

Protocol_bio217_Hendrik Bethge

1. Metagenome assembled Genomes (MAGs)

1.1 Background information

The goal of this project was to learn how to handle and interpret metagenomic data from the raw read to the assembled MAGs.

This process consists of five steps:

1. Pre-processing of the raw reads
2. Assembling the reads into contigs/fasta
3. Assessing the quality of assemblies
4. Binning the contigs into MAGs
5. Assessing completeness, contamination and strain heterogeneity of the MAGs

Dataset

The samples were taken from a biogas plant near Cologne over the course of 587 days.

Samples were taken once a month and analysed based on 16S amplicon sequences by Martin Fisher (<https://sfamjournals.onlinelibrary.wiley.com/doi/full/10.1111/1751-7915.13313>).

In addition to the amplicon sequencing, the samples were also sequenced for metagenomics.

In this protocol three exemplary samples from this dataset were analysed.

HPC system

All steps were run on the HPC-server of the CAU (CAU-cluster) via **Bash scripts** unless specified otherwise.

To access the HPC-server:

```
ssh -X sunam???@caucluster.rz.uni-kiel.de
```

The scripts were written in **Sublime Text** and executed through a Linux Shell.

```
sbatch <script.sh>
```

A script should contain the following information:

1. The shebang
2. Processing requirements as #SBATCH commands
 - Reservation
 - Nodes to use
 - CPUs (for multithreading)
 - memory requirements
 - time

- working directory
 - log files
 - partitions
3. Stdin and Stderr paths
 4. Slurm modules needed for the task
 5. The command
 6. jobinfo

Example bash script:

```
#!/bin/bash
#SBATCH --nodes=1
#SBATCH --cpus-per-task=8
#SBATCH --mem=32G
#SBATCH --time=01:00:00
#SBATCH --job-name=example
#SBATCH --output=example.out
#SBATCH --error=example.err
#SBATCH --partition=all
#SBATCH --reservation=biol217

#activate environment
module load miniconda3/4.7.12.1
conda activate /home/sunam226/.conda/env/anvio
#Commandlines
example.command
#this prints the required resources into your logfile
jobinfo
```

For the course we were able to access the HPC with special permissions through the reservation command. Unless specified otherwise all commands were run in a **miniconda environment**.

Tools used:

Tool	Version	Repository
fastqc	0.11.9	FastQC
fastp	0.22.0	fastp
megahit	1.2.9	megahit
samtools	1.9	samtools
QUAST	5.0.2	quast
Bowtie2	2.4.5	bowtie2
Concoct	1.1.0	CONCOCT
MetaBAT2	2.12.1	Metabat2

Tool	Version	Repository
DASTool	1.1.5	DAS_Tool
anvi'o	7.1	anvi'o
GUNC	1.0.5	GUNC

1.2 Pre-processing of the raw reads

The raw reas were copied into the working directory

```
mkdir /work_beegfs/sunam236/day2

cp /home/sunam226/Day2/0_raw_reads/*.fastq.gz /work_beegfs/sunam236/day2
cd ./day2
```

The quality of the reads was analysed with **FastQC**, based on the phred quality score.

```
for i in *.gz; do fastqc $i -o output_folder/; done
```

The sequences were trimmed with **Fastp**

```
for i in `ls *_R1.fastq.gz`;
do
    second=`echo ${i} | sed 's/_R1/_R2/g'`
    fastp -i ${i} -I ${second} -R _report -o ../clean_reads/"${i}" -O
    ../clean_reads/"${second}" -t 6 -q 20
done
```

1.3 Assembling the reads into contigs/fasta

The processed reas were assembled into contigs using **Megahit**.

```
cd /work_beegfs/sunam236/day2/clean_reads

megahit -1 BGR_130305_R1.fastq.gz -1 BGR_130527_R1.fastq.gz -1
BGR_130708_R1.fastq.gz -2 BGR_130305_R2.fastq.gz -2 BGR_130527_R1.fastq.gz -2
BGR_130708_R2.fastq.gz --min-contig-len 1000 --presets meta-large -m 0.85 -o
../megahit -t 20
```

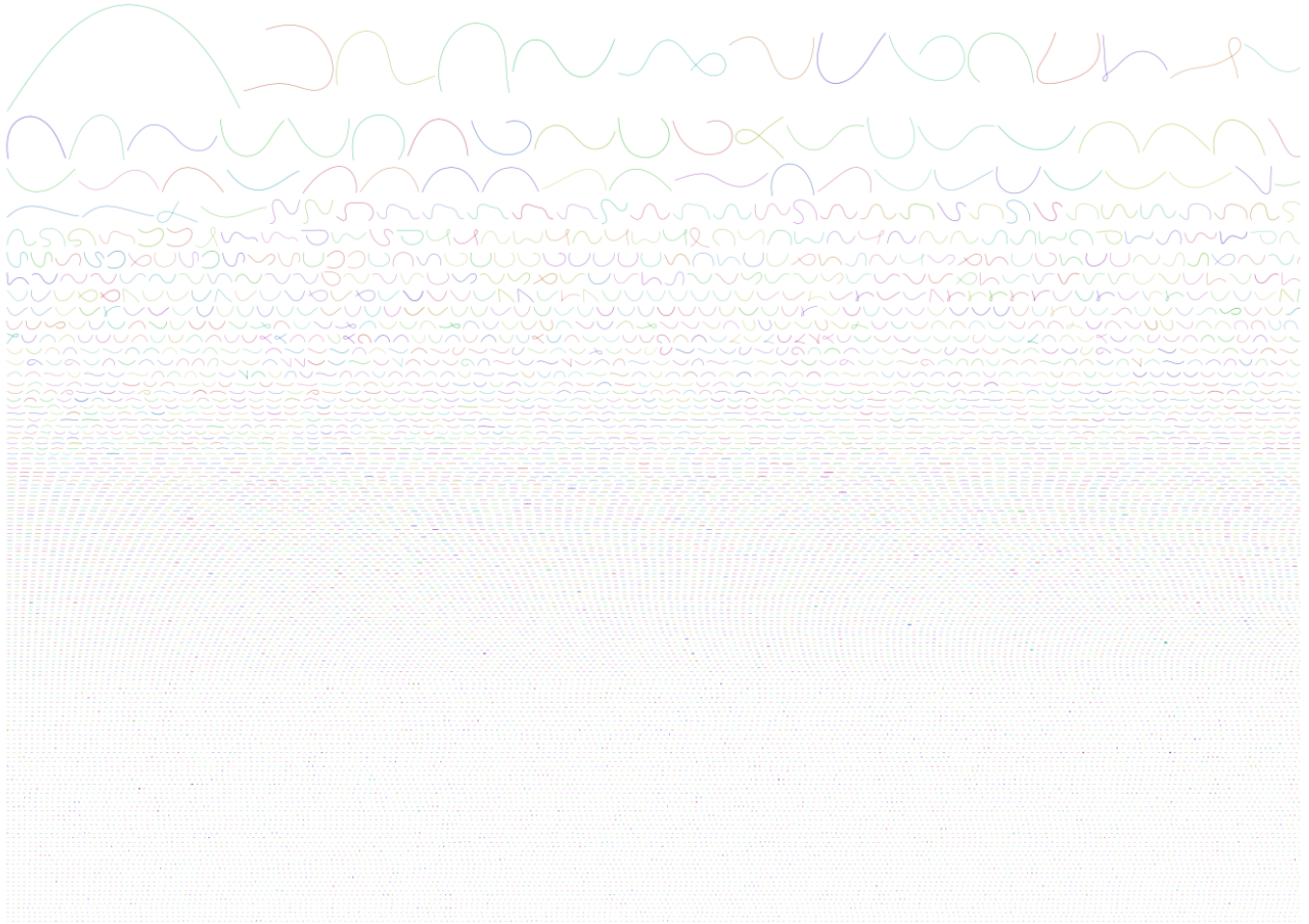
The contigs were transformed into FASTG files to visualized in **Bandage**.

```
cd /day3/3_coassembly/  
  
megahit_toolkit contig2fastg 99 final.contigs.fa > final.contigs.fastg
```

Opened Bandage and loaded the .fastg file

Question 1:

Submit your generated figure and explain shortly what you can see



The picture shows the de novo created contigs from the processed reads

- The length of each line represents the length of the contig sequence
- Loops represent k-mers with multiple continue options
- The colour is selected based on assumed MAG assignment (bins)
- The smaller fragments are reads, that were not able to be assigned to any of the longer contigs

1.4 Assessing the quality of assemblies

The quality assessment of the megahit results was performed with **Quast**. The output is given as PDF and html.

```
cd /work_beegfs/sunam236/3_coassembly/  
  
metaquast -t 6 -o ../3_metaquast/ -m 1000 final.contigs.fa
```

This step did not work due to a permission error, so the output was copied from a backup

Question 2:

What is your N50 value?

