



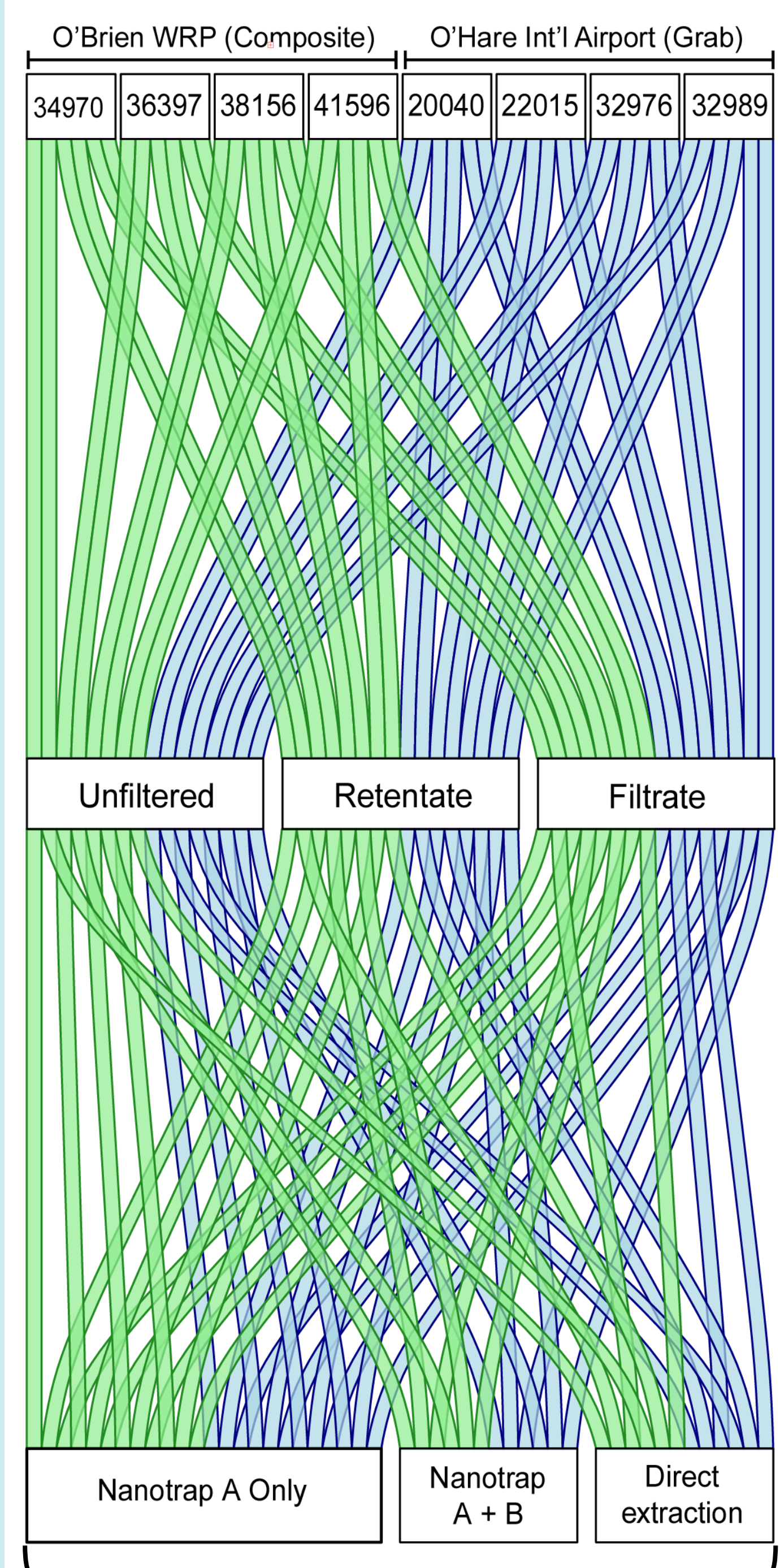
Improving the Signal-to-Noise Ratio of Metagenomic Wastewater Surveillance for Human Pathogens and ARGs

