0.74, sICAM-1=442.80, and E- Selectin=441.00. To compare cytokine levels between study groups, unpaired t-tests were used with a significance threshold of p<0.05 and no assumption of consistent standard deviation.

- Minimum Entropy Decomposition analysis of sequence libraries
-

 Poor quality reads were discarded using flexbar (Dodt *et al.*, 2012). Reads were then treated as single end reads and trimmed for quality and to a length of 160bp using Trimmomatic (Bolger, Lohse and Usadel, 2014). Minimum Entropy Decomposition (MED) was performed as described in the oligotyping pipeline (Eren *et al.*, 2015), generating a list of node representative sequences and their relative abundances. Taxonomic assignment of representative sequences were generated based on aligning 16s rRNA sequences at 98.5% identity against the eHOMD database (Escapa *et al.*, 2018) and further evaluated using the Ribosomal Database Project (RDP) database (Cole *et al.*, 2014). MED nodes were assigned species level taxonomy based 220 on $>98.5\%$ identity matches to eHOMD and/or RDP with eHOMD designations used as 1st priority as our intial aim was to identify oral taxa. Node frequency tables now assigned taxonomy were then used in association with sample metadata for Linear Discriminate Effect Size analysis (LEfSe) (Segata *et al.*, 2011).

- QIIME2 analysis of sequence libraries
-
- *All Scripts at end of this document. All work was performed on the University of Rhode Island High Performance Computing Bluewaves cluster:
- <https://web.uri.edu/hpc-research-computing/cluster-specifications/>
- *Data Import*
- Fastq sequence data were imported using the **import-fastq.sh** script. Generated demux.qzv 231 files were viewed using https://view.giime2.org/ via the "quality viewer" function to determine location for sequence quality trimming used in the following step.
- *Fastq Trimming, DADA2 Analysis, Run-run Merging*

 A second sequencing run of the prepared Illumina libraries was performed to ensure maximum sequencing depth without further PCR amplification of starting template. At the end of the "D2- merge.sh" script we then used the **Rarefaction-alpha.sh** script output (Fig. S2) to determine read depth (22,000) to use in downstream commands.

Initial Alpha and Beta Diversity Measurements and Run-run Comparison Testing

Merged libraries were used in the **Alpha-Beta-An.sh** script to determine initial phylogenetics on

unfiltered data using the QIIME2 *core-metrics-phylogenetic* function, Alpha and Beta diversity

calculations via QIIME2 *alpha-* or *beta-group-significance* functions as well as Beta diversity

difference calculation (*beta-group-significance*) between sequencing Run 1 vs. Run 2 (Figs. S3,

S4).

Mock Community Standards Quality Control Testing

- This section uses the **mocklibQC1.sh** and **mocklibQC2.sh** scripts. Taxonomy files used are
- 246 described below. The *-bar-plots.qzv file generated in the $1st$ script was viewed at
- 247 https://view.giime2.org/ and the level 6 data from this plot was exported as a .csv file which
- provides the reads per each taxa. This data was converted to proportion of total reads per each
- 249 taxa and converted into the Zymo_actual.tsv file. The Zymo_expected.tsv file was generated by
- editing the Zymo_actual.tsv file with the proportion for each taxa present as provided in the manufacturers reference:
- [https://files.zymoresearch.com/protocols/_d6305_d6306_zymobiomics_microbial_community_d](https://files.zymoresearch.com/protocols/_d6305_d6306_zymobiomics_microbial_community_dna_standard.pdf) [na_standard.pdf](https://files.zymoresearch.com/protocols/_d6305_d6306_zymobiomics_microbial_community_dna_standard.pdf)
- These .tsv inputs were then used in the **mocklibQC2.sh** script to output the visualizations seen in Figure S1.
- *Classifier Setup and Editing*
- Initial import of the SILVA 132 classifier was performed by running the following command:
- *wget https://data.qiime2.org/2019.7/common/silva-132-99-nb-classifier.qza*

Classifier setup was then performed using the **classifier.sh** script. Upon searching the silva-

taxonomy.qzv file generated, we noted that there were no reads that matched *A. baumanii*

sequences at all. We used the rep-seqs.merge.qzv file output from the **D2merge.sh** script

saved as a FASTA file from within the QIIME2 viewer function and performed a pairwise blast vs

the full length *A. baumanii* 16S sequence from NCBI:

>A.baumanii 16s

 AACGCTGGCGGCAGGCTTAACACATGCAAGTCGAGCGGGGGAAGGTAGCTTGCTACCGG ACCTAGCGGCGGACGGGTGAGTAATGCTTAGGAATCTGCCTATTAGTGGGGGACAACATC TCGAAAGGGATGCTAATACCGCATACGTCCTACGGGAGAAAGCAGGGGATCTTCGGACCT TGCGCTAATAGATGAGCCTAAGTCGGATTAGCTAGTTGGTGGGGTAAAGGCCTACCAAGG CGACGATCTGTAGCGGGTCTGAGAGGATGATCCGCCACACTGGGACTGAGACACGGCCC

 AGACTCCTACGGGAGGCAGCAGTGGGGAATATTGGACAATGGGGGGAACCCTGATCCAG CCATGCCGCGTGTGTGAAGAAGGCCTTATGGTTGTAAAGCACTTTAAGCGAGGAGGAGGC TACTTTAGTTAATACCTAGAGATAGTGGACGTTACTCGCAGAATAAGCACCGGCTAACTCT GTGCCAGCAGCCGCGGTAATACAGAGGGTGCGAGCGTTAATCGGATTTACTGGGCGTAAA GCGTGCGTAGGCGGCTTATTAAGTCGGATGTGAAATCCCCGAGCTTAACTTGGGAATTGC ATTCGATACTGGTGAGCTAGAGTATGGGAGAGGATGGTAGAATTCCAGGTGTAGCGGTGA AATGCGTAGAGATCTGGAGGAATACCGATGGCGAAGGCAGCCATCTGGCCTAATACTGAC GCTGAGGTACGAAAGCATGGGGAGCAAACAGGATTAGATACCCTGGTAGTCCATGCCGTA AACGATGTCTACTAGCCGTTGGGGCCTTTGAGGCTTTAGTGGCGCAGCTAACGCGATAAG TAGACCGCCTGGGGAGTACGGTCGCAAGACTAAAACTCAAATGAATTGACGGGGGCCCGC ACAAGCGGTGGAGCATGTGGTTTAATTCGATGCAACGCGAAGAACCTTACCTGGCCTTGA CATACTAGAAACTTTCCAGAGATGGATTGGTGCCTTCGGGAATCTAGATACAGGTGCTGCA TGGCTGTCGTCAGCTCGTGTCGTGAGATGTTGGGTTAAGTCCCGCAACGAGCGCAACCCT TTTCCTTACTTGCCAGCATTTCGGATGGGAACTTTAAGGATACTGCCAGTGACAAACTGGA GGAAGGCGGGGACGACGTCAAGTCATCATGGCCCTTACGGCCAGGGCTACACACGTGCT ACAATGGTCGGTACAAAGGGTTGCTACACAGCGATGTGATGCTAATCTCAAAAAGCCGATC GTAGTCCGGATTGGAGTCTGCAACTCGACTCCATGAAGTCGGAATCGCTAGTAATCGCGG ATCAGAATGCCGCGGTGAATACGTTCCCGGGCCTTGTACACACCGCCCGTCACACCATGG GAGTTTGTTGCACCAGAAGTAGCTAGCCTAACTGCAAAGAGGGCGGTTACCACGGTGTGG CCGATGACTGGGGTGAAGT

 One ASV from the rep-seqs.merge.qzv file (e875355f3179838110485d8f5013d4a6) returned a 100% match homology to the above *A. baumanii* sequence. This ASV was just annotated as *Acinetobacter* in the SILVA-132-99 taxonomy. We then edited the silva classifier taxonomy file by first exporting it using the following commands:

- *module load QIIME2/2019.7*
- *qiime tools export *
- *--input-path silva-taxonomy.qza *
- *--output-path silva-taxonomy*