- 2963

How many contigs are assembled?

- 57414

Whats is the total length of the contigs?

- 145675865 Bp = 145,7 Mb

1.5 Binning contigs into MAGs

Binning of the contigs:

To make the next steps easier the contig names were shortened. *This step was performed directly in the console, without a bash script*

```
anvi-script-reformat-fasta final.contigs.fa -o
/work_beegfs/sunam236/day3/contigs.anvio.fa --min-len 1000 --simplify-names--
report-file name_conversion.txt
```

Afterwards the bins were mapped in a loop, so every final.contigs.fasta recieved a corresponding mapping file.

```
module load bowtie2
cd ./2_fastp/

for i in `ls *mapped_R1.fastq.gz`;
do
    second=`echo ${i} | sed 's/_R1/_R2/g'`
    bowtie2 --very-fast -x ../4_mapping/contigs.anvio.fa.index -1 ${i} -2
    ${second} -S ../4_mapping/"$i".sam
done
```

The output are sequence mapping files (.sam), which were converted to binary alignment and map files (.bam).

```
module load samtools

cd ./4_mapping/

for i in *.sam; do samtools view -bS $i > "$i".bam; done
```

1.5.1 Anvi'o

From here on **anvi'o** commands were used.

The **An**alysis and **Vi**sualization platform for microbial **'O**mics combines many of the computational strategies of data-enabled microbiology.

A contigs.db file was created, which contains key information associated with the sequences.

This command computes k-mer frequencies for each contig, soft-split contigs larger than 20000 bp into smaller ones and identifies open reading frames using Prodigal.

```
cd /work_beegfs/sunam236/day3/  
  
anvi-gen-contigs-database -f ./4_mapping/contigs.anvio.fa -o  
./5_anvio_profiles/contigs.db -n 'biol217'
```

Afterwards a Hidden Markov Models (HMM) was run on the contigs, searching for specific single copy core genes (SCGs) with known functions.

```
anvi-run-hmms -c ./5_anvio_profiles/contigs.db
```

The result was visualized with anvi'o interactive (directly in the console):

```
srun --pty --mem=10G --nodes=1 --tasks-per-node=1 --cpus-per-task=1 --  
partition=all /bin/bash
```

Note the accessed node!

```
conda activate  
/home/sunam225/miniconda3/miniconda4.9.2/usr/etc/profile.d/conda.sh/envs/anvio-7.1  
  
anvi-display-contigs-stats contigs.db
```

Open a new terminal, not logged into HPC

```
ssh -L 8060:localhost:8080 sunam236caucluster-old.rz.uni-kiel.de  
  
ssh -L 8080:localhost:8080 node???
```

Open <http://127.0.0.1:8060/> in your browser

1.5.2 Binning with Anvio

Sorted and indexed the .bam files with **samtools** in anvi'o.

```
cd ./4_mapping/

for i in *.bam; do anvi-init-bam $i -o ../5_anvio_profiles/"$i".sorted.bam; done
```

An anvio profile was created to store sample-specific information about the contigs in a single profile. It processed only contigs >2500 nt.

Processing includes:

- Recovery of mean coverage, SD of coverage and average coverage for the inner quartiles.
- Characterization of single nucleotide variants (SNVs)

```
cd ./5_anvio_profiles

for i in `ls *.sorted.bam | cut -d "." -f 1`; do anvi-profile -i
"$i".fastq.gz.sam.bam.sorted.bam -c contigs.db -o ../6_profiling/$i; done
```

This created a folder containing profiles.db and a .txt log.

In the next steps all profiles of the different samples were merged into one:

```
cd ./6_profiling/

anvi-merge ./BGR_130305_mapped_R1/PROFILE.db ./BGR_130527_mapped_R1/PROFILE.db
./BGR_130708_mapped_R1/PROFILE.db -o ./ -c ../5_anvio_profiles/contigs.db --
enforce-hierarchical-clustering
```

In this course two different Binner were used: **Metabat2** and **Concoct**. The result of both Binner were consolidated using **DASTool**. *The binning steps did not work, so the files were copied from a back up!*

Binning with Metabat2

```
cd ./6_profiling/

anvi-merge ./BGR_130305_mapped_R1/PROFILE.db ./BGR_130527_mapped_R1/PROFILE.db
./BGR_130708_mapped_R1/PROFILE.db -o ./ -c ../5_anvio_profiles/contigs.db --
enforce-hierarchical-clustering
```

Binning with CONCOCT

```
anvi-cluster-contigs -p /PATH/TO/merged_profiles/PROFILE.db -c /PATH/TO/contigs.db
-C consolidated_bins --driver dastool -T 20 --search-engine diamond -S
```

```
METABAT,CONCOCT --log-file log_consolidation_of_bins --just-do-it

anvi-summarize -p /PATH/TO/merged_profiles/PROFILE.db -c /PATH/TO/contigs.db -o
/PATH/TO/SUMMARY_consolidated_bins -C consolidated_bins
```

Consolidating bins with DASTool

```
anvi-cluster-contigs -p /PATH/TO/merged_profiles/PROFILE.db -c /PATH/TO/contigs.db
-C consolidated_bins --driver dastool -T 20 --search-engine diamond -S
METABAT,CONCOCT --log-file log_consolidation_of_bins --just-do-it

anvi-summarize -p /PATH/TO/merged_profiles/PROFILE.db -c /PATH/TO/contigs.db -o
/PATH/TO/SUMMARY_consolidated_bins -C consolidated_bins
```

From here the previously mentioned copied files were used to answer the questions and proceed with the quality estimation.

Question 3:

- Number of Archaea bins from MetaBAT2:
3
- Number of Archaea bins from CONCOCT:
2
- Numer of Archaea bins from consolidating:
2

1.5.3 MAGs quality estimation

To see the amount of bins

```
anvi-estimate-genome-completeness -p PROFILE.db -c contigs.db --list-collection
```

Visualized it through anvi-interactive (directly in the console).

```
srunk --pty --mem=10G --nodes=1 --tasks-per-node=1 --cpus-per-task=1 --
partition=all /bin/bash
```

Note the accessed node!

```
conda activate
/home/sunam225/miniconda3/miniconda4.9.2/usr/etc/profile.d/conda.sh/envs/anvio-7.1

anvi-interactive -p ./PROFILE.db -c ./contigs.db -C ../../consolidated_bins
```