Background

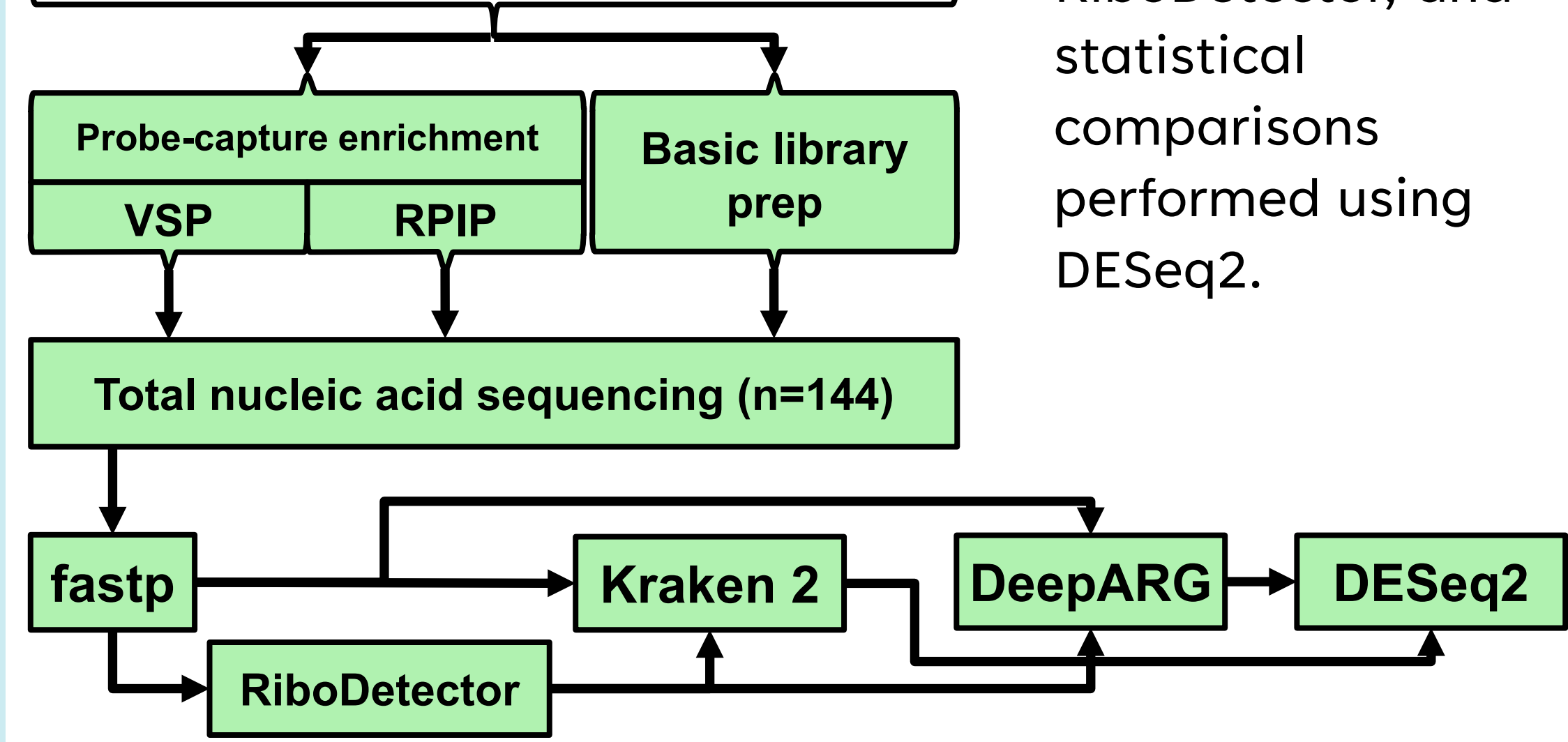
Much of the genetic material found in wastewater originates in the viromes and microbiomes of the surrounding population. This genetic material can therefore be used to infer the presence of pathogens and functional genes such as antimicrobial resistance genes (ARGs) in the local community. However, most genetic sequences in wastewater are not relevant to public health. We subjected wastewater samples to a variety of processing methods intended to increase the relative abundance of public health-relevant sequences.

