

1 **Supplementary Material**

2 **Materials and Methods**

3 **Dataset Collection and Validation:** The datasets used for this study were identified by body part
4 or tissue being affected in samples collected – chronic lower extremity ulcers for CW samples
5 and sputum samples from people with CF, presence of original metatranscriptomic data, presence
6 of comorbidities in patients and treatment course. We passed only reads with a minimum length
7 of 22 to improve the specificity of mapping to reference databases in downstream application as
8 previous data has shown that read length influences mapping (1).

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10 **Community Composition Analysis:** We reduced our Metaphlan4 output data to genera and
11 species that had at least one percent abundance prevalence in at least 3 samples. Then we
12 grouped these organisms at both genera and species levels as either aerobes, facultative
13 anaerobes or obligate anaerobes. We defined obligate/facultative anaerobes based on culture
14 requirements and how these organisms have frequently been identified/grouped in previous
15 literature and textbooks (2-9).

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17 **Functional Profiling:** Two bioinformatic tools were used for functional profiling. The Simple
18 Annotation of Metatranscriptomes by Sequence Analysis tool (SAMSA2) identified 4527 level 4
19 enzyme classes and the DIAMOND_analys_counter.py scripts were used to aggregate the
20 outputs which were eventually exported to R studio for statistical analysis. The diversity_stats.R
21 script of the SAMSA2 package was used to compute the mean Shannon and Simpson diversity
22 indices for the two infection communities and the diversity_gaphs.R script was used to plot the
23 graphs of these. We performed differential expression analysis of the features in the both
24 communities using the run_DESeq_stats.R script. The HMP Unified Metabolic Analysis Network
25 (HUMANn3) tool, identified 594,273 UniRef protein families which were further regrouped to 2459
26 unique ECs using the humann_regroup_table command and exported to Rstudio for differential
27 expression analysis. For pathway mapping by HUMANn3, 490 pathways with metacyc IDs were
28 identified and 162 of these had differential expression across the two infection communities.

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31 **Assigning Functions to Functional Categories:** To assign E.C. numbers to functional
32 categories, we first converted E.C. numbers to KEGG terms and used these to assign them to
33 KEGG pathways. Pathways were then assigned to broader unified functional categories (e.g.
34 Histidine metabolism and Tryptophan metabolism were both reclassified to “Amino acid
35 metabolism”). Any functions without a KEGG ID were annotated with the ExplorEnz, BRENDA,
36 and InterPro databases if possible. Pathways were assigned to MetaCyc “superclasses” based
37 on their pathway IDs and then added to unified functional categories if necessary (e.g.
38 peptidoglycan biosynthesis and lipid A biosynthesis were both assigned to “Cell Wall & Cell
39 Membrane biosynthesis”). While E.C classes allowed us to perform functional category
40 assignment, some of the additional identified features from SAMSA2 were not assigned E.C
41 numbers or were not enzymes such as flagella, pili, typeIII secretion systems, and these were
42 assigned functional categories based on description in previous literature (10-12).

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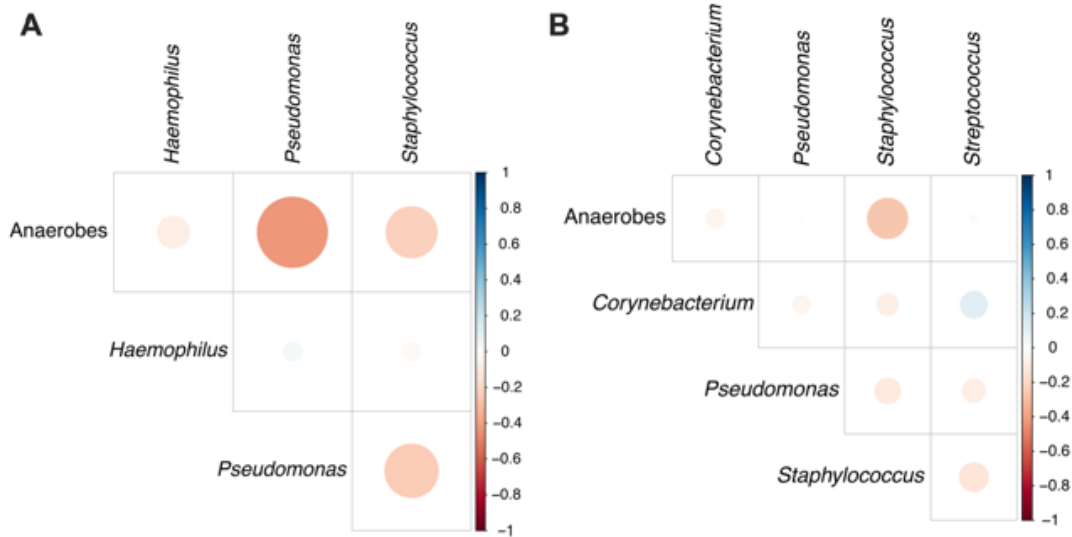
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46 **Statistical Analyses:**

