

Master Degree in Quantitative and Computational Biology

Project group Computational Microbes Genomics

Group:

Andrea Tonina, Lorenzo Santarelli, Gloria Lugoboni

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1 Introduction

1.1 Metagenome Sequencing, Assembly, and Binning

Metagenome sequencing enables the construction of metagenomes-assembled genomes (MAGs). A MAG can be seen as a microbial genome obtained by a preliminary passage of genome assembly of high quality contigs. This kind of analysis enables us to identify novel species thanks to a passage of annotation and taxonomic classification [1].

A typical metagenome project involves a specific pipeline, a step of sample processing and sequencing, a step of assembly and finally a step of binning followed by genome-annotation. This whole process is than completed with a statistical analysis [2].

Metagenomics is possible thanks to the study of DNA genomes, the sequencing is possible using a variety of novel sequencing technologies and platforms like Roche 454 sequencing, Illumina sequencing, and ion torrent Personal Genome Machine (PGM) [3].

Thanks to the process of assembly it is possible to reconstruct genomes. This method is based on a process of alignment and merging of overlapping sequences, creating large contiguous regions (contigs) [4].

After the process of assembly is completed, contigs are grouped by their organism of origin into bins, using a process known as binning [5]. The selection of high quality bins enables the identification of MAGs, these are characterized by a high completness and low levels of contamination and are used to operate taxonomic annotation and gene prediction [6]. These can be grouped together in the same species genome bin (SGB) if they exceed a certain threshold of nucleotide identity, with a treshold of the 5% for genomic identity. It is possible to assign a taxonomic label based on the presence (or not) of characterized genomess [7]. If a genome with associated taxonomy is not availabe, we talk about known SGB (kSGB), while in the opposite case, we talk about unknown clades (uSGB) [8].

With the term pangenome, we indicate the union of the *core genome*, containing genes present in all strains, and the *dispensable genome*, also called *accessory genome*, containing genes present in two or more strains and genes unique to single strains [9].

DA SPIEGARE IL PANGENOME ANAYSIS E LE ALTRE ANALISI CHE FACCIAMO??

2 Methods

2.1 Softwares and parameters used

2.1.1 Genome annotation (Prokka)

Prokka is a fast and accurate command line software tool used to annotate prokaryoyic genomes. It produces standards-compliant output files that can be used for further analysis or viewing in genome browsers.

Prokka expect one single input file in a FASTA format, containing an assembled genome. The process of annotation is possible thanks to the comparison of the gene codes with a large database of known sequences, identifying the best match as the most significative one and therefore associating the labelling and the relevant features to the gene codes. Prokka use this method in an hieratical manner, using initially small and reliable databases moving only at the end of the process to protein family databases. Prokka produces several outputs file, listed in the Figure 2.1 [10].

We need to specify several parameters, specifically, the input files, our MAGs, the output directory --outdir and the parameter --kingdom Bacteria that is needed to specify the annotation mode, to make the prokka more fast.

prokka --kingdom Bacteria --outdir SGB15132_prokka_output .f*.

Suffix	Description of file contents
fna	FASTA file of original input contigs (nucleotide)
faa	FASTA file of translated coding genes (protein)
ffn	FASTA file of all genomic features (nucleotide)
fsa	Contig sequences for submission (nucleotide)
bl	Feature table for submission
sqn	Sequin editable file for submission
gbk	Genbank file containing sequences and annotations
gff	GFF v3 file containing sequences and annotations
og	Log file of Prokka processing output
xt	Annotation summary statistics

Figure 2.1: Prokka outputs files [10].

2.1.2 Pangenome analysis and Phylogenetic analysis (Roary, Roary + FastTree)

Roary is a tool that enables the construction of large-scale prokaryote pangenomes, identifying the core and accessory genes.

The input file to Roary is a GFF file containing sequences features.

Roary collects the coding regions from the annotated input genome. It operates a clustering process creating a network and defining a phylogenetic tree. A matrix is therefore obtained and the pangenome (core genes and accessory genes) is defined. The process of clustering is based on the minimum percentage of identity, setted to 95% by default [11].