Methods

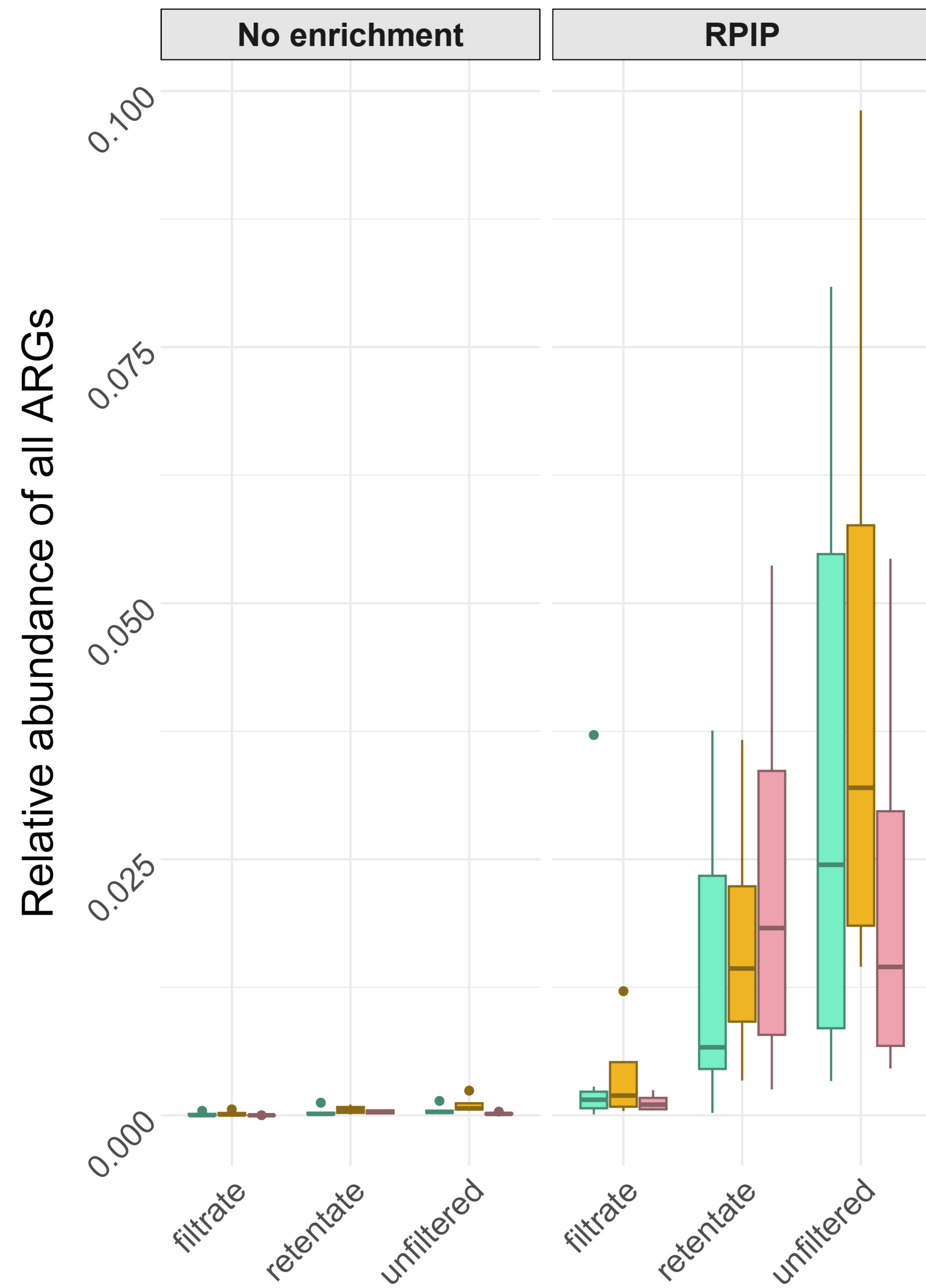
- 4 wastewater samples each from the Terrence J. O'Brien Water Reclamation Plant and O'Hare International Airport.
- Each sample was split into aliquots and subject to several treatments.
- First, we centrifugally filtered aliquots through 24-well plates with 0.45 μm pore size (above). We used both filtrate and resuspended retentate. An unfiltered aliquot of each sample served as a baseline.
- Next, we concentrated aliquots using Nanotrap Microbiome Particles (Ceres Nanosciences)—magnetic beads that adhere to surface structures of cells and viral capsids. We used 2 Nanotrap protocols, designated “A” and “A + B”, as well as direct extraction as a baseline.
- Finally, we sequenced each aliquot 3 ways: using two hybridization capture panels plus untargeted total nucleic acid sequencing. The panels we used were the Viral Surveillance Panel (VSP) and Respiratory Pathogen ID and AMR Enrichment Panel (RPIP), both from Illumina.



