Broliden_5325

10 December, 2020

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#Load libraries and other scripts				
#Defining some variables for the analysis				
Loading data and metadata				
Calculate QC metrics				

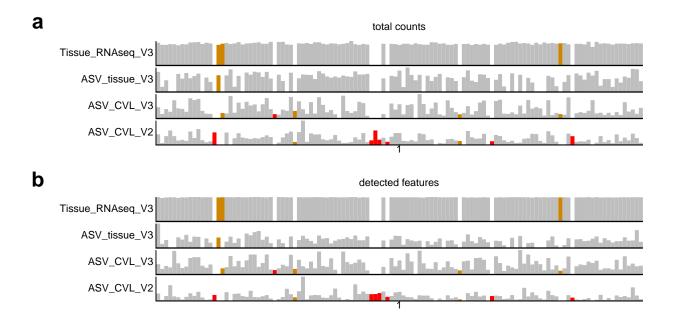


Figure 1. Comparative barplot for the a) total counts and b) number of non-zero detected features (genes / bacteria) for each of the sequencing datasets. Samples are ordered alfabetically according to the patient ID. Samples that are present in exactly two datasets are shown in orange (P016,P017,P035,P076,P101). Samples that are present in exactly two datasets are shown in red (P015,P030,P054,P055,P056,P058,P084,P104).

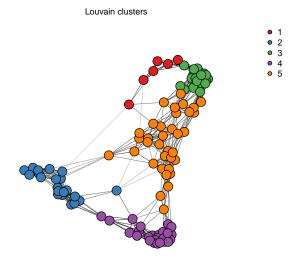


Figure 1b. Sample embedding of 20-SNN graph clustered using Louvain based on the normalized bacterial abundances across all microbiome datasets (including samples not present in all datasets).

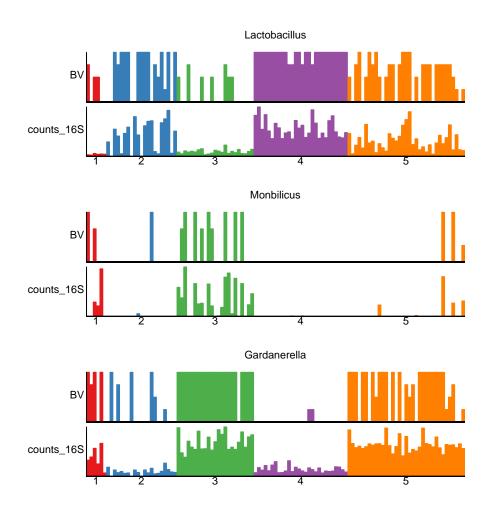


Figure 2. Differential bacterial abundance across joint clustered samples and compared across microbiome datasets. The results are shown both as **a)** Dot plots and **b)** barplots Bacteria with log2FC above 0.25 and p-value below 0.01 were considered significant and were sorted by the highest expression.

Figure 2. Differential bacterial abundance across microbiome datasets. The results are shown both as **a)** Dot plots and **b)** barplots Bacteria with log2FC above 0.25 and p-value below 0.01 were considered significant and were sorted by the highest expression.

Figure 2. Differential bacterial abundance across CVL2 and CVL3 microbiome datasets. The results are shown both as **a)** Dot plots and **b)** barplots Bacteria with log2FC above 0.25 and p-value below 0.01 were considered significant and were sorted by the highest expression.

Figure 2b. Differential bacterial abundance across tissue and CVL microbiome week3 datasets. The results are shown both as **a)** Dot plots and **b)** barplots Bacteria with log2FC above 0.25 and p-value below 0.01 were considered significant and were sorted by the highest expression.

Computing differential expression across microbiome datasets

Figure 4. Differential bacterial abundance across all groups and all microbiome datasets. The results are shown both as **a)** Dot plots and **b)** barplots Bacteria with log2FC above 0.25 and p-value below 0.01 were considered significant and were sorted by the highest expression.