 This output the file in *.tsv format which was then opened in a text editor and searched for the ASV node identity and then modified as indicated here:

before

- e875355f3179838110485d8f5013d4a6
- D_0__Bacteria;D_1__Proteobacteria;D_2__Gammaproteobacteria;D_3__Pseudomonadales;D
- _4__Moraxellaceae;D_5__Acinetobacter 0.9999662516375396
- **after**
- e875355f3179838110485d8f5013d4a6
- D_0__Bacteria;D_1__Proteobacteria;D_2__Gammaproteobacteria;D_3__Pseudomonadales;D
- _4__Moraxellaceae;D_5__Acinetobacter;D_6__Acinetobacter baumanii
- 0.9999662516375396
- The edited file was then saved as taxonomy-1.tsv and imported via the following command:
- *qiime tools import *
- *--input-path /data/mramseylab/classifiers/silva-taxonomy/taxonomy-1.tsv *
- *--type 'FeatureData[Taxonomy]' *
- *--input-format TSVTaxonomyFormat *
- *--output-path silva-mod-taxonomy.qza*
- *qiime metadata tabulate *
- *--m-input-file silva-mod-taxonomy.qza *
- *--o-visualization silva-mod-taxonomy.qzv*
- The silva-mod-taxonomy.qza file was used for the rest of our analyses.
- *Data Filtering*

 Data presented throughout the manuscript is generally described as "Minimally Filtered" or 321 "Strict Filtered". We $1st$ began with all ASV assigned data from the above scripts and used metadata based filtering to separate out all data from control samples which included no- template and PCR only control indexes. This was done using the **control-filter.sh** script. Data output from the script was exported so it could be viewed and then used to subtract taxa from human-derived datasets in downstream filtering steps. Next we applied the **meta-filter.sh** script to our data to extract only human derived samples for further analysis and filtering. After this we applied the **minimal-filter.sh** script. Outputs from this script were used in all "Minimally Filtered" 328 described data in the manuscript. Minimal filtering included removal of instances of taxa that
329 appeared in only 2 samples or less, features (ASV sequences) that appeared less than 10 tin appeared in only 2 samples or less, features (ASV sequences) that appeared less than 10 times across all samples and any taxa with less than 20 reads per sample.

 Next we took our output files from the above **control-filter.sh** script and used them as part of the input for the **auto-filter.sh** script to exclude these taxa from the remaining minimally filtered data. While this removed many spurious taxa from our samples we observed high abundance of known aquatic microbial contaminant sequences not typically human associated (ex: *Sphingomonas, Ralstonia*). These and other taxa were manually excluded using the **strict- filter.sh** script and further removal of highly abundant *Pseudomonas* sequences were also removed via the **nopa-filter.sh** script. Data filtered to this extent are referred to as "Strictly" filtered in the manuscript.

Re-analysis of Filtered Data

340 A reanalysis of filtered data was performed $1st$ for Alpha and Beta diversity measurements via the **alpha-beta2.sh** script (Fig. 1). Further description and statistical analysis of Beta diversity differences were performed in R using the *vegan* package via the metaMDS and Adonis functions primarily as demonstrated in the **MMR20_ellipses_NMDS.R** script below.

LEfSe Analysis Comparison

 Initial LEfSe analysis was performed using ASV output abundance data directly from the initial MED pipeline (Fig. 3). Further analysis on QIIME2 / DADA2 assigned strictly filtered data (Fig. S5) was performed first using the **lefse-noTax.sh** script to export and format data for analysis on the LEfSe Galaxy server [\(https://huttenhower.sph.harvard.edu/galaxy/\)](https://huttenhower.sph.harvard.edu/galaxy/). This script only outputs ASV node information without taxonomic assignment. Manual taxonomic assignment was performed comparing ASV node names to the rep-seqs-merge.qza file from the

- **D2merge.sh** script. LEfSe analysis was performed using default settings, *.svg output files were
- further edited in InkScape software for clarity.