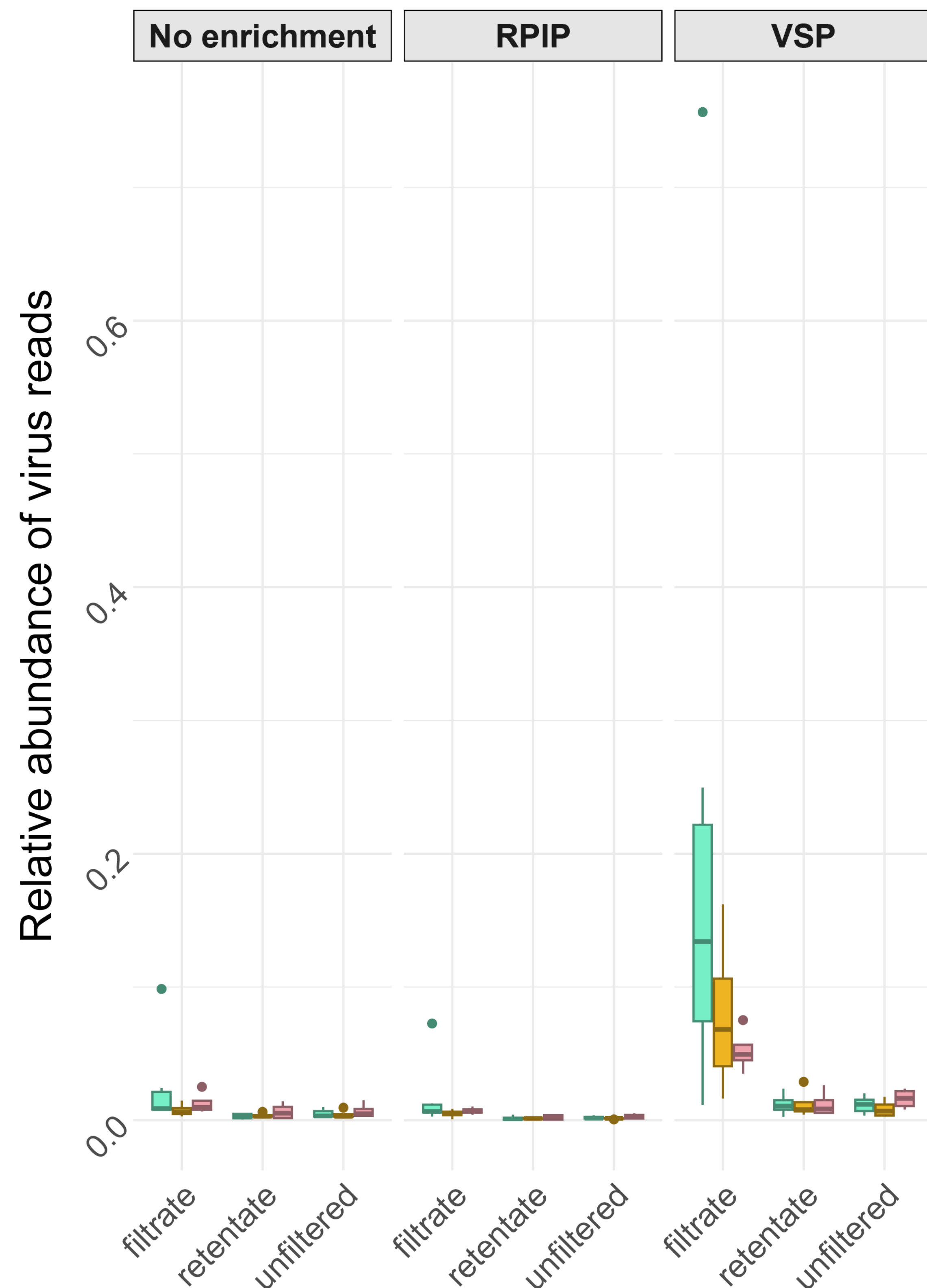
- VSP Targets full genomes of 66 human pathogenic viruses.
- RPIP targets > 280 respiratory pathogens (bacteria, fungi, and viruses), > 2000 AMR markers, and full genomes of SARS-CoV-2 and Influenza A and B.
- After performing read trimming and quality filtering with fastp, we assigned taxonomic IDs with Kraken 2 and identified ARG-like sequences with DeepARG. Ribosomal RNA reads were identified using RiboDetector, and statistical comparisons performed using DESeq2.



