Discordant bioinformatic predictions of antimicrobial resistance from whole-genome sequencing data of bacterial isolates: an interlaboratory study

Supplementary Methods

Each participating team was asked to provide a description of the pipeline used within this study. The submitted methods sections are included below:

Lab_1

Workflow Overview

Raw reads were checked for quality using FastQC, trimmed/assembled/corrected/reassembled using shovill pipeline (see https://github.com/tseemann/shovill). Species ID was achieved with the raw reads using Kraken-HLL (see https://github.com/fbreitwieser/krakenhll) and Bracken (for abundances, see https://github.com/jenniferlu717/Bracken). AMR analysis involved searching the assembled contigs CARD database with RGI tool and ResFinder/ARG-ANNOT databases with the software tool c-SSTAR (see https://github.com/chrisgulvik/c-SSTAR).

Commands

Step 1 (~/AMRIL_data/AMRIL_analysis_raw_data->cat 00.cmds_ID)

"Unzipping fastq.gz"
gunzip *.gz
"Running FastQC"
mkdir 00.fastqc
fastqc -t 10 -o 00.fastqc -f fastq *.fastq
"Running shovill"
shovilloutdir 01.shovillR1 AMRIL_7_R1_001.fastqR2 AMRIL_7_R2_001.fastqdepth 0trim - -force &

(NOTE: for sample AMRIL_2 required to add tag -min_cov 1 - since the majority of the contigs had a coverage of 2X, and were discarded after the pilon correction step when using the default settings)

"Running KrakenHLL and Braken (with Krona for html reports)"

mkdir 02.krakenhll

krakenhll --threads 10 --report-file 02.krakenhll/krakenhll_report --db "minikraken_20171101_8GB_dustmasked" --output 02.krakenhll/krakenhll_results --preload -fastq-input --paired AMRIL_7_R1_001.fastq AMRIL_7_R2_001.fastq

python2.7 ~/ktoolu/kt_summarize.py --include-unclassified --draw-krona-plot 02.krakenhll/krakenhll_results_krona.plot.html --path-to-krona ~/KronaTools-2.7/bin/ 02.krakenhll/krakenhll_results

kraken-report --db "minikraken_20171101_8GB_dustmasked" 02.krakenhll/krakenhll_results > 02.krakenhll/krakenhll_results_kraken-report

python ~/Bracken/est_abundance.py -i 02.krakenhll/krakenhll_results_kraken-report -k ~/Bracken/minikraken_8GB_200mers_distrib.txt -o 02.krakenhll/krakenhll_results_Bracken_mini8GB_200mers_est_abundance.txt

(NOTE: for samples 2 and 5 there was difficulty in assigning the taxonomy to the species level, this may be due to a combination of low quality/depth in the raw sequence data and the fact that the genus is Enterobacter).

Step 2 (~/AMRIL_data/AMRIL_analysis_raw_data->cat 00.cmds_AMR)

#RGI-CARD (contigs)

sudo docker run -v ~/AMRIL_data/AMRIL_analysis_raw_data/AMRIL_8/:/myData -w /myData finlaymaguire/rgi rgi main --input_sequence 01.shovill/contigs.fa --output_file 14.rgi_card.out -input_type contig

#c-SSTAR (contigs) - resfinder and argannot

~/ess_apps/c-SSTAR/c-SSTAR -d ~/ess_apps/c-SSTAR/db/ResGANNOT_srst2.fasta -g 01.shovill/contigs.fa -o 16.c-SSTAR_ResGANNOT > 16.c-SSTAR_ResGANNOT/stdout

Software versions

FastQC 0.11.5shovill 0.9.2using SPADes (version 3.10.1)KrakenHLL0.3.2using database minikraken_20171101_8GB_dustmasked (Nov 1 2017)

KronaTools2.7Bracken1.0.0using database minikraken_8GB_200mers_distrib.txt (Nov 2 2017)CARD-RGI4.0.2using card database (Jul 24 2018)c-SSTAR1.2cusing resfinder and arg-annot databases from (https://github.com/tomdeman-
bio/Sequence-Search-Tool-for-Antimicrobial-Resistance-SSTAR-
/blob/master/Latest AR database/ResGANNOT srst2.fasta)