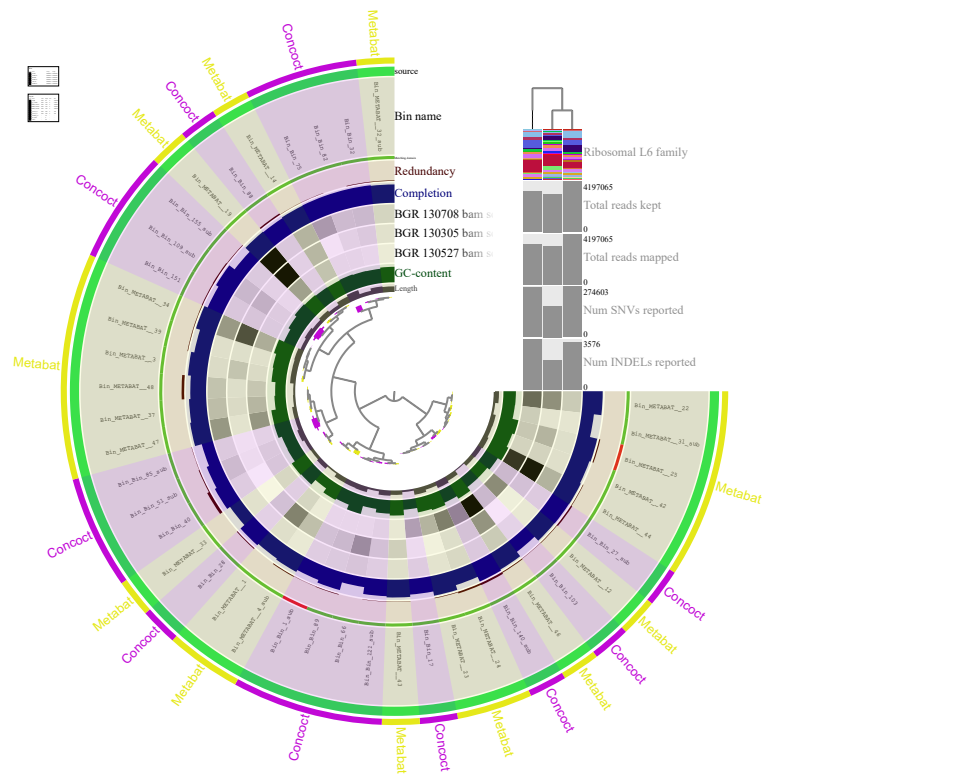

open a new terminal, not logged into CAUcluster

```
ssh -L 8060:localhost:8080 sunam236@caucluster-old.rz.uni-kiel.de  
ssh -L 8080:localhost:8080 node010
```

Open in a Browser: <http://127.0.0.1:8060/>

Collection 'consolidated bins' for merged profiles

Items order: Mean coverage (D. Euclidean; L. Ward) | Current view: mean_coverage | Samples order: custom



Matching domain

bacteria (1.0) (16) bacteria (0.9) (11) bacteria (0.8) (5) bacteria (0.7) (3) bacteria (0.6) (3) archaea (0.9) (1) archaea (0.5) (1) bacteria (0.5) (1)

Bin name

Bin_Bin_103 (1) Bin_Bin_109_sub (1) Bin_Bin_121_sub (1) Bin_Bin_140_sub (1) Bin_Bin_151 (1) Bin_Bin_155_sub (1) Bin_Bin_17 (1) Bin_Bin_1_sub (1)

Bin_Bin_27_sub (1) Bin_Bin_28 (1) Bin_Bin_32 (1) Bin_Bin_40 (1) Bin_Bin_51_sub (1) Bin_Bin_66 (1) Bin_Bin_75 (1) Bin_Bin_82 (1) Bin_Bin_85_sub (1)

Bin_Bin_88 (1) Bin_Bin_89 (1) Bin_METABAT_1 (1) Bin_METABAT_12 (1) Bin_METABAT_14 (1) Bin_METABAT_19 (1) Bin_METABAT_22 (1) Bin_METABAT_23 (1)

Bin_METABAT_24 (1) Bin_METABAT_25 (1) Bin_METABAT_3 (1) Bin_METABAT_31_sub (1) Bin_METABAT_32_sub (1) Bin_METABAT_33 (1) Bin_METABAT_34 (1)

Bin_METABAT_37 (1) Bin_METABAT_39 (1) Bin_METABAT_42 (1) Bin_METABAT_43 (1) Bin_METABAT_44 (1) Bin_METABAT_46 (1) Bin_METABAT_47 (1)

Bin_METABAT_48 (1) Bin_METABAT_4_sub (1)

Source

dastool (41)

DTU013 sp002385815 (36.8185986599562) DTU030 sp001513125 (3.921875) Sphaerochaeta sp004294855 (3.63671875) Heriborax sp012517995 (6.12796815507096)

Lenti-01 sp002304915 (16.7680505555553) DTU029 sp001512435 (3.78879310344828) JAAYTL01 sp012518035 (7.4995605501012) DTU067 sp001512995 (16.06365063918256)

Paludibacter sp002070265 (12.4985931772575) UBA1402 sp002305085 (12.36400511018845) JAAYEV01 sp012728565 (7.989707889707891) Unknown_genus_Thermoclostridium_30944 (24.944980742517792)

JAAYQA01 sp012519495 (4.8789209928744794) DTU059 sp012523685 (4.7537037037037) UBA4923 sp012841015 (9.266666666666666) Acetomicrobium sp012518015 (12.697994987468679)

Unknown_genus_UBA4923_4080 (18.601342525399133) JAAZMG01 sp012798935 (2.911111111111111) Geofilum sp002411385 (8.82654166243905) Methanococcus bourgenis (6.126142137200454)

DTU012 sp012837335 (14.1262618093325) UBA5420 sp012518325 (5.56818740029147) UBA3900 sp002391675 (91.6623344340136) DTU018 sp002305335 (3.04972375690607)

DTU012 sp012520795 (9.63177841223661) UBA9076 sp003557975 (3.26792722586574) DTU069 sp012738095 (5.3091510740041295) Heribinx sp012518075 (10.957410216381199)

DUNF01 sp012839485 (2.226053639846744) Methanococcus sp012797575 (16.07302141368665) Sedimentibacter sp012520455 (5.8845341609766) UBA4179 sp0023240405 (18.2282513661203)

UBA4923 sp012841145 (46.6636611574563) DTU059 sp012523705 (12.06126653738659) JAAZJW01 sp012519835 (11.74253654745671) UBA5420 sp012838305 (3.89716312056738)

Heribinx sp012837605 (4.83089770354906) UBA5455 sp012518415 (7.01762452107279) UBA5436 sp012522435 (29.977330279017988) Peritomonas mucosa (1.592920339823)

Methanosarcina flavescens (6.905140545441871) UBA3906 sp002391555 (56.3795961674341) JAAZLA01 sp012799545 (9.6756186612576) JAAZIN01 sp012510625 (4.92930089359465)

DTU049 sp01512885 (5.7820025560825) UBA1361 sp002306335 (48.725792931369156) UBA3946 sp002385755 (11.8512440265695) UBA4179 sp0023281125 (33.14622347922728)

UBA3950 sp002385475 (0.936482229882325) DUQW01 sp012800005 (1) Fermentomonas sp012838585 (3.53873873873874) JAAZSA01 sp012518455 (44.13619492733992)

UBA1426 sp002385825 (7.358974358974359) UBA3910 sp002391465 (52.114699445369766) Methanococcus thermolyticus (10.726257287462358) JAAZIN01 sp012798795 (5.15258215962441)

DTU010 sp012837385 (5.475929259593) UBA3906 sp002391555 (56.3795961674341) JAAZLA01 sp012799545 (9.6756186612576) JAAZIN01 sp012510625 (4.92930089359465)

JAAZPQ01 sp012797175 (20.34923905866428) Heribinx sp012523775 (1.7340813648294) Paludibacter sp012519425 (1.74683344037077) DTU010 sp002318315 (13.47249744975099)

UBA5453 sp002427375 (14.774360009871849) Fermentomonas sp012839475 (15.7470119521912) Amphibacillus sp012838505 (2.834723837508997) DUQW01 sp012520715 (5.935405753147687)

DTU033 sp01513365 (3.09107468123862) UBA4971 sp000019985 (6.98727177882108) DUQW01 sp012837585 (32.37337886058821) UBA1046 sp012519085 (3.7457627118644004)

UBA3947 sp00351575 (8.1277122519502) DUPA01 sp012838515 (5.96611721611721) JAAYPV01 sp012523875 (2.957875457875461)

Ribosomal_L6_species :: Ribosomal L6 species

Ribosomal_L6_genus :: Ribosomal L6 genus

Ribosomal_L6_family :: Ribosomal L6 family

Ribosomal_L6_order :: Ribosomal L6 order

Ribosomal_L6_class :: Ribosomal L6 class

Ribosomal_L6_phylum :: Ribosomal L6 phylum

Ribosomal_L6_domain :: Ribosomal L6 domain

Question 4:

- Which binning strategy gives the best quality archae bins:
dastool (consolidated) gives the best quality, which was expected since it is not a binner itself and just merges the results of the two binners. Metabat2 seems to be better than CONCOCT for our data.
- How many archaea bins do you get of high quality.
2

The binning quality for all three was checked. Result was saved in working directory

```
anvi-estimate-genome-completeness -c ./contigs.db -p ./PROFILE.db -C consolidated_bins > genome_completeness_dastool

anvi-estimate-genome-completeness -c ./contigs.db -p ./PROFILE.db -C METABAT > genome_completeness_metabat2

anvi-estimate-genome-completeness -c ./contigs.db -p ./PROFILE.db -C CONCOCT > genome_completeness_concoct
```

1.5.4 Bin refinement

From this point only the archae bins were used.