Supplemental References

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400 Supplemental Data (Data S1.) Related to Transparent Methods
401 Scripts used for analyses:
402
403 import-fastq.sh
404 module load QIIME2/2019.7
405
406 for d in /data/mramseylab/raw reads/2018 Serum/Run* ;
407 do
408 SUFX=${d#*Run}
409
410 qiime tools import \
411 --type 'SampleData[PairedEndSequencesWithQuality]' \
412 --input-path /data/mramseylab/raw reads/2018 Serum/Run$SUFX \
413 --input-format CasavaOneEightSingleLanePerSampleDirFmt \
414 --output-path /data/mramseylab/raw_reads/2018_Serum/Run$SUFX/demux-
415 $SUFX.qza
416
417 qiime demux summarize \
418 --i-data /data/mramseylab/raw_reads/2018_Serum/Run$SUFX/demux-
419 $SUFX.qza \setminus420 --o-visualization /data/mramseylab/visualizations/demux-$SUFX.qzv
421
422 done
423
```


```
456 --o-merged-table 
457 /data/mramseylab/raw reads/2018 Serum/Merge runs/denoise-table-
458 merge.qza
459
460 qiime feature-table merge-seqs \
461 --i-data /data/mramseylab/raw reads/2018 Serum/Run1/rep-seqs-1.qza \
462 --i-data /data/mramseylab/raw reads/2018 Serum/Run2/rep-seqs-2.qza \
463 --o-merged-data 
464 /data/mramseylab/raw_reads/2018_Serum/Merge_runs/rep-seqs-merge.qza
465
466 qiime feature-table summarize \
467 --i-table /data/mramseylab/raw_reads/2018_Serum/Merge_runs/denoise-
468 table-merge.qza \
469 --o-visualization /data/mramseylab/visualizations/denoise-table-
470 merge.gzv \
471 --m-sample-metadata-file /data/mramseylab/metadata/Serum5.tsv
472
473 qiime feature-table tabulate-seqs \
474 --i-data /data/mramseylab/raw reads/2018 Serum/Merge runs/rep-seqs-
475 merge.qza \
476 --o-visualization /data/mramseylab/visualizations/rep-seqs-merge.qzv
477
478 qiime phylogeny align-to-tree-mafft-fasttree \
479 --i-sequences /data/mramseylab/raw reads/2018 Serum/Merge runs/rep-
480 seqs-merge.qza \
481 --o-alignment 
482 /data/mramseylab/raw_reads/2018_Serum/Merge_runs/aligned-rep-seqs-
483 merge.qza \
484 --o-masked-alignment 
485 /data/mramseylab/raw_reads/2018_Serum/Merge_runs/masked-aligned-rep-
486 seqs-merge.qza \
487 --o-tree /data/mramseylab/raw_reads/2018_Serum/Merge_runs/unrooted-
488 tree-merge.qza \
489 --o-rooted-tree 
490 /data/mramseylab/raw reads/2018 Serum/Merge runs/rooted-tree-merge.qza
```


```
511 Alpha-Beta-An.sh
512 module load QIIME2/2019.7
513
514 rawdir=/data/mramseylab/raw_reads/2018_Serum/Merge_runs/
515 procdir=/data/mramseylab/proc_reads/
516 metadir=/data/mramseylab/metadata/
517 cmr="core-metrics-results"
518
519 qiime diversity core-metrics-phylogenetic \
520 --i-phylogeny $rawdir\rooted-tree-merge.qza \
521 --i-table $rawdir\denoise-table-merge.qza \
522 --p-sampling-depth 22000 \
523 --m-metadata-file $metadir\Serum5.tsv \
524 --output-dir $procdir$cmr\-initial
525
526 qiime diversity alpha-group-significance \
527 --i-alpha-diversity $procdir$cmr\-initial/faith pd vector.qza \
528 --m-metadata-file $metadir\Serum5.tsv \
529 --o-visualization $procdir$cmr\-initial/faith-pd-group-
530 significance.qzv
531
532 qiime diversity alpha-group-significance \
533 --i-alpha-diversity $procdir$cmr\-initial/evenness vector.qza \
534 --m-metadata-file $metadir\Serum5.tsv \
535 --o-visualization $procdir$cmr\-initial/evenness-group-
536 significance.qzv
537
538 array=( unweighted unifrac distance matrix
539 weighted unifrac distance matrix bray curtis distance matrix )
540
```