The main output of Roary is a tree obtained using the presence and absence of the accessory genes. It is a tree used to have an initial insight of the data, grouping in a quick way the genomes based on their accessory genes [11]. It can be visualized using iTOL, an online tool for phylogenetic tree display [12].

Roary returns three graphs, the newick tree associated to the pangenome table, a pie chart of the breakdown of genes and the number of isolate they are present in, a graph with the frequency of genes versus the number of genomes.

[13]

There are some main parameters that need to be specified to roary, specifically, the input .gff files; the output directory -f roary_out; the , -i parameter, specifing the percentage identity of blastp, here used at 95%; the , -cd parameter, percentage of isolates a gene must be in to be considered part of the core genome, here setted at

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90%.

```
roary .gff -i 95 -cd 90
```

With Roary it is also possible to perform a core gene alignment to generate a more reliable tree. A core-genome alignment is more scalable with respect to the whole-genome alignment. It is an alignment useful to identify the core genes conserved in all aligned genomes and that can be very useful to infer the phylogeny [14].

The main parameters sto be specified are the <code>-e</code> parameter, needed to perform a core gene alignment; the <code>-n</code> parameter, to use mafft as the tool for the multiple sequence alignmet, making the process faster and finally the parameter <code>-p</code>, needed to specify the number of threads, increasing therefore the speed [11].

```
roary .gff -i 95 -cd 90 -e -n -p 8.
```

The core gene alignment can be used to construct a phylogenetic tree. This is possible using FastTree, a tool for constructing large phylogenies, estimating their reliability. FastTree exploit Neighbor-Joining and nearest neighbor interchanges to create a phylogenetic tree. [15] Specifically, we used FastTreeMP, that allow the parallelization of the steps needed in computing a tree [16]. The tree is obtained using the following code,

```
FastTreeMP -gtr -nt -out core_gene.tre core_gene_alignment.aln,
```

in which the parameter -gtr express the generalized time-reversible model (to be used with nucleotide alignments only) while the parameter -nt is used to specify that the alignment is performed on nucleotides.

2.1.3 Taxonomic assignment (PhyloPhlAn 3.0)

PhyloPhlAn 3.0 is an accurate and rapid tool to perform microbial genome characterization and phylogenetic analysis both of newly assembled microbial genomes and metagenomes. PhyloPhlAn 3.0 can integrate public genome resources/information to the genomes in input and is also accurate at the strain and species level and allow the assign to each bin obtained via metagenomic assembly its closest species-level genome bins [17]. There are some main parameters needed to be specified, mainly, the input folder with the <code>-i</code> parameter; the output folder, with the parameter <code>-o</code>; the <code>--nproc</code> parameter, used to specify the CPUs that can be used; the <code>-n</code> parameter that allow us to decide how many SGBs (sorted by increasing average genomic distance) will be reported for each input bin in the output file; the <code>--database_update</code> parameter to update the databases file, the <code>-d</code> parameter to specify the name of the output database adn finally, the <code>--verbose</code> parameter to print to the bash [18]. The final command is the following:

3 Results and discussion

3.1 uSGB 15132

We were provided with a set of 31 high-quality prebinned metagenomes grouped in the same uSGB labelled SGB15132. As shown in the Figure 3.1 the bins have a completeness higher that 97.3 and the maximum redundacy registered is equal to 2.25. The MAGs were associated to a study condition, as shown in the Figure 3.2, we used this information to interpret the analysis that were effectuated later on.

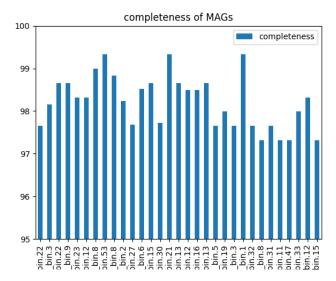


Figure 3.1: Completeness distribution of the given MAGs.