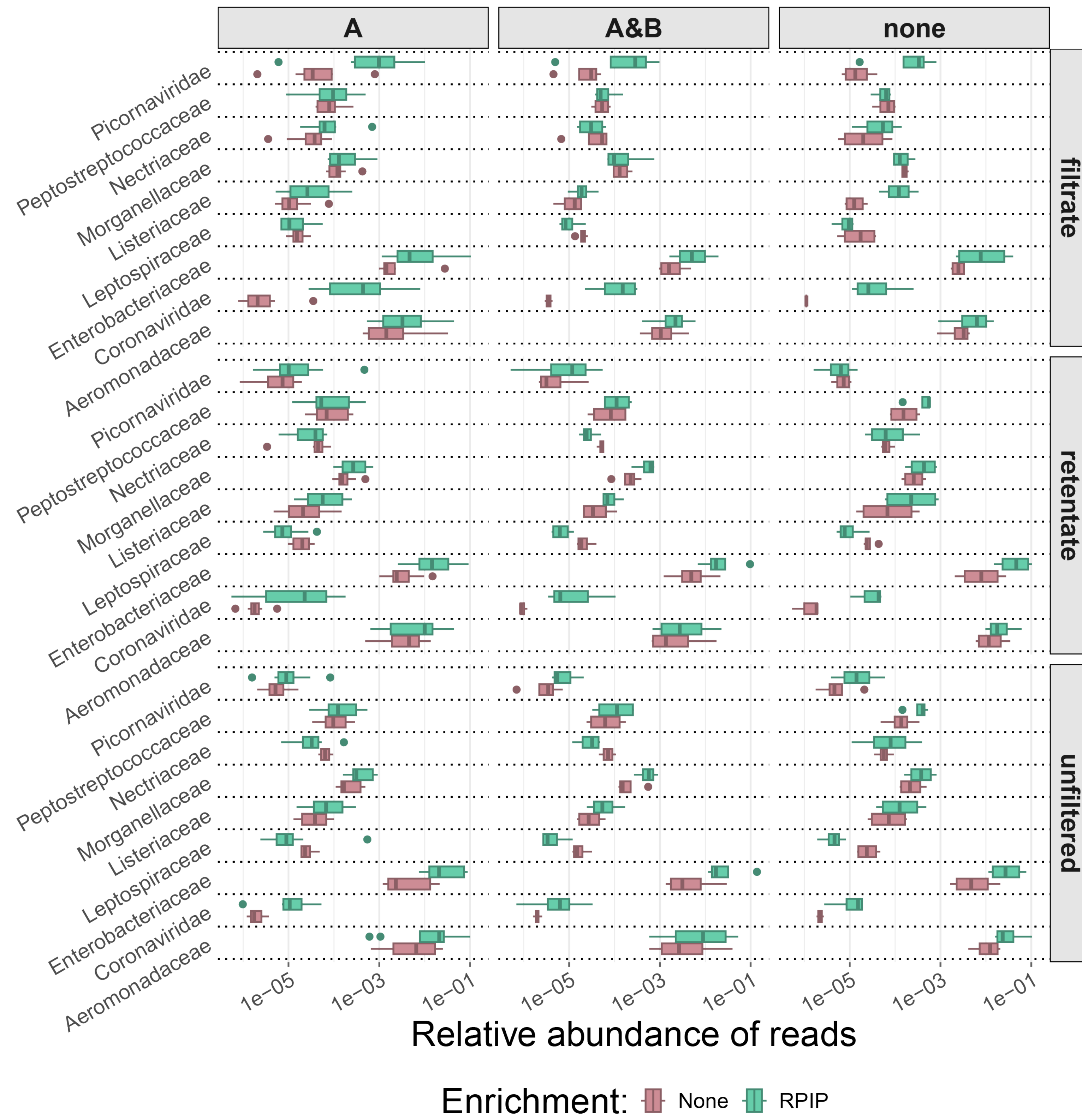
Results & Discussion



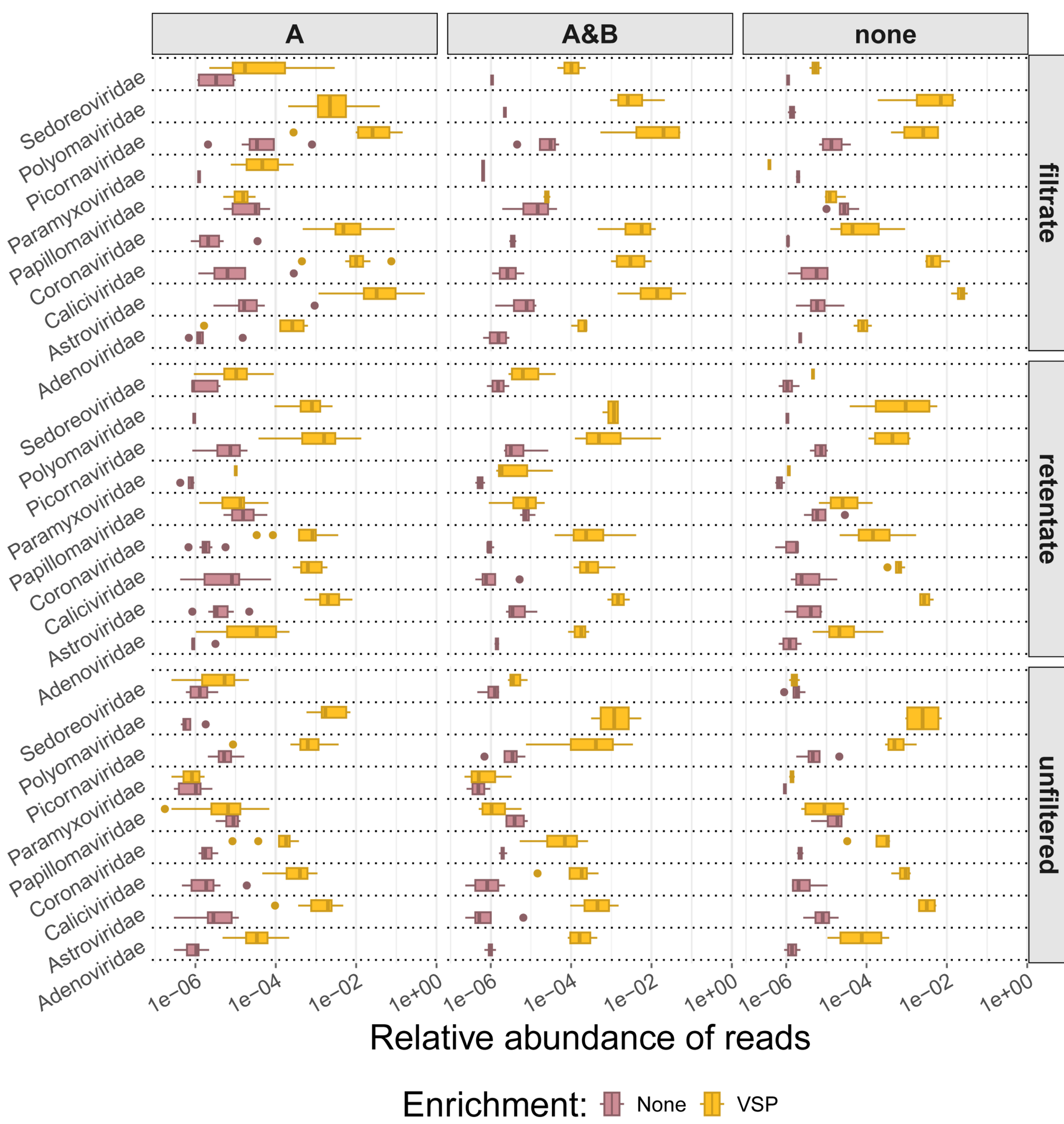
Nanotrap protocol A A&B none
↑ Total portions of predicted ARG reads. RPIP enrichment increases the overall portions of ARG reads. Filtration decreases portions of ARG reads, likely because most are associated with cells that do not pass through the filter.