A summary folder was created with *anvi-summarize* containing a comprehensive overview of the collection and statistics created by anvio.

```
anvi-summarize -c ./contigs.db -p ./5_anvio_profiles/merged_profiles/PROFILE.db -C consolidated_bins -o ./5_anvio_profiles/summary --just-do-it
```

The archae bins were copied to a separate folder

```
cd ./5_anvio_profiles/summary/bin_by_bin

mkdir ../../archaea_bin_refinement

cp ./Bin_Bin_1_sub/*.fa ../../archaea_bin_refinement/

cp ./Bin_METABAT__25/*.fa ../../archaea_bin_refinement/
```

1.6 Assess completeness, contamination and strain heterogeneity of the MAGs

1.6.1 Chimera detection

Genome **UNC**cluttered (**GUNC**) was used to detect chimeric MAGs.

```
conda activate /home/sunam226/gunc

cd ../../archea_bin_refinement
mkdir GUNC

cd /work_beegfs/sunam236/Day5/5_anvio_profiles/archea_bin_refinement

for i in *.fa; do gunc run -i "$i" -r /home/sunam226/Databases/
gunc_db_progenomes2.1.dmnd --out_dir GUNC --threads 10 --detailed_output; done
```

To see if a MAG is chimeric or not (clade separation score) the output file (.tsv) was opened.

Question 5:

Do you get Archea bins that are chimeric:

clade separation index close to one = chimeric

- metabat archea: not chimeric
 - species level =0.9
- Concoct archea: chimeric
 - kingdom, phylum, class = chimeric
 - After class it isn't marked as chimeric, because the groups have not been named yet

Explain what a chimeric bin is in your own words

A chimeric bin contains a MAG that is created from more than one genome. So a mixture of at least two different species.

1.6.2 Manual bin refinement

Only for the non chimeric bins.

The redundant parts were removed through the interactive anvio interface, by pre-selecting the better MAGs (<70% completeness).

Before starting, a copy/backup of the unrefined bins was created in ARCHEA_BIN_REFINEMENT. Because the manual refinement will overwrite the unrefined files.

Run directly in the console:

```
srunk --reservation=biol217 --pty --mem=10G --nodes=1 --tasks-per-node=1 --cpus-per-task=1 /bin/bash
```

Note the accessed node

```
anvi-refine -c ./4_mapping/contigs.db -C consolidated_bins -p
./5_anvio_profiles/merged_profiles/PROFILE.db --bin-id Bin_METABAT__25
```

Open a new terminal, not logged into CAUcluster

```
ssh -L 8060:localhost:8080 sunam236@caucluster-old.rz.uni-kiel.de  
ssh -L 8080:localhost:8080 node010
```

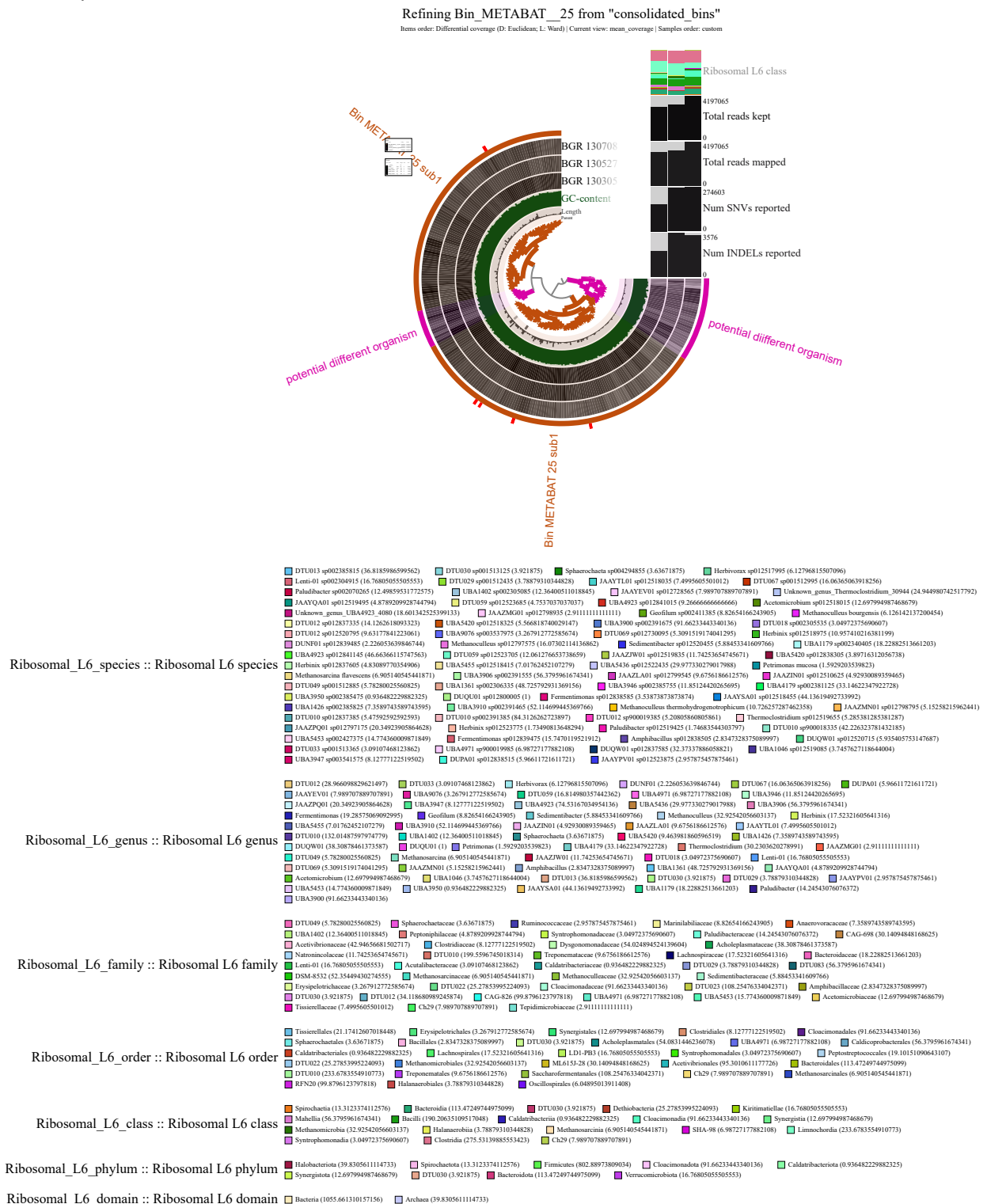
Open in a browser: <http://127.0.0.1:8060/>

Question 6:

- Does the quality of your Archea improve?:
 - Before: Comp.:97.4 Red.:5.3
 - After: Comp.:93.4 Red.:5.3
 - after removing some bins clustered by difference, the composition reduced but the redundancy stayed the same. So visually the quality increases, but the redundancy score did not reduce.
- Explain what a chimeric bin is in your own words

A chimeric bin contains a MAG that is created from more than one genome. So a mixture of at least two

different species.