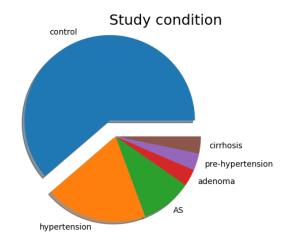


Figure 3.2: Pie chart of the study condition of our MAGs $\,$

3.2 Genome annotation

After the gene annotation process, possible using prokka, we were able to identify that the number of the CDS (protein coding sequence) is slightly variable, spanning from a minimun value of 2651 to a max of 3935. For each MAG, as shown in the Figure 3.3 more or less a half of the CDS are known proteins, while the other half is represented by RNA or hypotetical proteins.

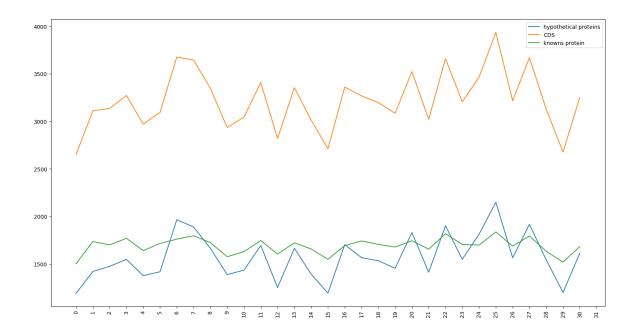


Figure 3.3: Main annotation results

3.3 Pangenome analysis

Each SGB strain was found to contain an average of 1317 genes that are present in every strain (core genome), plus 8407 genes that are absent in more than one strains (strain < 31) (accessory genome). This can be seen in the Figure 3.4, showing the frequency plot of the genes per genome, this plot gives a general overview of the frequency of genes within a whole genome set, typically these plots have an shape and most genes can be detected in a single genome or in all genomes [19]. Here it is shown that the number of genes present in all the genomes correspond to the number of core genes. The accessory genes are also divided into genes present in only one strain (cloud genome) or genes present in two or more strains but not all strains (shell genome) [9]. The figure 3.5 show the subparts constituting the pangenome, associated with the total number of genes while the Figure 3.6 show the tree obtained via the analysis associated to heatmap in which we can easily define the core genes that are present in all the genomes. It is important to remember that the tree is not obtained with a complete clustering methodology.

Looking at the Figures 3.7 and 3.8 we can obtain/derive an idea on the pangenome, specifically if it is open or closed. It is important to remember that these results were obtained using only 31 metagenomes, making the process of discussion more complex. Even though, looking at the Conserved vs Total genes plot in Figure 3.7, we can argue that the total genes initial slope is very high but at the end, when almost all the genomes were added, the slope start to decrease. The conserved gene line is almost stationary, with little changes when the genomes are all added. Only looking at this graph it is difficult to define if the pangenome is open or closed, therefore, the observation of the Unique vs New genes plot can be useful. Looking at the Figure 3.8, we can see that, adding new genomes doesn't give new information, the number of new genes is almost at zero when we add the last genome. This help us to state with a good security that the pangenome is closed.

3.4 Phylogenetic analysis and association with host data

PhyloPhlAn enable us to operate a phylogenetic analysis. The main output correspond to a list of the closest SGBs sorted by their increasing average sequence distance (Mash distance) in a tab-separated file [18]. The information of

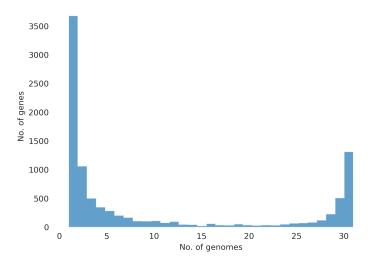


Figure 3.4: Pangenome Frequency

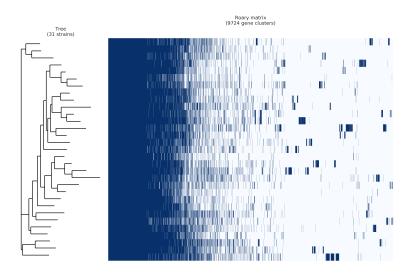


Figure 3.5: Pangenome matrix

each SGB contains specific columns:

my_bin: is the input bin name

(k|u)SGB_ID: the SGB ID, k indicate a known SGB, u indicate an unknown SGB

taxa_level: taxonomic level the SGB has been assigned to taxonomy: the full taxonomic label assigned to the SGB

average_mash_distance: the distance is calculated with respect to all the genomes in the SGB.