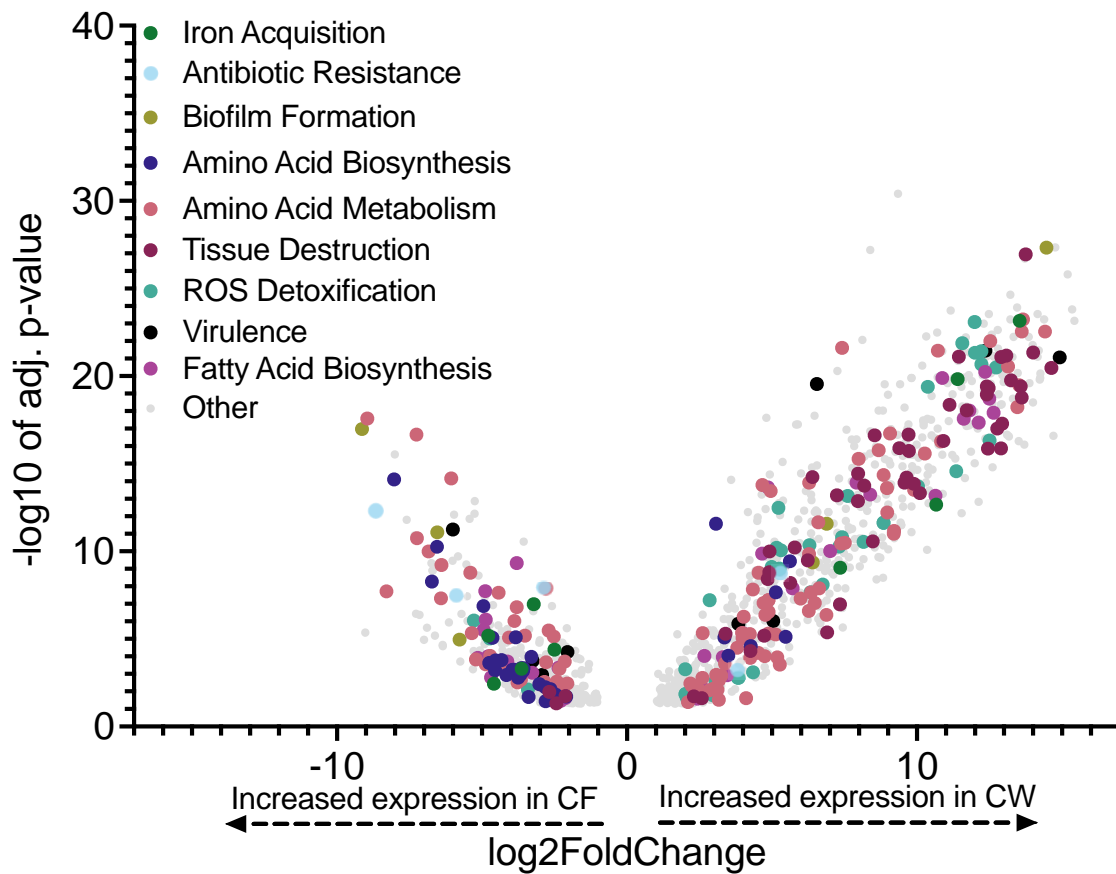
47 For the HUMANn3 functional analyses, we performed differential expression analysis using