Question 7:

- How abundant are the archaea bins in the 3 samples
 - Metabat: 1.76 | 1.14 | 0.58
 - Concoct: 0.96 | 0.00 | 0.40
- you can also use `anvi-inspect -p -c, anvi-script-get-coverage-from-bam or, anvi-profile-blitz`. Please look up the [help page](#) for each of those commands and construct the appropriate command line

```
anvi-inspect -p ./5_anvio_profiles/merged_profiles/PROFILE.db -c
./4_mapping/contigs.db --split-name
```

1.6.3 Taxonomic assignment

Taxonomic annotations were added to the MAGs through the single copy core genes (SCGs):

```
anvi-run-scg-taxonomy -c ./4_mapping/contigs.db -T 20 -P 2

anvi-estimate-scg-taxonomy -c ./4_mapping/contigs.db --metagenome-mode
```

The abundance of rRNAs in the dataset was estimated:

```
anvi-estimate-scg-taxonomy -c ./4_mapping/contigs.db -p
./5_anvio_profiles/merged_profiles/PROFILE.db --metagenome-mode --compute-scg-
coverages --update-profile-db-with-taxonomy
```

The output was saved from the terminal:

```
anvi-estimate-scg-taxonomy -c ./4_mapping/contigs.db -p
./5_anvio_profiles/merged_profiles/PROFILE.db --metagenome-mode --compute-scg-
coverages --update-profile-db-with-taxonomy > temp.txt
```

Received one final summary of all the info from the consolidated bins

```
anvi-summarize -c ./contigs.db -p ./5_anvio_profiles/merged_profiles/PROFILE.db -C
consolidated_bins -o ./5_anvio_profiles/summary2 --just-do-it
```

Renamed bins_summary.txt to .tsv and opened with libre office

2. Pangenomics - comparing genomes

2.1 Evaluation of the starting databases

For pangenomics a new set of contigs.dbs were used. They contained already prepared MAGs from the Biogasreactor and a complete Methanogen genome.

Files that were created for us in Day6:

- 02_contigs-dbs
- 03_pangenome

To compare the bins a summary overview was visualized.

For this direct access to a HPC compute node was acquired.

Enter directly in the console:

```
srun --reservation=biol217 --pty --mem=10G --nodes=1 --tasks-per-node=1 --cpus-per-task=1 /bin/bash
```

Note the accessed node and start anvi'o interactive display

```
cd /02_contigs-dbs/  
anvi-display-contigs-stats *db
```

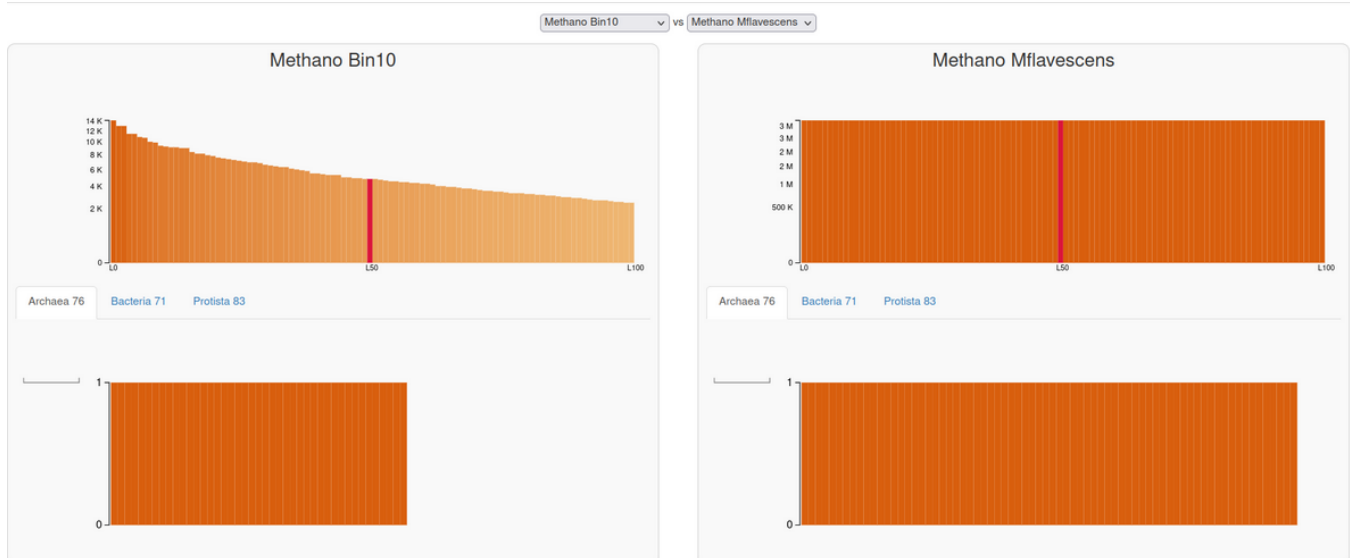
Open a new terminal, not logged into CAUcluster

```
ssh -L 8060:localhost:8080 sunam236@caucluster-old.rz.uni-kiel.de  
ssh -L 8080:localhost:8080 node010
```

Open in a Browser: <http://127.0.0.1:8060/>

Question 7:

How do the MAGs compare in size and number of contigs to the full genome?



- The bins are all smaller than the full *Methano_mflavescens* genome. They differ from 1.3 mbp to 2.6 mb
- The contig number is much higher, since the full genome only consists of 1 contig. Most of the contigs are > 2.5 kb
- The longest contigs vary between 14 kb and 65 kb (compared to 3.2 mb of the full genome)
- HMM (single copy core genes) differ from 43 to 80. The complete genome contains 72.

Based on the contigs number, size and number of marker genes (HMM hits), which two MAGs are the best and which are the worst?

The worst MAGs are:

- 10
 - Contains a lot of Contigs smaller than 2.5 kb
 - The least HMM archaea hits
 - Shortest total length
- 5
 - No approx. genome found
 - To few HMM hits
 - Alot of contigs below 2.5 kb
 - The best MAGs are:
- 13
 - Most contigs above 20kb
 - Most archaea HMM hits
- 9
 - Most contigs above 10 kb
 - Longest total length
 - Contains no approx. bacterial genome

2.2 Create a pangenome

A Pangenome visualizes entire genomes for comparison. The Comparison is based on essential and accessory gene clusters, phylogenetic relationships and genome qualities.

2.2.1 Create a external genome file

First a "external genome file" needed to be created to input which genomes and MAGs should be used, as a table with two columns:

name	contigs_db_path
Bin1	/path/to/contigs-bin-01-120311.db
Genome-name	/path/to/contigs-genome-name.db

A anvio script can create the input information.

```
cd /03_pangenome/
anvi-script-gen-genomes-file --input-dir ../02_contigs-dbs/ -o
external_genomes.txt
```

Question 8:

Use the **head** or **cat** command to verify if it worked

```
cat external_genomes.txt

name contigs_db_path
Methano_Bin1 /work_beegfs/sunam236/Day6/02_contigs-dbs/Bin1.db Methano_Bin10
/work_beegfs/sunam236/Day6/02_contigs-dbs/Bin10.db Methano_Bin13
/work_beegfs/sunam236/Day6/02_contigs-dbs/Bin13.db Methano_Bin3
```

```
/work_beegfs/sunam236/Day6/02_contigs-dbs/Bin3.db Methano_Bin5
/work_beegfs/sunam236/Day6/02_contigs-dbs/Bin5.db Methano_Bin8
/work_beegfs/sunam236/Day6/02_contigs-dbs/Bin8.db Methano_Bin9
/work_beegfs/sunam236/Day6/02_contigs-dbs/Bin9.db Methano_Mflavescens
/work_beegfs/sunam236/Day6/02_contigs-dbs/Mflavescens.db
```

2.2.2 Etsimate genome completeness

The genome completeness was estimated to recieve information about the quality of the MAGs. The most important factors to assess the quality are **redundancy** and **completeness**

```
anvi-estimate-genome-completeness -e external_genomes.txt -o ./genome-
completeness.txt
```

Question 9:

How do the bins compare to isolate genomes? Would you remove one, based on the output of the completeness estimation?

- The bins should show a lower completion rate and higher redundancy compared to isolated genomes, since the are unlikely to be as pure and continous.
- All bins with a completion < 75% and redundancy > 10 should be removed
--> Methano_Bin10 + 5

2.2.3 Remove unwanted genomes

The previously detected unwanted MAGs (completion <70% & redundancy >10%) were directly removed from the folder. For this a new folder was created in the 02_contig-dbs directory called "discarded" and the unwanted MAGs were moved into it.

```
cd ../02_contigs-dbs/
mkdir discarded
mv Bin10.db Bin5.db ../discarded/
```

Afterwards proceeded with the "good" MAGs.