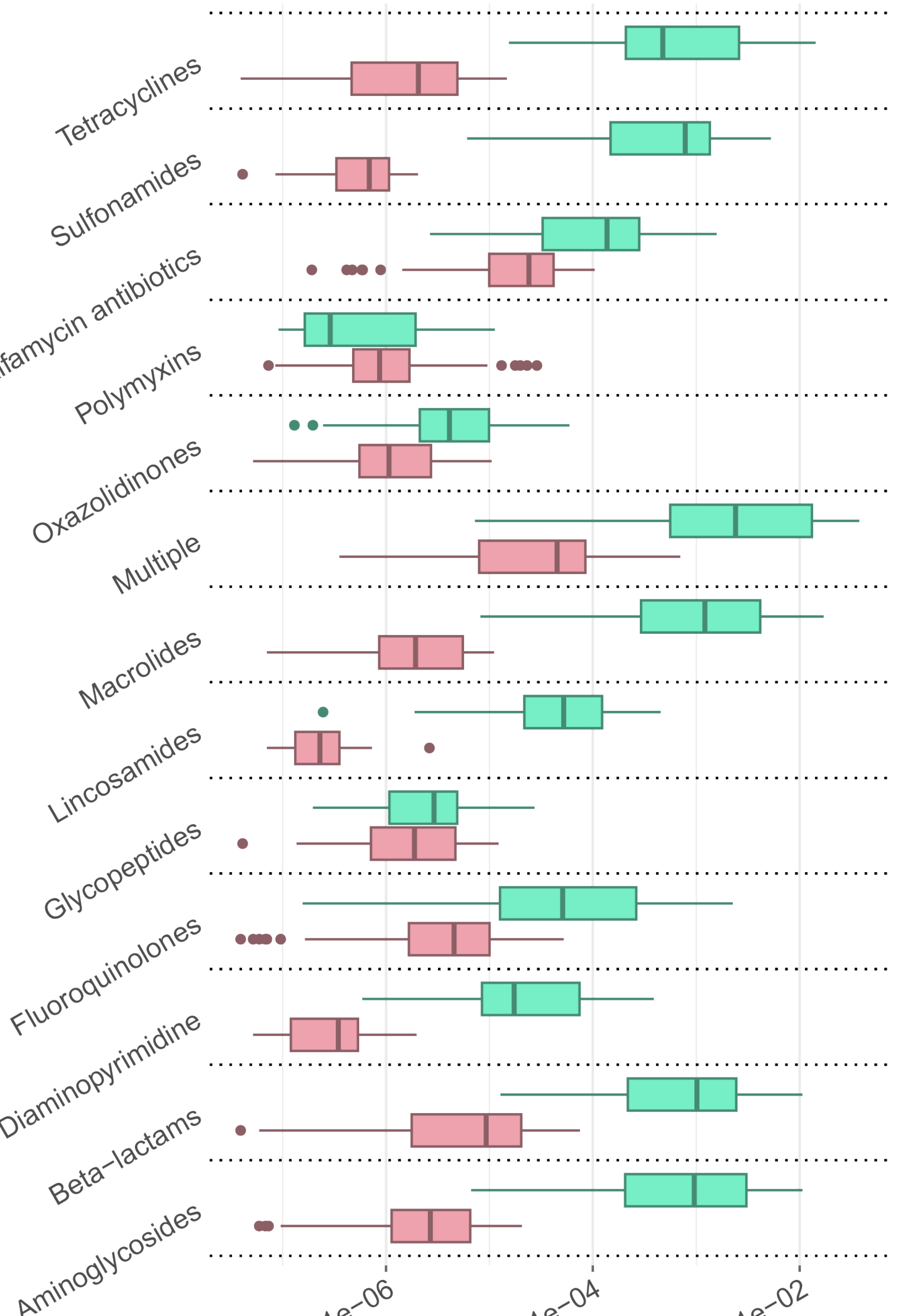
Nanotrap protocol A A&B none
↑ Portion of all reads identified by Kraken2 as belonging to viruses. Filtration combined with VSP enrichment produce the greatest increase in viral reads.