48 DESeq2 1.38.3 and MaAsLin2 1.12.0 in R studio.

49 **Figures**

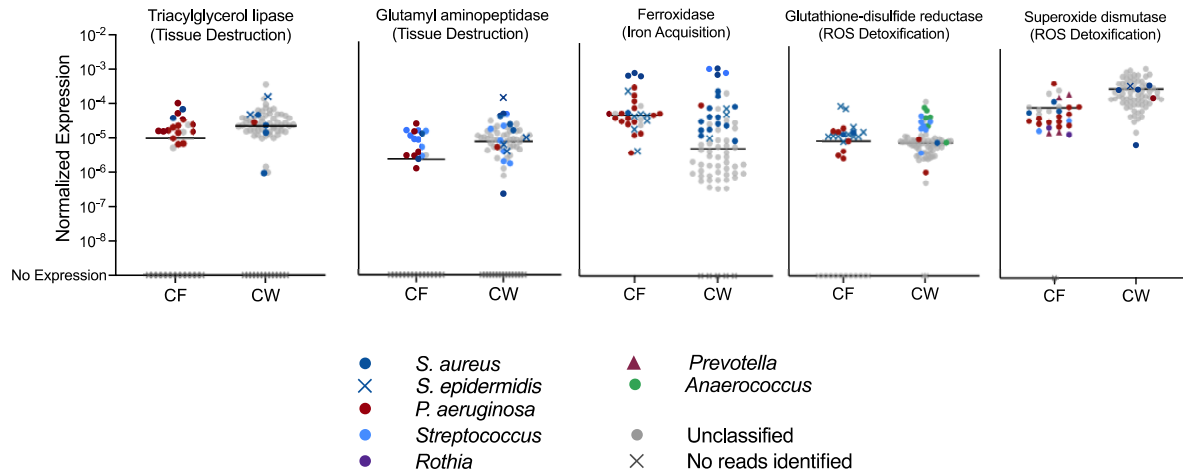


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 51 **Figure S1:** Co-occurrence of obligate anaerobes and traditional pathogens in CF and CW
 52 samples. **A)** Obligate anaerobes (*Prevotella*, *Actinomyces*, *Veillonella* & *Fusobacterium*) have
 53 negative correlation with CF pathogens (*Pseudomonas*, *Staphylococcus*, & *Haemophilus*). **B)**
 54 Obligate anaerobes (*Finnegoldia*, *Anaerococcus*, *Peptoniphilus*, *Peptostreptococcus*, *Parvimonas*
 55 and *Peptococcus*) have negative correlation with CW pathogens (*Staphylococcus*,
 56 *Streptococcus*, *Pseudomonas*, and *Corynebacterium*). **A&B** show the Pearson correlation
 57 coefficients between the sum of the relative abundances of the anaerobes, and each indicated
 58 pathogen. The size of the circles corresponds to their Pearson correlation coefficients while the
 59 color denotes if they are positive (blue) or negative (red).



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Figure S2: Volcano plot to highlight differentially expressed functions in both infection communities as identified by HUMANN3. 42.98% of the functions were differentially expressed (adjusted p-value < 0.05, \log_2 FoldChange > 1) with 753 and 303 functions highly expressed in CW and CF samples respectively.



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70 **Figure S3:** Dot plot to highlight bacterial contribution to the expression of differentially expressed
 71 functions in CF and CW environments. Normalized expression (y-axis) refers to the sum of the
 72 relative abundance of the bacteria species per sample for each function in log scale.

73 **Supplementary Datasets**

74 **Dataset S1:** Detailed metadata on all samples used for the study.

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76 **Dataset S2:** Functional analysis data. Sheet 1 - highlighted conserved functions identified with
77 HUMANN3 and SAMSA2 and bacterial contribution to these functions. Sheet 2 - highlighted
78 differentially expressed functions identified with HUMANN3 and SAMSA2 and bacterial
79 contribution to these functions. Sheet 3 - all differentially expressed functions obtained from
80 HUMANN3. Sheet 4 - all differentially expressed functions obtained from SAMSA2.

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82 **Dataset S3:** Pathway analysis data. Sheet 1- all pathway analysis data as identified by
83 HUMANN3. Sheet 2 - all differentially expressed pathways obtained from HUMANN3.

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87 **SUPPLEMENTAL REFERENCES**

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