A new external-genomes output was created for the updated bins.

```
cd ../03_pangenome/
anvi-script-gen-genomes-file --input-dir ../02_contigs-dbs/ -o
external_genomes_final.txt
```

2.3 Create Pangenome database

In Anvi'o two artifacts needed to be created (similar to working with assemblies).

The first one was a `genomes-storage.db` file, which merges all genomes into one database, making it easier to work with and publish.

The database contains:

1. All genome fasta files
2. The gene annotations (HMMs, SCGs) added before
3. Any new annotations and genome comparisons made afterwards

The second file is the `pan-genome.db` (similar to the profile generated to annotate the bins).

The file contains:

1. Genome similarities based on gene amino acid sequences
2. Resolved gene cluster
3. Any post-analysis of gene clusters, downstream analysis and visualisations.

Both files were created within one script, with different computing requirements from before.

```
#!/bin/bash
#SBATCH --nodes=1
#SBATCH --cpus-per-task=10
#SBATCH --mem=500M
#SBATCH --time=00:05:00
#SBATCH --job-name=pangenome
#SBATCH --output=pangenome.out
#SBATCH --error=pangenome.out
#SBATCH --partition=all
#SBATCH --reservation=biol217

#load and activate anvio environment
module load miniconda3/4.7.12.1
source activate
/home/sunam225/miniconda3/miniconda4.9.2/usr/etc/profile.d/conda.sh/envs/anvio-7.1

# Commands
cd /work_beegfs/sunam236/Day6/03_pangenome/

anvi-gen-genomes-storage -e external_genomes_final.txt -o genomes_storage-
GENOMES.db

anvi-pan-genome -g genomes_storage-GENOMES.db --project-name pangenome --num-
threads 10
#this prints the required resources into your logfile
jobinfo
```

2.4 Compare the data phylogenetically (ANI)

Afterwards the genome similarity was calculated (the **A**verage **N**ucleotide **I**ntity (ANI) with `MUMmer` to align each genome).

This was used to measure how related the genomes were and if a new species was discovered.

Species boundary: 95-96% identity over a 90% genome coverage

Depending on the amount of genomes used, this can be very memory intensive, so more Memory needed to be assigned in the bashscript:

```
#!/bin/bash
#SBATCH --nodes=1
#SBATCH --cpus-per-task=10
#SBATCH --mem=600M
#SBATCH --time=00:02:00
#SBATCH --job-name=pangenome
#SBATCH --output=pangenome.out
#SBATCH --error=pangenome.out
#SBATCH --partition=all
#SBATCH --reservation=biol217

#load and activate anvio environment
module load miniconda3/4.7.12.1
source activate
/home/sunam225/miniconda3/miniconda4.9.2/usr/etc/profile.d/conda.sh/envs/anvio-7.1

# Commands
cd /work_beegfs/sunam236/Day6/03_pangenome/

anvi-compute-genome-similarity --external-genomes external_genomes_final.txt --
program pyANI --output-dir ./ANI --num-threads 10 --pan-db ./pangenome/Pangenome-
PAN.db

#this prints the required resources into your logfile
jobinfo
```

Once the genome similarities were calculated, the interactive interface was calculated

2.5 Visualize the pangenome

Direct access to a HPC node to anvio-display-pan.

Run directly in the console:

```
srunk --pty --mem=10G --nodes=1 --tasks-per-node=1 --cpus-per-task=1 --
reservation=biol217 --partition=all /bin/bash

conda activate
/home/sunam225/miniconda3/miniconda4.9.2/usr/etc/profile.d/conda.sh/envs/anvio-7.1

anvi-display-pan -p ./pangenome/Pangenome-PAN.db -g ./genomes_storage-GENOMES.db -
P 8080
```

Question 10:

Scroll to the top of the help and find out which INPUT FILES you need. Write the command and use the additional flag -P. What is the -P flag for?

- Pangenome-PAN.db
- genomes_storage-GENOMES.db
- The -P flag declares the port number used (default 8080)

```
anvi-display-pan -p ./pangenome/Pangenome-PAN.db -g ./genomes_storage-GENOMES.db -P 8080
```

2.6 Interpret and order the pangenome

In this course interpreting and ordering the pangenome was done in the interactive interface. It was performed according to three tasks:

Task 1: Genome similarity

1. Remove combined homogeneity, functional homogeneity, geometric homogeneity, max num parsimony, number of genes in gene cluster and number of genomes gene cluster has hits from the active view (by setting Height to 0)
2. Create a "Bin-highlight" including all SCGs and name it accordingly.
3. Cluster the genomes based on Frequency

Question 11:

Based on the frequency clustering of genes, do you think all genomes are related? Why?

All of them are loosely related. All SCGs are clustered in the same region and they all should be Archaea. But Bin9 seems to be closer related to *M. flavescens* than the other three bins. The other three are closer related to each other.

How does the reference genome compare to its closest bin?

Tip: Consider the genome depiction and layers above.

They are pretty similar. Bin9 has most of the gene clusters of the reference genome.

When looking at some example genecluster alignments, Bin9 and the reference share most of the time the same sequence. Even in the SCG region

What ranges are used to determine a prokaryotic species? How high can you go until you see changes in ANI?

What does the ANI clustering tell you about genome relatedness?

- The cut off for prokaryotic species is 95%.
 - So Bin9 and reference seem to be one species. The other four differ from the reference, but share a species with each other.
- Differences within the clustered bins appear at 99%.
 - 99.5%: Bin1 not related to the other
 - 99.6%: Bin13 related to 3 and 8, Bin8 and 3 loosely related
 - 99.7%: Only relatedness between Bin9 and reference
- The ANI clustering gives information about the approximate relatedness (higher percentage=higher approximate relatedness).

Task 2: Functional Profiling

1. Using the Search Function, highlight all genes in the KEGG Module for Methanogenesis

- a maximum of 1 times (Single copy gene)
- has a high variability in its functional homogeneity (max. 0.80)
--> This gene will be highly conserved, but has diversified in the AA make up

2. Highlight the found genes on the interface. Inspect one of the gene-clusters more closely (Inspect gene-cluster).

Question 13:

What observations can you make regarding the geometric homogeneity between all genomes and the functional homogeneity?

- 20 Genes match the search parameters. All of them are in the previously marked SCGs regions.
- The functions should not be impaired, since SCGs are highly conserved, due to their very important and essential functionality. The aminoacid structure differ, which is probably not effecting the polarity or structure in the protein
 - M.flavescens and Bin 9 differ from the other four. They share the same SNPs, deletions, or insetions compared to the other four bins

2.7 BlastKOALA (Bonus)

Besides Anvi'o there are a lot of tools available for metabolism investigations. BlastKOALA is one of them, which creates a metabolic profile of the genome based on KEGG database.

In this course BlastKOALA Results for this Methanogen were prepared for us, so we could reconstruct the pathway.

Question 14:

Can the organism do methanogenesis? Does it have genes similar to a bacterial secretion system?

- Yes it can do methanogenesis
 - Through CO₂, Acetate and Methanol
- Yes it contains 7 genes also used in bacterial secretion systme

03070 Bacterial secretion system (7)

K03072	1727
K03074	1726
K03076	2075
K03106	1569
K03110	1206
K03116	194
K03118	196, 197

3. RNA-Seq analysis

3.1 Download the Data

The dataset used came from a publication by [Prasse et al. 2017](#).

To download the data used in the paper the SRR number was used:

```
conda activate /home/sunam226/.conda/envs/grabseq
grabseqs -t 4 -m ./metadata.csv SRR4018514 SRR4018515 SRR4018516 SRR4018517
```

or download the entire project:

```
conda activate /home/sunam226/.conda/envs/grabseq
grabseqs -t 4 -m SRP081251
```

Or download the data manually and put it in the respective folder

Question 15:

How to find the SRR number from a paper?

- Search in the paper directly for SRR, or NCBI, or Accession with ctrl+f
- Click the link
- Look for a SRA or project number
- Click on each samples and look for SRR-Number

3.2 Quality control (fastqc & fastp)

A fastqc output folder was created and **fastqc** was performed.

Afterwards **fastp** was run.