```
541 for i in "${array[@]}"
542 do
543
544 qiime diversity beta-group-significance \
545 --i-distance-matrix $procdir$cmr\-initial/$i.qza \
546 --m-metadata-file $metadir\Serum5.tsv \
547 --m-metadata-column Condition \
548 --o-visualization $procdir$cmr\-initial/$i.qzv \
549 --p-pairwise
550
551 done
552
553 #Generates data for run1 vs run2 variability to measure batch effect
554 qiime diversity beta-group-significance \
555 --i-distance-matrix $procdir$cmr\-
556 initial/unweighted unifrac distance matrix.qza \
557 --m-metadata-file $metadir\Serum4.tsv \
558 --m-metadata-column Run \
559 --o-visualization $procdir$cmr\-initial/unweighted-unifrac-Run-
560 significance.qzv \
561 --p-pairwise
```


```
583 Control-filter.sh
584 module load QIIME2/2019.7
585
586 tablein=/data/mramseylab/raw_reads/2018_Serum/Merge_runs/denoise-
587 table-merge.qza
588 clsdir=/data/mramseylab/classifiers/
589 metadir=/data/mramseylab/metadata/
590 visdir=/data/mramseylab/visualizations/
591 filtdir=/data/mramseylab/proc_reads/
592 # filter status of input files, "ctrl-filter" is just for taxa 
593 belonging to no template controls
594 fil=ctrl-filter
595
596
597 #must make the directory you are filtering to 1st or else it will 
598 error
599 mkdir $filtdir$fil
600
601
602 qiime feature-table filter-samples \
603 --i-table $tablein \
604 --m-metadata-file $metadir\Serum5.tsv \
605 --p-where "[Source]='Control'" \
606 --o-filtered-table $filtdir$fil/$fil-table.qza
607
608 qiime taxa collapse \
609 --i-table $filtdir$fil/$fil-table.qza \
610 --i-taxonomy $clsdir\silva-mod-taxonomy.qza \
611 --p-level 6 \setminus612 --o-collapsed-table $filtdir$fil/$fil-collapse-table.qza
```

```
613
614 qiime feature-table relative-frequency \
615 --i-table $filtdir$fil/$fil-collapse-table.qza \
616 --o-relative-frequency-table $filtdir$fil/$fil-relative-collapse-
617 table.qza
618
619 qiime tools export \
620 --input-path $filtdir$fil/$fil-relative-collapse-table.qza \
621 --output-path $filtdir$fil/
622
623 biom convert \
624 -i $filtdir$fil/feature-table.biom \
625 -o $filtdir$fil/$fil-relative-collapse-table.txt \
626 --header-key "taxonomy" \
627 --to-tsv
628
629 #Use above taxa table to filter out based on taxa present in controls
630
631 qiime feature-table filter-features \
632 --i-table $tablein \
633 --m-metadata-file $filtdir$fil/$fil-collapse-table.qza \
634 --o-filtered-table $filtdir$fil/$fil-excluded-table.qza \
635 --p-exclude-ids
636
637 #Use excluded table to generate barplot for checking
638
639 qiime taxa barplot \
640 --i-table $filtdir$fil/$fil-excluded-table.qza
641 --i-taxonomy $clsdir\silva-mod-taxonomy.qza \
```
- --m-metadata-file \$metadir\Serum5.tsv \
- --o-visualization \$filtdir\$fil/\$fil-excluded-table.qzv

```
645 Meta-filter.sh
646 module load QIIME2/2019.7
647
648 tablein=/data/mramseylab/raw_reads/2018_Serum/Merge_runs/denoise-
649 table-merge.qza
650 clsdir=/data/mramseylab/classifiers/
651 metadir=/data/mramseylab/metadata/
652 visdir=/data/mramseylab/visualizations/
653 filtdir=/data/mramseylab/proc_reads/
654 # filter status of input files, "initial" is the 1st pass no filter at 
655 all
656 # "initial-human" is the initial data but for only human samples, no 
657 mock or control samples
658 fil=initial-human 
659
660 #note must make directores needed BEFORE running the below command
661 mkdir $filtdir$fil
662
663 qiime feature-table filter-samples \
664 --i-table $tablein \
665 --m-metadata-file $metadir\Serum5.tsv \
666 --p-where "[Source]='Human'" \setminus667 --o-filtered-table $filtdir$fil/$fil-table.qza
668
```