Lab_2

Raw reads trimmed using Trimmomatic

TrimmomaticPE -phred33 \$path4\$file'_R1_001.fastq.gz' \$path4\$file'_R2_001.fastq.gz' \$path4\$file'.pe_1.fq' \$path4\$file'.upe_1.fq' \$path4\$file'.pe_2.fq' \$path4\$file'.upe_2.fq' ILLUMINACLIP:TruSeq3-PE.fa:2:30:10 LEADING:3 TRAILING:3 SLIDINGWINDOW:4:24 MINLEN:36 2> \$file'.log'

and Trim Galore.

```
trim_galore --length 36 --paired --retain_unpaired $path4$file'.pe_1.fq' $path4$file'.pe_2.fq'
```

trim_galore --length 36 \$path4\$file'.upe_1.fq' \$path4\$file'.upe_2.fq'

mv \$path4\$file'.pe_1_val_1.fq' \$path4\$file'.pe_1.fq'

cat \$path4\$file'.pe_1_unpaired_1.fq' \$path4\$file'.upe_1_trimmed.fq' > \$path4\$file'.upe_1.fq'

mv \$path4\$file'.pe_2_val_2.fq' \$path4\$file'.pe_2.fq'

cat \$path4\$file'.pe_2_unpaired_2.fq' \$path4\$file'.upe_2_trimmed.fq' > \$path4\$file'.upe_2.fq'

Reads assembled using SPAdes.

spades --careful --cov-cutoff auto -t 8 --pe1-1 \$fq\$file'.pe_1.fq' --pe1-2 \$fq\$file'.pe_2.fq' --s1 \$fq\$file'.upe_1.fq' --s2 \$fq\$file'.upe_2.fq' -k 21,33,55,77,99 -o \$wd'spades'

Resistance genes detected using abricate with the CARD database on assemblies. BLAST (<u>https://blast.ncbi.nlm.nih.gov/Blast.cgi</u>) on 16S and other genes.

Lab_3

Read QC, Assembly

10 paired end fastq files were downloaded from the GOSH server. Reads were initially screened for quality using FastQC (<u>https://www.bioinformatics.babraham.ac.uk/projects/fastqc/</u>)

Raw reads, were assembled using UniCycler (Wick *et al.*, 2017) in "Illumina-only" assembly mode. UniCycler uses SPAdes' built-in read correction module. Here we used SPAdes version 3.10.1 (Bankevich *et al.*, 2012) and generate a SPADes assembly graph. UniCycler then performs additional assembly improvement steps through identifying multiplicity of contigs, scaffolding, overlap removal and bridging (see Wick *et al.* for details). The resultant assemblies were visualized in Bandage (Wick *et al.*, 2015) to check overall quality and merge overlapping graphs.

Species Assignment

Taxonomic assignment was performed using two approaches. The first utilized the raw sequencing reads, without any assembly step. Raw reads were screened using the distance based tool MASH(Ondov *et al.*, 2016), which implements the minHASH k-mer matching algorithm. MASH distances were calculated for each of the 10 genomes against an archive of RefSeq genomes (Pruitt *et al.*, 2012), release 70, sketched using k=21 and s=1000. The closet matching reference was selected based on the proportion of matching k-mers.

In addition, and to validate the assignments using MASH, the *de novo* assemblies were uploaded to WGSA (wgsa.net), which provides a rapid taxonomic assignment.

AMR Gene Detection

Denovo assemblies were uploaded to 2 independent reference databases for AMR gene identification: CARD (Jia *et al.*, 2017) and ResFinder (Zankari *et al.*, 2012), the latter was run with a %ID threshold of 90% and a selected minimum length of 80%. Resulting output files were downloaded and inspected.

Assigning Resistance Phenotype

Resistance phenotypes were considered in the context of the presence, absence and co-occurence of different AMR genes. We took a conservative approach corresponding to the 'strict' classification in CARD, calling a resistance gene present if there was >80% identity to the reference. We operated on a precautionary principle (appropriate for clinical work, where WGS could be used to triage samples) and called a gene as present even in cases where not all of the reference gene was covered (e.g. AMRIL_2, with 22% coverage of SHV-156).

The CARD reference database ontology terms were used to relate resistance genes to resistance to specific antibiotics. In addition to using this ontology, we discussed the following heuristic reasoning for our phenotype classifications based on our existing knowledge:

Ciprofloxacin (fluoroquinolone): a combination of specific genes (e.g. *patA* and *patB*, *Qnr1*) and mutations (e.g. in *gyrA* and *gyrB*).