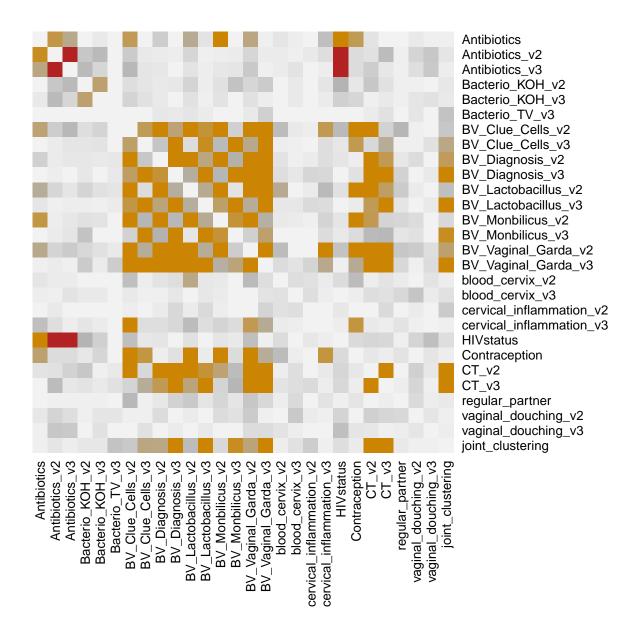


Figure 6. Association analysis across several patient categorical metadata parameters, including patient groupings annotations from microbiome.

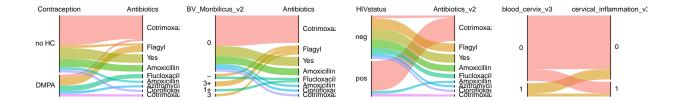
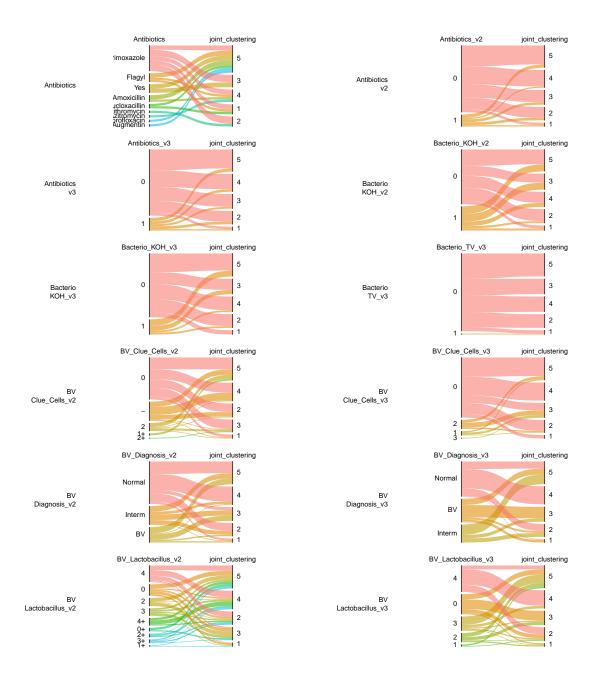
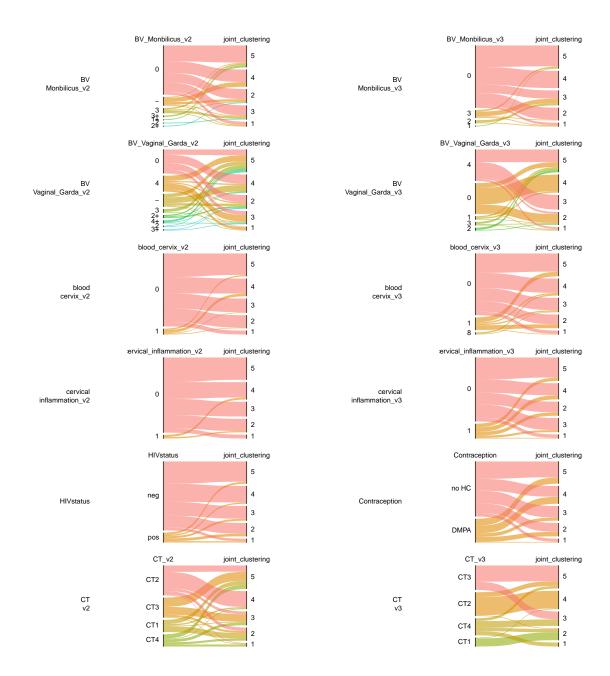


Figure 7. A few examples of significant association between metadata parameters shown as sankey plots.





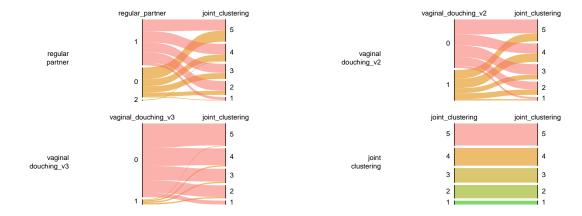


Figure 8. Sankey plots for all tested associations between the patient groups identified in in the microbiome datasets.