```
700 --o-visualization $filtdir$fil/$fil-excluded-table.qzv
701
702 #Generate viewable feature table to look for ASVs of interest / 
703 filtering stats
704 qiime feature-table summarize \
705 --i-table $filtdir$fil/$fil-excluded-table.qza
706 --o-visualization $filtdir$fil/$fil-excluded-table2.qzv \
707 --m-sample-metadata-file $metadir\Serum5.tsv
```


```
740 --i-table /$filtdir$fil/$fil-table.qza \
741 --i-taxonomy $clsdir\silva-mod-taxonomy2.qza \
742 --p-mode contains \n\743 --p-exclude "$i" \
744 --o-filtered-table /$filtdir$fil/$fil-table.qza
745
746 done
747 #Use excluded table to generate barplot for checking
748 qiime taxa barplot \
749 --i-table /$filtdir$fil/$fil-table.qza \
750 --i-taxonomy $clsdir\silva-mod-taxonomy2.qza \
751 --m-metadata-file $metadir\Serum5.tsv \
752 --o-visualization /$filtdir$fil/$fil-table-barplot.qzv
753
754 #Generate viewable feature table to look for ASVs of interest / 
755 filtering stats
756 qiime feature-table summarize \
757 --i-table /$filtdir$fil/$fil-table.qza
758 --o-visualization /$filtdir$fil/$fil-table.qzv \
759 --m-sample-metadata-file $metadir\Serum5.tsv
760
```


791 --m-metadata-file \$metadir\Serum5.tsv \

```
792 --o-visualization $filtdir$fil/$fil-table-barplot.qzv
```
793

- 794 #Generate viewable feature table to look for ASVs of interest /
795 filtering stats filtering stats
- 796 qiime feature-table summarize \
- 797 --i-table \$filtdir\$fil/\$fil-table.qza
- 798 --o-visualization \$filtdir\$fil/\$fil-table.qzv \
- 799 --m-sample-metadata-file \$metadir\Serum5.tsv

```
801 Alpha-beta2.sh
802 module load QIIME2/2019.7
803
804 rawdir=/data/mramseylab/raw reads/2018 Serum/Merge runs/
805 procdir=/data/mramseylab/proc_reads/
806 metadir=/data/mramseylab/metadata/
807 visdir=/data/mramseylab/visualizations/
808 cmr="core-metrics-results"
809 # filter status of input files, "initial" is the 1st pass no filter at 
810 all
811 fil=-initial
812
813 #note change name of input tables for the 1st command below. Some 
814 input tables did not have standardized filename conventions
815 #changed to AF-hum-table.qza and table3.qza to minF-hum-table.qza
816
817 array=( minF AF SF NP )
818
819 for i in "${array[@]}"
820 do
821
822 qiime diversity core-metrics-phylogenetic \
823 --i-phylogeny $rawdir\rooted-tree-merge.qza \
824 --i-table $procdir$i\-hum/$i\-hum-table.qza \
825 --p-sampling-depth 5000 \setminus826 --m-metadata-file $metadir\Serum5.tsv \
827 --output-dir $procdir$i\-hum/$cmr
828
829 qiime diversity alpha-group-significance \
830 --i-alpha-diversity $procdir$i\-hum/$cmr/faith pd vector.qza \
```

```
831 --m-metadata-file $metadir\Serum5.tsv \
832 --o-visualization $procdir$i\-hum/$cmr/faith-pd-group-
833 significance.qzv
834
835 qiime diversity alpha-group-significance \
836 --i-alpha-diversity $procdir$i\-hum/$cmr/evenness vector.qza \
837 --m-metadata-file $metadir\Serum5.tsv \
838 --o-visualization $procdir$i\-hum/$cmr/evenness-group-
839 significance.qzv
840
841 done
842
843 #dual array taking the directories above and then running the next 
844 command on the 3 filenames in array2 for each directory prefix in
845 array 1
846 #note different array command syntax from above vs below
847
848 array1=( minF AF SF NP ) 
849 array2=(unweighted unifrac distance matrix
850 weighted unifrac distance matrix bray curtis distance matrix )
851 for indirs in ${array1[@]} 
852 do
853 for infils in ${array2[@]}
854 do 
855
856 qiime diversity beta-group-significance \
857 --i-distance-matrix $procdir$indirs\-hum/$cmr/$infils\.qza
858 --m-metadata-file $metadir\Serum5.tsv \
859 --m-metadata-column Condition \
860 --o-visualization $procdir$indirs\-hum/$cmr/$infils\.qzv \
861 --p-pairwise
```


done