In our case, for each MAG we get the same taxa_level, the species Flavonifractor plautii, a strictly anaerobic rod shaped bacterium. It is a commensal of the human intestinal microbiota which is usually difficult to isolate from samples [20].

DA DIRE ALTRO?

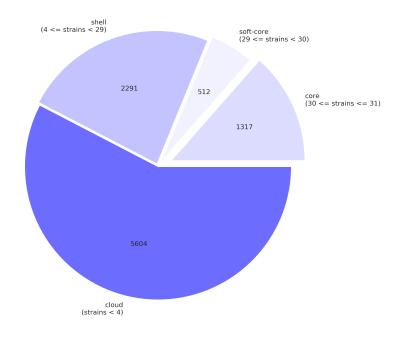


Figure 3.6: Core and accessory genome

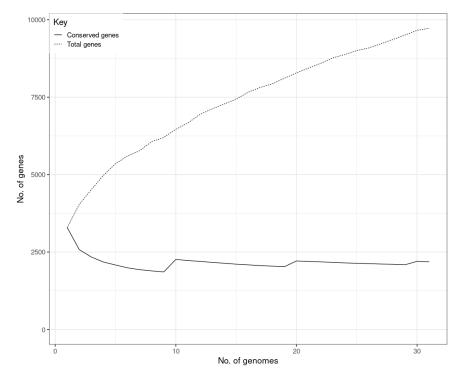


Figure 3.7: Conserved vs Total genes

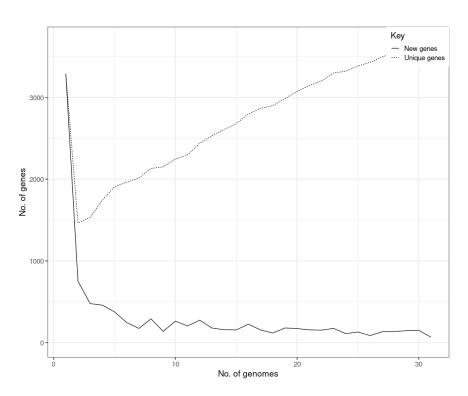


Figure 3.8: Unique vs New genes

4 Conclusion

Thanks to this project it was possible to obtain an insight on the metagenome workflow and the main analysis that are need when dealing with unkown genomes. We initially investigated the uSGB we were given computing general statistics on the quality and the associated metadata. We then performed a process of gene annotation and retrived informations about the number of coding sequences and hypotetical proteins in these CDS (Figure 3.3). We than were able to perform a pangenome analysis and hypotize that the pangenome of our uSGB is closed as the addition of new genomes gives us little to no information as shown in the Figure 3.8. Also, it was possible to identify the genes shared by at least ??TO CHECK?? 90% of our MAGs, the core genome. These correspons to $\sim 15\%$ of the total gene. From the taxonomix assignment of the uSGB it was possible to identify the Oscillospiraceae family and specifically the Flavonifractor plautii species. Thanks to further invastigation on this bacteria it was found that this strain was recently obtained/created via the unification of two previuosly distinct species, specifically, Clostridium orbiscindens and Eubacterium plautii [21].

It is known from literature that this species is correleated with hip joint infection [22], chronic kidney disease, varoisu autoimmune disorders and epithelial invasive potential [23]. Finally, this strain was associated with blood infections in immunosuppressed patients [24].

We think that this foundings are to keep in consideration for further analysis on this strain, since the 33% of our samples were obtained from patients with correlated diseases like rheumatoid arthritis (a kind hip joint infection) and hypertension.

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