Visualise the data

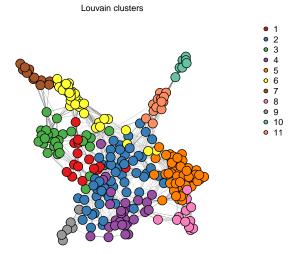


Figure 9a. Bacterial community embedding of 5-SNN graph clustered using walktrap community detection algorithm based on the normalized bacterial abundances across all microbiome datasets (including samples not present in all datasets).

community1	community2	community3	community4	community5	community6
Collinealia perofaciens Corprehacterium peeudogenitalium Enhydrobacter aerosaccus Haemorphilius parainfulenzae other Gordonia other Lactococcus other Mogibacterium other Rhodobacteriareae other Sitephophyta Streptococcus dysgalactiae Tejoldmonas artifensis Veillonella gparvula other Gammaproteobacteria	Aerococus christenseni Alopoburn deltre Bildobacterium adolescentis Bildobacterium longum/breve Brevbacterium longum/breve Brevbacterium pauckovans Brevbacterium revensurgense Campylobacter hominis Chamydis trachomatis Coynebacterium initians Coynebacterium initians Coynebacterium pitrulium Dermabacter hominis Dorea formicigenerans Enterococus villorum/hiraefactum Gemella mothilorum/sanguinishaeri Gemella mothilorum/sanguinishaeri Gemella mothilorum/sanguinishaeri Laciobacilius reuceria Laciobacilius cipiant Laciobacilius reuceria Laciobacilius cipiant	Bacteria Bacterioles carcae Bacterioles dorei Bacterioles dorei Bildobacterium pseudocatenulatum/ Coprococcus comes Delfita tsuruhatensis Escherichia/Shigelia coli/llexneri/boyi other Aciretobacter other Asiretobacter other Asiretobacter other Asiretobacter other Chitinophagaceae other Chitinophagaceae other Chitinophagaceae other Chitinophagaceae other Chitinophagaceae other Chitinophagaceae other Palerobacter other Holdemanella other Janthinobacterium other Peloromoas other Relastonia other Raistonia other Staphylococcus other	Actinomyces neuli Actinomyces turionesis Anaerococcus hydrogenalis Anaerococcus murdochii Bifdobacterium bilidum Dialister propionidraciens Dialister succinatiphilus Finegoldia maginama Granulicatella elegana Granulicatella elegana Granulicatella elegana Horricatella Other Burkholdrai other Eubacterium other Obsenella other Hydrogenophaga other Methylobacterium other Obsenella other Hydrogenophaga other Methylobacterium other Obsenella other Velilonella Prevotella bibria Prevotella bibria Prevotella Prevotella Sibria Sickica esigua Solobacterium moorei Stackica esigua Solobacterium moorei Streptococcus agalactias Streptococcus agalactias Streptococcus agalactias Streptococcus agalactias Welhoactum cambriense Varibaculum cambriense Velilonella montpellierensis Methylobacterium radiotolerans	Actinomyces broglycogenia Actinomyces urogentalis Aniaerococcus prevoilitetradius Atopoblum vaginae Dialister micraerophilus Gardnerella vaginalis Gemella asaccharolytica Mobiluncus curtisi Mobiluncus other Arcanobacterium other Bulledia Other Bulledia Other Bulledia Other Bulledia Other Bulledia Other Bulledia Other Gardnerella Other Bulledia Other Falsiporphyromonas Other Falsiporphyromonas Other Fandibacter Other Parvibacter	Brachybacterium paraconglomeratun Chrysoboaterium haidness Corynebacterium freneyis Couprisvidus metallidurans Lactobacilius plantarum/fabilermenta Mesorhizobium loti other Actinomyces other Actinomycess other Brachybacterium other Brachybacterium other Brachybacterium other Dietzia other Frimicutes other Brachybacterium other Cloactbacterium other Checklandia other Firmicutes other
community7	community8	community9	community10	community11	
other Arthrobacter other Chryseobacterium other Exquobacterium other Lanibacter other Particulations of the Particulation of the Particulation other Phyliobacterium other Neothizobium other Neothizobium other Sphingobacterium	Anaerococcus lactolyticus Anaerococcus obesiensis Campylobacter ureolyticus Fusobacterium equinum Fusobacterium nucleatum other Campylobacter other Paraprevotella other Prevotellaceae other Prevotellaceae other Proteobacteria Streptococcus salivarius/vestibularis/ Sutterella sanguinus	Lactobacillus coleohominis Lactobacillus crispatus/acidophilus Lactobacillus iners Lactobacilus jenseni Lactobacilus reuteri/oris/frumenti/ani other Lactobacillales	Bacteroides vulgatus Blautia obeum Faecalibacterium prausnitzii other Bacillus other Badynhizobium other Roseburia Roseburia faecis	Anaerostipes hadrus Dorea longicatena Lactobacillus vaginalis other Alloprevotella other Blautia other Faecallbacterium other IV other Oscillibacter other Ruminococcaceae other Ruminococcaceae Prevotella copt Streptococcus intermedius/constellat Fusicatenibacter saccharivorans	