Multiqc was used to see a summary of all files at once.

```
mkdir ../qc_reports

for i in *.fastq.gz; do fastqc -t 4 -o ../qc_reports/fastqc_output $i; done

for i in *.fastq.gz; do fastp -i $i -o ${i}_cleaned.fastq.gz -h
../qc_reports/${i}_fastp.html -j ${i}_fastp.json -w 4 -q 20 -z 4; done

cd ..
multiqc -d . -o ../qc_reports/multiqc_output
```

3.3 READemption preperation

READemption is a pipeline used for computational evaluation of RNA-Seq data. Since multiple different steps take a while to complete, the entire pipeline will be run in one script.

The READemption pipeline is divided into 5 steps:

1. **Create** a project folder and the required subfolders


```
conda activate /home/sunam226/.conda/envs/reademption

reademption create --project_path READemption_analysis --species
salmonella="Salmonella Typhimurium"
```

This will create a folder structure as shown below. It contains both the input and the output folders.

```
READemption_analysis
├── config.json
├── input
│   ├── Salmonella_annotations
│   ├── Salmonella_reference_sequences
│   └── reads
└── output
    ├── align
    │   ├── alignments
    │   ├── index
    │   ├── processed_reads
    │   ├── reports_and_stats
    │   │   ├── stats_data.json
    │   │   └── version_log.txt
    └── unaligned_reads
```

2. **Prepare** the reference sequences and annotations

- Save the general URL for *Salmonella typhimurium* SL1344 as the variable FTP_SOURCE. This makes it easier to access the URL in following codes

```
FTP_SOURCE=ftp://ftp.ncbi.nih.gov/genomes/archive/old_refseq/Bacteria/Salmonella_e
nterica_serovar_Typhimurium_SL1344_uid86645/
```

- Download the reference sequences (chromosome + 3 plasmids) as FASTA (.fa) files

```
wget -O READemption_analysis/input/salmonella_reference_sequences/NC_016810.fa
$FTP_SOURCE/NC_016810.fna
wget -O READemption_analysis/input/salmonella_reference_sequences/NC_017718.fa
$FTP_SOURCE/NC_017718.fna
wget -O READemption_analysis/input/salmonella_reference_sequences/NC_017719.fa
$FTP_SOURCE/NC_017719.fna
wget -O READemption_analysis/input/salmonella_reference_sequences/NC_017720.fa
$FTP_SOURCE/NC_017720.fna
```

- Download the genome annotation file (.gff3) and unzip it

```
wget -P READemption_analysis/input/salmonella_annotations
https://ftp.ncbi.nlm.nih.gov/genomes/all/GCF/000/210/855/GCF_000210855.2_ASM21085v
2/GCF_000210855.2_ASM21085v2_genomic.gff.gz

gunzip
READemption_analysis/input/salmonella_annotations/GCF_000210855.2_ASM21085v2_genom
ic.gff.gz
```

- Change the headers of the fasta files to the header of annotation file (annotation file + reference file need to look the same: Sequence ID first):
 - By hand: Remove everything in front of the NZ-number in sublime text editor, than safe.
 - Or use command

```
sed -i "s/>/>NC_016810.1 /"
READemption_analysis/input/salmonella_reference_sequences/NC_016810.fa
sed -i "s/>/>NC_017718.1 /"
READemption_analysis/input/salmonella_reference_sequences/NC_017718.fa
sed -i "s/>/>NC_017719.1 /"
READemption_analysis/input/salmonella_reference_sequences/NC_017719.fa
sed -i "s/>/>NC_017720.1 /"
READemption_analysis/input/salmonella_reference_sequences/NC_017720.fa
```

Check the files if the IDs are at the start and match

- Then download the RNA-Seq libraries containing the reads. To make analysis faster in the course, these librarians only contain subsamples of the original libraries

```
wget -P READemption_analysis/input/reads http://reademptiondata.imib-
zinf.net/InSPI2_R1.fa.bz2
wget -P READemption_analysis/input/reads http://reademptiondata.imib-
zinf.net/InSPI2_R2.fa.bz2
wget -P READemption_analysis/input/reads http://reademptiondata.imib-
zinf.net/LSP_R1.fa.bz2
wget -P READemption_analysis/input/reads http://reademptiondata.imib-
zinf.net/LSP_R2.fa.bz2
```

3.3 Run READemption pipeline

Because multiple steps in the pipeline take a long time, everything is run in one script (displayed beneath the explanation of the different steps)

1. **Read processing:** remove the poly a tails with --poly_a_clipping
 2. **Mapping:** align the short reads to the reference sequence
- The result can be seen in **output/align** folder.

```
$ reademption align --processes 4 --poly_a_clipping --project_path
READemption_analysis.
```

3. **Coverage:** See how often each part of the reference sequence appears in the short reads.

- It is important to normalize the coverage across the genome, to reduce differences coming from library size and composition.
- A forward and reverse strand evaluation is performed (forward=positive values, reverse=negative values)
- Output files are stored as wiggle format (.wig) for genome browsers
 - `READemption_analysis/output/species1_coverage-raw/` contains raw counting values without normalization
 - `READemption_analysis/output/species1-tnoar_min_normalized/` contains coverage values normalized by total number of aligned reads (tnoar) multiplied by lowest number of aligned reads of all libraries
 - `READemption_analysis/output/species1_coverage-tnoar_mil_normalized/` contains coverage values normalized by the total number of aligned reads multiplied by one million

```
reademption coverage -p 4 --project_path READemption_analysis --paired_end
```

4. **Gene wise quantification:** Compare the sequence reads to a genomic or transcriptomic reference sequence

- Quantify the number of reads aligning with the annotations found in .gff3 file
- Output:
 - `READemption_analysis/output/species1_gene_quanti_per_lib/` contains a coverage file for each sample showing the number of reads covering each entry of the annotations file
 - `READemption_analysis/output/species1_gene_quanti_combined/` contains normalized coverage values of each sample combined into one file
- 5. **Differential gene expression analysis** is performed by the R library DESeq2. It compares all conditions with each other and create a .scv file for each comparison.
- Output in `READemption_analysis/output/salmonella_deseq/deseq_with_annotations/`
 - BaseMean: mean of normalized counts of all samples. *Note that it does not take into account gene length, it is used in DESeq2 to calculate the dispersion of a gene.*
 - Log2FoldChange: Describes the difference when comparing two things. Fold change is calculated as the ratio of A/B or B/A with log2.

```
reademption deseq --libs
InSPI2_R1.fa.bz2,InSPI2_R2.fa.bz2,LSP_R1.fa.bz2,LSP_R2.fa.bz2 -c
InSPI2,InSPI2,LSP,LSP -r 1,2,1,2 --libs_by_species
salmonella=InSPI2_R1,InSPI2_R2,LSP_R1,LSP_R2 project_path READemption_analysis
```

6. **Plot creation** to visualize the results.

```
reademption viz_align --project_path READemption_analysis
```

Visualize the distribution of read length as histograms.

- `input_read_length_distribution.pdf` before clipping
- `processed_reads_length_distributions.pdf` after clipping

```
reademption viz_gene_quanti --project_path READemption_analysis
```

`viz_gene_quanti` generates two documents:

- `expression_scatter_plots.pdf` a scatter plot of the raw gene wise quantification values for each library pair
- `rna_class_sizes.pdf` shows the proportion of the features selected in `gene_quanti` command

```
reademption viz_deseq --project_path READemption_analysis
```

`viz_deseq` generates 3 plots:

- `MA_plot`: shows the log2fold changes (M) against the mean (A) of normalized counts for all the samples in the dataset (baseMean). Values coloured red have an adjusted p value $\text{padj} \leq 0.1$.
- 2 `volcano_plots`: show the log 2 fold change plotted against the -log transformed p value and against the -log transformed, adjusted p value.
-

