Supplementary Information

Association of food insecurity on gut microbiome and metabolome profiles in a diverse college-based sample

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Page S-11, Figure S2. Predicted functional metabolic pathways by food security status. **a** Comparison of the log ratio of the 10 lowest ("Set 1") and 10 highest ("Set 2") ranked pathways associated with food security status (after filtering, $n = 58$). **b** Food secure status had a significantly greater log ratio of Set 2 compared to Set 1 (Mann-Whitney U test: *p* < 0.05).

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prevalence amplicon sequence variants (ASVs) were filtered out using the feature-table plugin with the filter-samples method (--p-min-frequency 10; --p-min-samples 2). **d** To account for uneven sequencing depth between samples, normalization was performed via alpha rarefaction for observed features and Shannon index. Based on the ASV feature table, a p-min-depth of 10 and a p-max-depth of 120,000 was used. **e** Based on assessment of alpha rarefaction a threshold of 22,000 sequences/sample was established leaving 58/60 high quality samples for analysis (participants DW09 and DW96 were removed).

Table S1. Alpha and beta diversity metrics for food insecure and food secure status.

Alpha diversity values for food security status displayed as mean ± SD.

Table S2. Top 10 lowest (set 1) and highest (set 2) ranked genera associated with food security

as produced by Songbird analysis.

Table S3. Top 10 lowest (set 1) and highest (set 2) ranked predicted microbial functions

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Figure S8. Quality-based filtering processes for microbiome data. Quality scores based on rarefaction assessment for **a** Forward and **b** Reverse reads. **c** Low abundance/low prevalence amplicon sequence variants (ASVs) were filtered out using the feature-table plugin with the filtersamples method (--p-min-frequency 10; --p-min-samples 2). **d** To account for uneven sequencing depth between samples, normalization was performed via alpha rarefaction for observed features and Shannon index. Based on the ASV feature table, a p-min-depth of 10 and a p-max-depth of 120,000 was used. **e** Based on assessment of alpha rarefaction a threshold of 22,000 sequences/sample was established leaving 58/60 high quality samples for analysis (participants DW09 and DW96 were removed).