Figure 9b. List of bacterial belongin to each community.

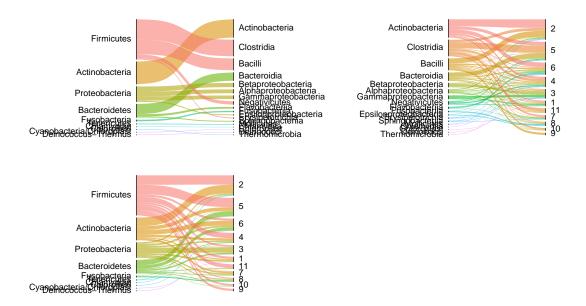


Figure 9b. Taxonomic annotation of bacterial communities.

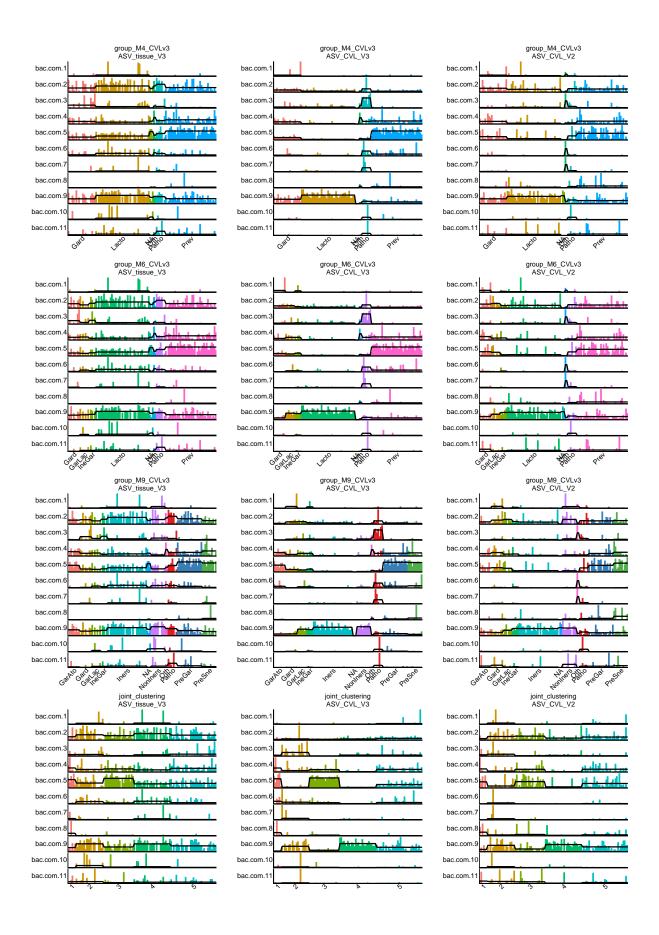
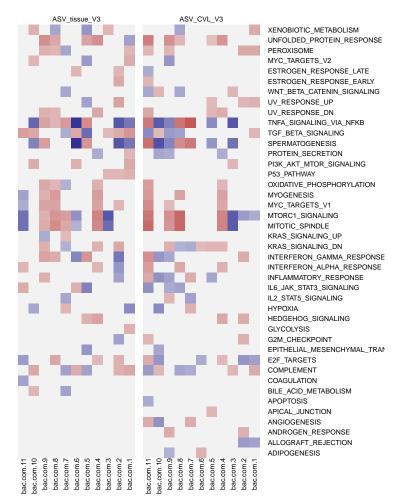
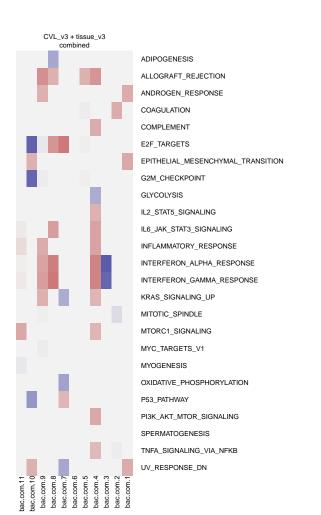


Figure 5. Comparisson of microbiome communities across all datasets and all clustering methods.