Final Script

```
#!/bin/bash
#SBATCH --job-name=pub_data
#SBATCH --output=pub_data.out
#SBATCH --error=pub_data.err
#SBATCH --nodes=1
#SBATCH --ntasks-per-node=1
#SBATCH --cpus-per-task=8
#SBATCH --mem=32G
#SBATCH --qos=long
#SBATCH --time=0-05:00:00
#SBATCH --partition=all
#SBATCH --reservation=biol217

source ~/.bashrc

module load miniconda3/4.7.12.1
```

```

module load python/3.7.4
conda activate /home/sunam226/.conda/envs/reademption
##### ---CALCULATIONS---
#aligning:
reademption align --fastq -f READemption_analysis --poly_a_clipping --
min_read_length 12 --segemehl_accuracy 95

# coverage:
reademption coverage -p 4 --project_path READemption_analysis

#gene wise quanty:
reademption gene_quanti -p 4 --features CDS,tRNA,rRNA --project_path
READemption_analysis

#differential gene expression:
reademption deseq -l sRNA_R1.fa,sRNA_R2.fa,wt_R1.fa,wt_R2.fa -c sRNA,sRNA,wt,wt \
-r 1,2,1,2 --libs_by_species Methanosarcina=sRNA_R1,sRNA_R2,wt_R1,wt_R2 --
project_path READemption_analysis

##### ---PLOTS---
reademption viz_align --project_path READemption_analysis
reademption viz_gene_quanti --project_path READemption_analysis
reademption viz_deseq --project_path READemption_analysis
conda deactivate
jobinfo

```

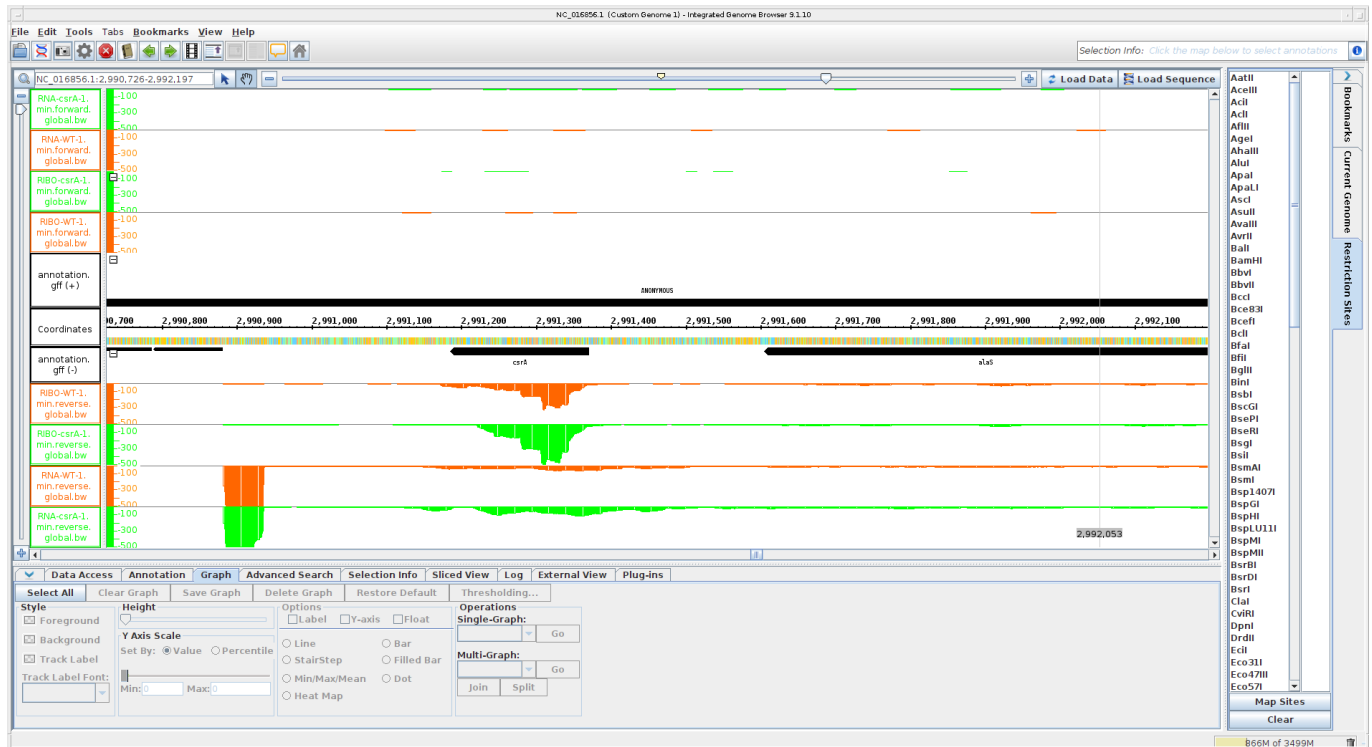
3.4 Genome Browser

Opened Ribo-seq and RNA-seq data in the **I**ntegrated **G**enome **B**rowser (IGB) to see the transcriptomics and translomics results.

The reference genome, as well as the gene annotations were loaded into the browser, to compare the generated mRNA to it.

csrA gene Infos

We looked at csrA in particular, because it was used in the paper.

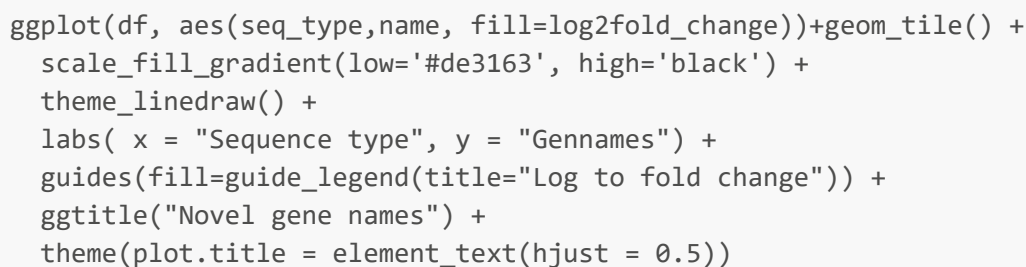


csrA lays on the reverse strand (negative values). IT seems to be transcribed and translated in both wildtype and mutant, based on the coverage of mRNA and ribosome-bound RNA.

- Start codon
 - ATG (reverse complementary TAC)
- Stop codon
 - TAA (reverse complementary ATT)
- Aminoacid length
 - $2991371 - 2991185 = 186$ bp
 - $186 \text{ bp} / 3 = 62$ AS
 - stop-codon is not a AA, so -1
--> 61 Aminoacids
 - small protein because <70 AS
- its SD (consensus at -7 or -4)
 - Shine dalgano
- The name of the upstream gene
 - upstream alaS
- Is csrA translated?
 - Yes, Ribosome coverage along the entire genes
 - UTRs clearly visible next to it

Find a significantly differential expressed gene (wt/scrA Mutant)

- Look at p-value <0.05 in table
- ribB**



4. R

Near the end of the course, we had a small excursion into R, to learn how to handle data and properly create plots.

Question 16:

Transfer dataframe iris from long to wide format

```
data("iris")

iris_long<-tidyr::gather(iris, key='Species', value='log2FC')
```

Question 17:

Upload 5 different graphs of trees

```
data("trees")

box<-ggplot(data=trees, mapping=aes(Height, Volume))+geom_boxplot()+ggtitle("Trees
Boxplot")+theme(plot.title = element_text(hjust = 0.5),
axis.text.y=element_blank())

jit<-ggplot(data=trees, mapping=aes(Volume, Girth), colour=Height) +
geom_jitter(color='darkred')+ggtitle("Trees Jitterplot")+theme(plot.title =
element_text(hjust = 0.5,color='green'))

pnt<-ggplot(data=trees, mapping=aes(Height, Volume, size=Girth)) +
geom_point()+ggtitle("Trees Pointplot")+theme(plot.title = element_text(hjust =
0.5), axis.text.y=element_blank() )+theme_bw()

dot<-ggplot(data=trees, mapping=aes(Girth, as.factor(Volume))) +
geom_dotplot()+ggtitle("Trees Dotplot")+theme(plot.title = element_text(hjust =
0.5), axis.text.y=element_blank())+theme_dark()

lin<-ggplot(data=trees, aes(x = Girth, y = Volume, color = Height)) + geom_line()
+ guides(color = guide_legend(title = "Height")) +ggtitle("Trees
Lineplot")+theme(plot.title = element_text(hjust = 0.5),
axis.text.y=element_blank())+theme_bw()

ggarrange(box, jit, pnt, dot, lin)
```