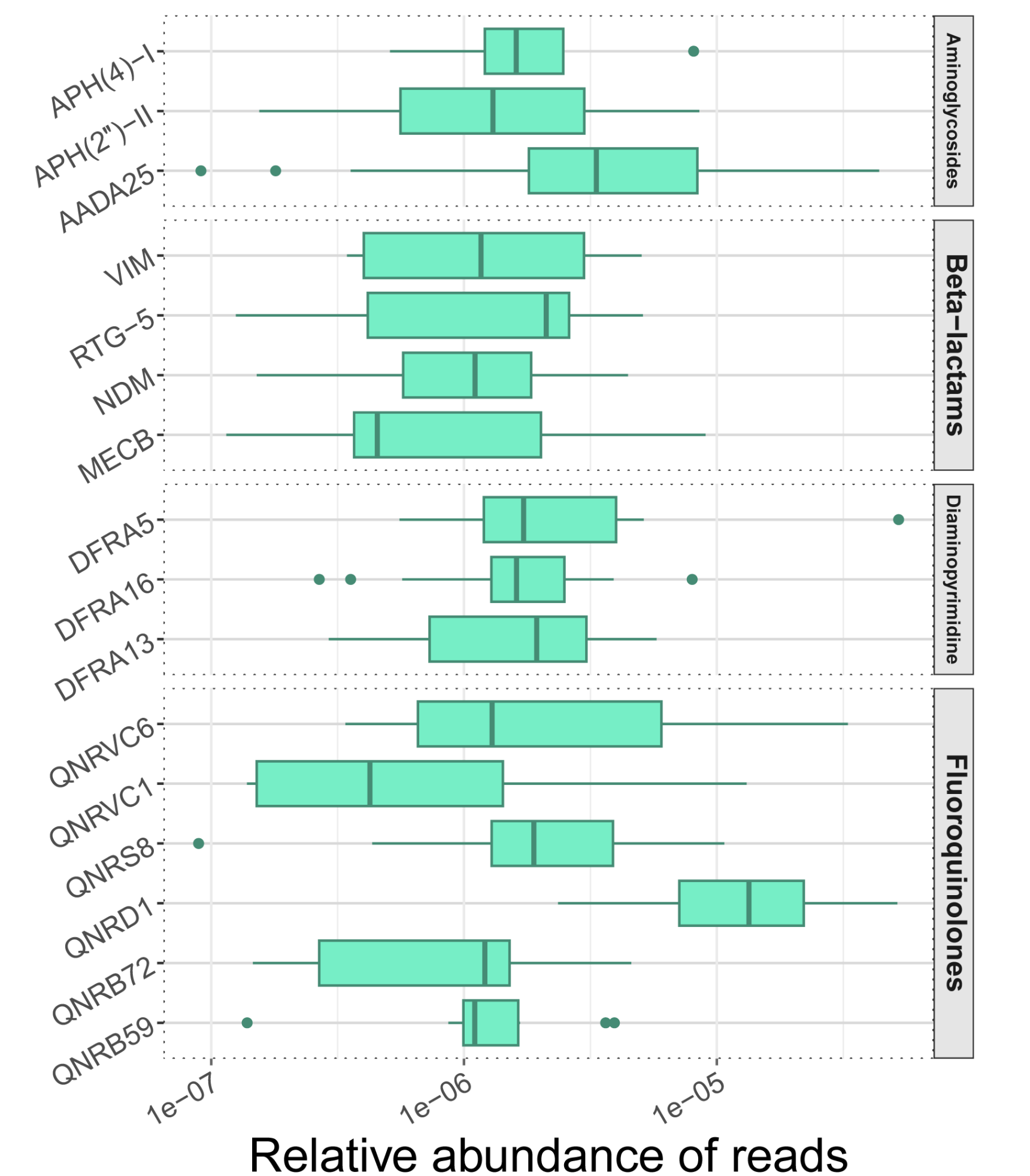
Enrichment: None RPIP
↑ Effect of RPIP enrichment (box color), Nanotrap protocol (columns) and fraction from filtration (rows) on portions of families containing RPIP-targeted taxa. Only targeted taxa with relative abundance at least doubled in at least one condition are included. Impact of filtration is most visible for viruses (*Picornaviridae* and *Coronaviridae*), but there is no clear effect of Nanotrap treatment.



Enrichment: None VSP
↑ Effect of VSP enrichment (box color), Nanotrap protocol (columns) and fraction from filtration (rows) on portions of families containing VSP-targeted taxa. Filtrates generally produce somewhat greater portions of viral reads (though this effect is not as pronounced in individual taxa), while there was not a clear consistent effect of Nanotrap treatment on these taxa.



Enrichment: None RPIP
↑ RPIP-targeted ARGs identified as having significant (based on DESeq2 analysis) differences in relative abundance between RPIP-enriched vs. unenriched samples, pooled by CARD resistance category.



Relative abundance of reads
↑ ARGs found in RPIP-enriched samples but absent from all unenriched samples.

Acknowledgements & References

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