HALLMARK

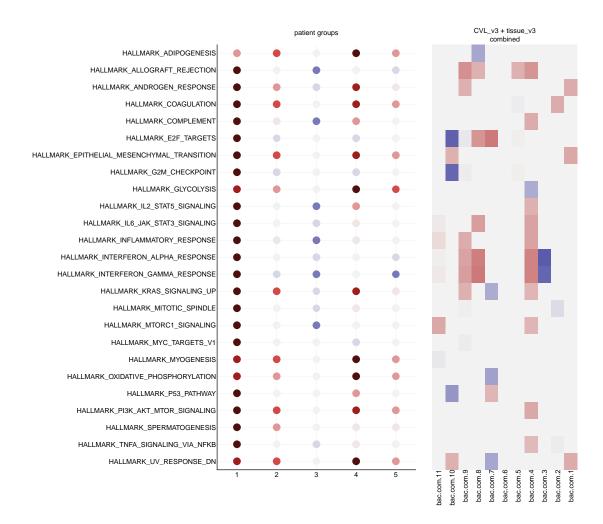
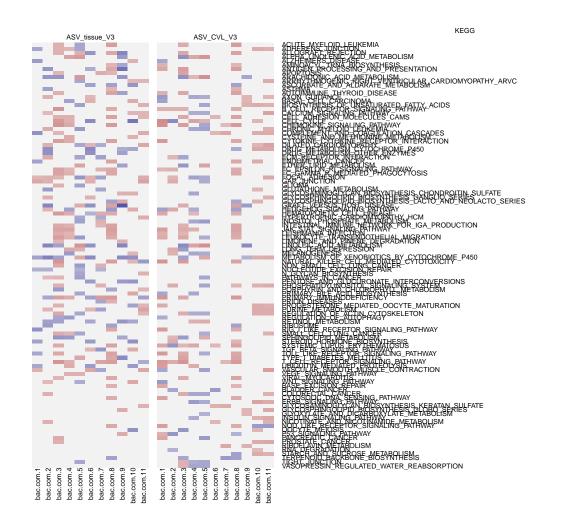


Figure 9. Heatmaps showing the functional association of microbiome datasets with the expression profiles in the RNAseq dataset. Briefly, bacterial abundances from each dataset were correlated with the gene expression of the top 5000 highly variable genes from the RNAseq dataset, generating a correlation matrix between bacteria and genes. Then for each bacteria, we rank genes based on their correlation to that bacteria and perform gene set enrichment anlaysis (GSEA) using the Hallmark MSigDB gene annotation database. This, in turn, will result in a matrix associating every bacteria with every Hallmark MSigDB process in the tissue. The heatmap shows the normalized enrichment score (NES). Only enrichments with pvalue below 0.05 are shown.



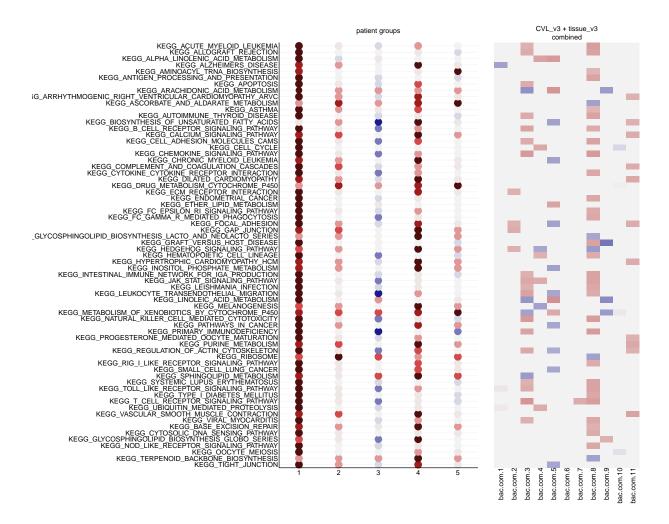


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