Gentamicin and **amikacin** (aminoglycoside): aminoglycoside resistance is conferred by specific genes (e.g. *AAC* family). AAC(3)-II confers resistance to gentamicin, but AAC(6')-I is additionally required for resistance to amikacin.

Cefotaxime (beta-lactam): many different beta-lactamases, with resistance correspondingly growing additively with the number of genes present.

Lab_4

Protocol summary

Ariba (https://github.com/sanger-pathogens/ariba) was used to identify antibiotic resistance genes and MLST sequence types by running local assemblies.

Since ariba does not allow species identification, kraken (https://ccb.jhu.edu/software/kraken/; https://genomebiology.biomedcentral.com/articles/10.1186/gb-2014-15-3-r46) was used to identify the species before ariba was run to identify the sequence type. The kraken analysis is based on a database from the following publication: Browne at al. Nature 2016, https://www.nature.com/articles/nature17645

Example kraken command:

bsub.py --threads 8 --queue normal 40 kraken1 'metagm_run_kraken -t 8 -noclean ~/lustre/Agata_YALE/kraken/InternalKraken_OCT2017 AMRIL_1.report ../fastq/AMRIL_1_R1_001.fastq.gz ../fastq/AMRIL_1_R2_001.fastq.gz'

To identify antibiotic resistance genes, the two following databases were used: The Comprehensive Antibiotic Resistance Database "CARD" version 2.0.1 (https://card.mcmaster.ca/home; https://www.ncbi.nlm.nih.gov/pubmed/23650175) and Antibiotic Resistance Gene-ANNOTation "ARG-ANNOT" (http://en.mediterranee-infection.com/article.php?laref=283&titre=; https://www.ncbi.nlm.nih.gov/pubmed/24145532). Databases were downloaded using ariba's getref command on June 27th 2018. Ariba was run using standard settings and results were checked manually. Disparate results between the two databases are specified in the comment column, only the hit with the best identity was kept. Possible contaminations were highlighted in the coverage of other contigs (column ctg_cov in the original ariba reports). Only hits of genes conferring resistance to aminoglycosides, beta-lactams and fluoroquinolones were listed in the Excel file "AMRIL_WGS_reporting". All hits were kept in the original ariba report files ("AMRIL_*_CARD_report.tsv"; "AMRIL_*_argannot_report.tsv") Example command:

bsub.py --queue small 2 log 'ariba run ~/lustre/ariba_databases/mlst/Klebsiella_pneumoniae/ref_db ../fastq/AMRIL_1_R1_001.fastq.gz ../fastq/AMRIL_1_R2_001.fastq.gz AMRIL_1'

Version information:

ARIBA version: 2.12.1

External dependencies:

bowtie2 2.2.3 /software/pathogen/external/apps/usr/bin/bowtie2 cdhit 4.6 /software/pathogen/external/apps/usr/bin/cd-hit-est nucmer 3.1 /software/pathogen/external/apps/usr/bin/nucmer

Python version: 3.6.0 (default, Jan 18 2017, 11:39:44) [GCC 4.6.3]

Python packages:

ariba 2.12.1 /software/pathogen/external/apps/usr/local/Python-3.6.0/lib/python3.6/sitepackages/ariba/__init__.py bs4 4.6.0 /software/pathogen/external/apps/usr/local/Python-3.6.0/lib/python3.6/sitepackages/bs4/__init__.py 4.2.0 /software/pathogen/external/apps/usr/local/Python-3.6.0/lib/python3.6/sitedendropy packages/dendropy/ init .py pyfastag 3.16.0 /software/pathogen/external/apps/usr/local/Python-3.6.0/lib/python3.6/sitepackages/pyfastaq/__init__.py 0.10.3 /software/pathogen/external/apps/usr/local/Pythonpymummer 3.6.0/lib/python3.6/site-packages/pymummer/__init__.py /software/pathogen/external/apps/usr/local/Python-3.6.0/lib/python3.6/sitepysam 0.11.2.2 packages/pysam/__init__.py

Kraken version: kraken-0.10.6-a2d113dc8f Database: /nfs/users/nfs_s/sb53/lustre/Agata_YALE/kraken/InternalKraken_OCT2017

Lab_5

FastQ files were downloaded and analyzed with the FastQC program (v0.11.7) by Babraham bioinformatics.