```
865 MMR20_ellipses_NMDS.R
866 setwd("E:/2019 Diabetic Serum/Figure Materials/R-Plots/")
867 library(vegan)
868 library(ggplot2)
869 library(dplyr)
870 set.seed(30)
871
872 metadata<-read.csv("Serum5.tsv", header=TRUE, sep="\t", 
873 stringsAsFactors = F)
874 featuredf<-read.csv("NP-hum-features.tsv", header=TRUE, sep="\t",
875 stringsAsFactors = F)
876
877 #making a subset of human only metadata and a few other criteria.
878 metadata <- metadata %>% filter(Source == 'Human') %>% 
879 select("sample.id","Condition","Run","Source","ABC")
880
881 #Fixing 1st row to make it row names
882 featuredf \leq data.frame(featuredf[,-1], row.names = featuredf[,1])
883
884 #transposing trial1 table to match Evelyns data format
885 featuredf <- as.data.frame(t(featuredf))
886
887 #Brings tables into agreement on matching sample.id values
888 featuredf = featuredf \frac{1}{2} mutate(sample.id = rownames(featuredf))
889
890 #Brings tables into agreement on matching sample.id values
891 table all = left join(featuredf, metadata)
892
893 #This gets metadata to agree with row numbers of the features table 
894 after splitting out from table all
```
 metadata <- 896 select(table all, "sample.id", "Condition", "Run", "Source", "ABC") #filter the "sample.id" column off the very end of the dataframe and turn it into rownames #NOTE THIS VALUE MUST BE CHANGED FOR EACH INPUT 901 featuredf \leq data.frame(featuredf[,-1101], row.names = featuredf[,1101]) # normalize data by sum of ASVs in each sample (this was from ZP's code) 906 featuredf \leq sweep(featuredf, 2, colSums (featuredf), $\dot{\ }$) ##Above this was all data input / manipulation, Below this is data analysis and plotting## ### The function metaMDS is used to calculate the dissimilarity matrix using the bray curtis distance metrics and at the same #time generates the values from the dissimilarity matrix for an ordination plot. 915 MDS \leq -metaMDS (featuredf, distance = "bray", $k = 3$, trymax = 500) 917 ## Next, extract the x and y coordinates from the MDS plot into a new data frame and add the metadata factors to the coordinates the data should be plotted based on. #MMR- added ABC=as.factor and changed [,3] to [,5] to agree with my own metadata file. 922 NMDS t1=data.frame(NMDS1=MDS\$point[,1],NMDS2=MDS\$point[,2], 923 Condition=table all\$Condition,ABC=table all\$ABC) ## set theme for following plots 926 theme set(theme bw())

928 ## Generate the ordination based on the solution from above and 929 selected grouping factor (Condition) 930 plot.new() 931 ord<-ordiellipse(MDS, table all\$Condition, 932 display = "sites", kind ="sd", conf = 0.95, label 933 $= T$) 934 dev.off() 935 936 ## Data frame df ell t1 contains values to show ellipses. It is 937 calculated with function veganCovEllipse which is hidden in vegan 938 package. 939 #This function is applied to each level of NMDS (group) and it uses 940 also function cov.wt to calculate covariance matrix. 941 veganCovEllipse<-function (cov, center = $c(0, 0)$, scale = 1, npoints = 942 100) 943 { 944 theta \leftarrow (0:npoints) * 2 * pi/npoints 945 Circle \leftarrow cbind(cos(theta), sin(theta)) 946 t(center + scale $*$ t(Circle $*$ * chol(cov))) 947 } 948 949 ##Generate ellipse points based on 95% confidence (SD) intervals 950 #Reference : http://stackoverflow.com/questions/13794419/plotting-951 ordiellipse-function-from-vegan-package-onto-nmds-plot-created-in-952 ggplot 953 #Data frame df ell contains values to show ellipses. It is calculated 954 with function veganCovEllipse which is hidden in vegan package. This 955 function is applied to each level of NMDS (group) and it uses also 956 function cov.wt to calculate covariance matrix. 957 df ell $t1$ <- data.frame() 958 for(g in levels(NMDS t1\$Condition)){ 959 if(g!="" && (g %in% names(ord))){ 960 df ell t1 \leftarrow rbind(df ell t1, 961 cbind(as.data.frame(with(NMDS t1[NMDS t1\$Condition==q,],

962 963 veganCovEllipse(ord[[g]]\$cov,ord[[g]]\$center,ord[[g]]\$scale))),Conditi 964 on=g)) } } 965 966 ## Calculate p-value: 967 adon t1<-adonis2(featuredf ~Condition, data=metadata, 968 by=NULL, method="bray", k=3) 969 970 NMDSplot $t1<$ -ggplot(data=NMDS $t1$, aes(NMDS1,NMDS2,col=Condition))+ 971 #update from Evelyn to add metadata text to plot if wanted 972 #This is useful to identify outlier points if using sample ID text 973 here, Remove for final plot 974 ### 975 geom text(aes(NMDS1, NMDS2, label=table all\$sample.id), size=2, vjust=0) + 976 # add the p-value in the bottom right corner 977 annotate("text", x=min(NMDS_t1\$NMDS1), y=min(NMDS_t1\$NMDS2-0.5), 978 label=paste("p= ", adon t1\$`Pr(>F)`[1]), size=3)+ 979 # draw the ellipses and define color based on the grouping factor 980 geom path(data=df ell t1, aes(x=NMDS1, y=NMDS2, linetype=Condition), 981 $size=1$)+ 982 #scale linetype manual(values=c("4-Con"="dotted","3-S4"="solid","2- 983 RI-old"="longdash"))+ 984 #scale colour manual(values=c("4-Con"="red", "3-S4"="darkgreen", "2- 985 RI-old"="purple"))+ 986 # add the points per sample and define shape based on TankLocation 987 geom point(aes(shape=ABC), size=2) + 988 # Reorder the legend 989 guides(color = guide legend(order=1), lty= guide legend(order=1), 990 shape = guide legend(order=2), legend.position = "bottom") 991 992 # Adding other Aethestics