Reads are assembled using the A5_miSeq pipeline (v20160825), using standard parameters.

```
for file in *_R1.fastq
do
file2=${file/R1/R2}
std=${file/_R1.fastq/}
a5_pipeline.pl $file $file2 $std
done
```

ID was obtained uploading the assembled contigs to the online KmerFinder tool (v2.5) from CGE. Conda is used for the posterior analysis, using the RGI (v 4.0.3) software. Also, last version of the CARD database (v 2.0.2) is used. Another bash script is used to run RGI.

```
conda create --name amr
conda activate amr
conda install rgi
rgi load -i /path/to/card.json
#Another bash script is used to run RGI.
#! /bin/env bash
#File: RGI.sh
for file in *.contigs.fasta
do
    std=${file/.contigs.fasta/}
    rgi main -i $file -n 4 -o $std
done
```

Results with a Cut-off marked as "Loose" were not taken into account.

Lab_6

We used BioNumerics version 7.6.3 with E.coli plug in tools from CGE (ResFinder).

Web tool KmerFinder (v2.5) was used to determine bacterial species <u>http://www.genomicepidemiology.org/</u>. We did not search for *gyrA* and *parC* mutations (quinolone resistance) in these sequences.

Lab_7

Reads were trimmed using trimmomatic. Species classification was performed using centrifuge. Resistance genes were found using SRST2. Resistance to specific drugs were predicted by assessing the literature

Read trimming

Program: trimmomatic Version: v0.38 Program call:

trimmomatic PE -threads 1 -phred33 AMRIL_1_R1_001.fastq.gz AMRIL_1_R2_001.fastq.gz -baseout AMRIL_1 LEADING:3 TRAILING:3 SLIDINGWINDOW:4:20 MINLEN:36

Species classification

Program: Centrifiuge Version: 1.0.3-beta Database: ftp://ftp.ccb.jhu.edu/pub/infphilo/centrifuge/data/p+h+v.tar.gz Program call:

centrifuge -x p+h+v -1 AMRIL_1_1P -2 AMRIL_1_2P > AMRIL_1.log centrifuge-kreport x p+h+v AMRIL_1.log > AMRIL_1.kreport

Drug resistance gene detection

Program: SRST2 Version: 0.2.0 Database: <u>https://github.com/katholt/srst2/blob/master/data/ARGannot_r2.fasta</u> Program call:

srst2 --input_pe AMRIL_1_R*.fastq.gz --forward _R1_001 --reverse _R2_001 --output test -gene_db ARGannot_r2.fasta

Lab_8

 Table 1| List and description of tools utilized in this study.

Name	Associated Workflow	Description
Conda	N/A	Package, dependency and environment
		management for multiple programming
		languages.
		(https://conda.io/docs/)
Snakemake ¹	N/A	Python based workflow management system.
		(https://snakemake.readthedocs.io/en/stable/)

Sickle	trim	C based program for adaptive trimming of FASTQ files [according to quality]. (<u>https://github.com/ucdavis-</u> <u>bioinformatics/sickle</u>)
Unicycler ²	assembly	Assembly pipeline for bacterial genomes from NGS reads; acts as an optimizer for SPAdes. (https://github.com/rrwick/Unicycler)
SPAdes ³	assembly	Python based genome assembler with built-in read error correction. (<u>http://cab.spbu.ru/software/spades/</u>)
ABRicate	amr	Perl based program for screening of assembled genomes [contigs] for antimicrobial resistance or virulence genes [database specific]. (https://github.com/tseemann/abricate)
OKraken ⁴	taxonomy; taxonomy- report	C++, Perl and Shell based program for taxonomic classification from sequences files [individual reads, assembled genomes etc]. (https://github.com/DerrickWood/kraken)

Workflow Overview

- 1. FASTQ reads are trimmed using **sickle** yielding trimmed paired reads 1 and 2, and another trimmed reads file containing the single, unpaired reads.
- 2. Trimmed FASTQ (1, 2 and singles) are then assembled using the Unicycler optimizer for SPAdes. The Unicycler optimizer produces a single 'assembly.fasta' file (in place of the 'contigs.fasta' and 'scaffolds.fasta' default output from SPAdes).

3. Antimicrobial resistance (AMR) is then predicted using the ABRicate tool, along with the ResFinder database (note that resfinder is selected over other AMR databases as it only focuses on acquired AMR genes, not chromosomal point mutations). Report is generated as '.csv' output.

```
abricate --db=resfinder assembly.fasta --csv > amr.csv
```

The output of 'abricate' is a list of predicted acquired AMR genes. These genes correspond to specific drug-class resistance (i.e. aminoglycosides) as indicated in the ResFinder database. These gene-class associations are used to predict the AMR phenotype of samples to specific drugs (by identifying the class associated with each drug, i.e. ciprofloxacin \rightarrow fluoroquinoline).

4. Taxonomic predictions are performed by Kraken (as well as report generation). Kraken utilizes a local database, downloaded from the Kraken repository (the standard database). Taxonomic prediction file is generated by:

kraken --preload --db <DATABASE_FOLDER> --fasta-input assembly.fasta --threads
<n_cpus> > kraken.out

And the subsequent [human-readable] report generated by:

kraken-report --db <DATABASE_FOLDER> kraken.out > kraken.txt

The species selection is done by identifying the species with the largest percentage abundance in the sample.

5. The 'all' rule is a Snakemake specific workflow rule that indicates all the above rules (as indicated in Fig. 1) are required to be carried out.

Lab_9

In-house program "qa_and_trim" calls Trimmomatic; nucleotides with a Phred score less than Q30 at the ends of the reads were removed with Trimmomatic. In house "kmerid" was used for kmer-based species identification (<u>https://github.com/phe-bioinformatics/kmerid</u>). Resistance gene detection was done using 'Genefinder', an in-house algorithm that uses bowtie2 to map sequence reads to reference sequences of interest and Samtools vs 0.1.18 to generate an mpileup file, which is then parsed for the rapid detection of sought sequences. Genes were called as present within a genome when detected with 100% coverage and >90% nucleotide identity to the reference gene.

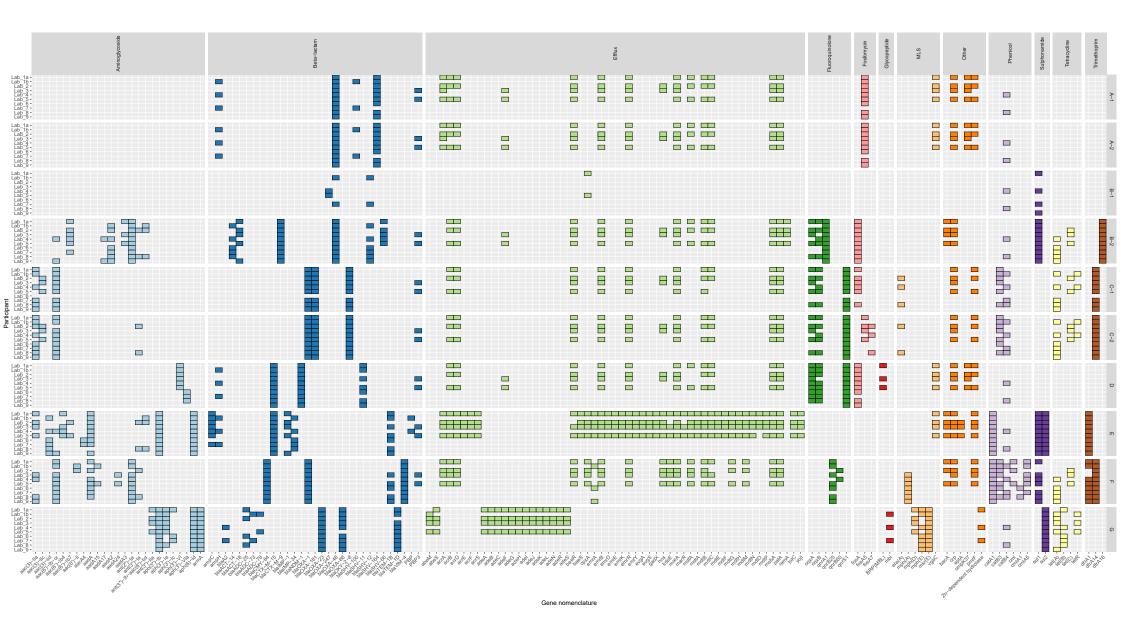


Figure S1. The presence of all AMR-associated genes in each sample by each participant. Genes are organised and coloured by the class of antibiotics they are associated with resistance, or if they are associated with the efflux of multiple classes of antibiotics.