```
993 NMDS by Trial1 <- NMDSplot t1 + theme(axis.text.y =
994 element text(size="12", color="black"), axis.title.y =
995 element text(face="bold", size="12", color="black"))+
996 theme(axis.text.x = element text(size="12", color="black"),
997 axis.title.x.bottom = element_text(face="bold",size="12", 
998 color="black")) +
999 theme(axis.text.x.top= element text(face="bold", size="12",
1000 color="black")) +
1001 #change name of the legend
1002 theme(legend.title=element_blank())+
1003 theme(legend.position = "right", legend.title = 
1004 element text(colour="black", size=16, face="bold"))
1005
1006
1007 print(NMDS_by_Trial1)
1008 print(NMDSplot t1)
1009
```

```
1010 Lefse-noTax.sh
1011 module load QIIME2/2019.7
1012
1013 clsdir=/data/mramseylab/classifiers/
1014 metadir=/data/mramseylab/metadata/
1015 visdir=/data/mramseylab/visualizations/
1016 filtdir=/data/mramseylab/proc_reads/
1017 rawdir=/data/mramseylab/raw_reads/2018_Serum/Merge_runs/
1018 procdir=/data/mramseylab/proc_reads/
1019
1020 array=( minF-hum AF-hum SF-hum NP-hum )
1021
1022 for i in "${array[@]}"
1023 do
1024
1025 qiime feature-table relative-frequency \
1026 --i-table $filtdir$i/$i\-table.qza \
1027 --o-relative-frequency-table $filtdir$i/$i\-table.notax.qza
1028
1029 qiime tools export \
1030 --input-path $filtdir$i/$i\-table.notax.qza \
1031 --output-path $filtdir$i/
1032
1033 #note, must use single hashes for -i / -i unlike other qiime commands.
1034 biom convert \
1035 -i $filtdir$i/feature-table.biom \
1036 -o $filtdir$i/$i\-notax.table.txt \
1037 --header-key "taxonomy" \
1038 --to-tsv
```
