





Automated identification of antibiotic resistance mutations in bacterial genomes

Creation of the ARM-find pipeline

Master's thesis in Bioinformatics

MARTIN BOSTRÖM

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Department of Mathematical Sciences Division of Applied Mathematics and Statistics Systems Biology and Bioinformatics CHALMERS UNIVERSITY OF TECHNOLOGY Gothenburg, Sweden 2017 Automated identification of antibiotic resistance mutations in bacterial genomes Creation of the ARM-find pipeline Martin Boström

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Cover: Logo for the ARM-find pipeline.

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Abstract

Antibiotic resistant bacteria are a fast-growing problem, worsened by the overuse of antibiotics. In treatment of infections, it is often necessary to determine a resistance profile for the infecting bacteria in order to establish the correct treatment, and the time required to accomplish that through cultivation is sometimes long. Recent advances in next-generation sequencing techniques have decreased the cost and time requirements of whole-genome sequencing to the point where a bioinformatical approach to resistance profile determination may prove faster than the traditional one. Software tools are already available for the detection of mobile resistance genes in bacterial genomes, but to my knowledge, no open-source tools exist that detect resistance mutations. This thesis describes the creation of the ARM-find pipeline, which can find such resistance mutations in assembled bacterial genomes, including draft genomes. It comes with a resistance mutation database that currently contains fluoroquinolone resistance mutations in *E. coli*, but is easily extensible to cover additional antibiotics and species. In addition to describing the pipeline, this thesis covers the prevalence of fluoroquinolone resistance mutations in $E. \ coli$ and Shigella. The pipeline was used to catalogue substitutions (in comparison to E. coli K-12 MG1655) in the genes encoding DNA gyrase and topoisomerase IV – the targets of fluoroquinolones – in all RefSeq genomes for both E. coli and Shigella. Fluoroquinolone resistance mutations were found to be common, and the relative frequencies of the mutations matched what has been reported in previous studies on the subject.

Keywords: Antibiotic resistance, pipeline, fluoroquinolones, bioinformatics, *E. coli*, *Shigella*, mutations

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Finally, I would like to thank the people behind www.armfinder.com for having already taken the name I had initially wanted for my pipeline (notice the lack of -er in ARM-find). I wish you all the best in your mission to "Register, Find and Contact Armwrestlers in the World".

Martin Boström, Gothenburg, February 2017

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Introduction

Antibiotics are molecules that either kill or inhibit the growth of bacteria. Most antibiotics that are in use as drugs today were originally found in nature, where they are used by bacteria or fungi against other microbes [1]. They are an important defence against bacterial infections, and are essential in healthcare. However, when antibiotics are given as treatment, antibiotic resistant bacteria are given an increased chance of achieving dominance through the elimination of the non-resistant competition. Through this mechanism, overuse of antibiotics has resulted in increasing resistance among bacteria, and the trend is worsening [2]. Without effective antibiotics, we could face a world where infections that have previously been easily treatable become a death sentence.

Antibiotics work by affecting certain targets – the main ones are bacterial cellwall biosynthesis, protein synthesis, and DNA replication and repair. There are different strategies that are used by bacteria to survive such attacks. They may alter cell permeability, e.g. by changing the amounts of efflux pumps and porins or by altering the cell wall, so that the concentration of the drug at the site of its target remains too low to cause severe harm. An alternative strategy is to modify the antibiotic so that it no longer binds to its target, or at least does so with lower affinity. The bacteria may also modify the drug's target, either through mutations or later modifications (such as the addition of molecules at binding sites) to reduce its binding affinity to the drug [1]. Finally, metabolic pathways can be altered to no longer rely on the targeted enzyme [3].

Acquiring or improving the properties described above can be done through acquisition of novel DNA or through alterations in pre-existing DNA. In the first case, resistance genes can be transferred horizontally between bacteria by plasmids, bacteriophages, naked DNA, or transposons. These genes could for instance code for enzymes that modify the antibiotic, or replace antibiotic-targeted enzymes in metabolic pathways. In the second case, antibiotic resistance arises through stepwise mutations, with each mutation resulting in less susceptibility to the antibiotic in question. The typical example of this is a reduced binding affinity between a drug and its target, caused by mutations to that target.

Whether resistance genes or resistance mutations are the most important varies between different antibiotics. In the case of fluoroquinolones, a broad-spectrum class of antibiotics that are highly effective for treating a variety of infections, chromosomal mutations have the highest impact on resistance. Since all fluoroquinolones have the same antibiotic mechanism, any mutation in the target genes that results in resistance to one fluoroquinolone will also yield resistance to all the others [4]. Fluoroquinolone resistance has been extensively studied in *Escherichia coli* [4], a common source of urinary tract infections. Studies have found that an alarmingly large portion of *E. coli* have achieved some degree of resistance to fluoroquinolones [2].

Because of the spread of antibiotic resistance, it is important to be able to quickly characterise resistance in an infection, so that the correct treatment may be provided. Currently, the antibiotic to give may be determined based on the symptoms of the patient, or ideally by isolating the bacteria responsible for the infection and evaluating their resistance profile. The latter is done by growing them in the presence of different antibiotics in the lab [5], which can be a slow process, especially when the infecting strain grows slowly, is difficult to cultivate, or is multi-resistant. However, we may now have another option at hand. Recent developments in next-generation sequencing (NGS) techniques have seen both the time requirement and the cost of sequencing entire genomes decrease exponentially [6]. If the genetic alterations that lead to antibiotic resistance are known, we could sequence the genomes of bacteria and use that information to infer what antibiotics are suitable for treatment. This has already been shown to be faster than cultivation-based resistance determination in the UK for the slow-growing M. Tuberculosis [7].

If we are to improve the speed of resistance profile determination for bacteria through whole-genome sequencing, we must have bioinformatical software tools that can find resistance-related genetic elements quickly and efficiently. We will need tools both for finding mobile resistance genes, and for finding resistance mutations, like the ones that are important for fluoroquinolone resistance. There are several software tools in existence for identifying mobile resistance genes, such as ResFinder [8], the Comprehensive Antibiotic Resistance Database (CARD) [9], and the Antibiotic Resistance Genes Database (ARDB) [10]. However, to my knowledge there are no open-source tools that are designed to identify resistance mutations in bacteria, which leads us to the aims of this project.

1.1 Aims

In this thesis, I have developed a pipeline for identifying antibiotic resistance mutations in bacterial genomes; it is called ARM-find, short for Antibiotic Resistance Mutation finding pipeline. To its database of resistance mutations, I have added mutations that confer resistance to fluoroquinolones in *E. coli*. To test the performance of the pipeline, I have searched for fluoroquinolone resistance mutations in all *E. coli* genomes available from NCBI's database RefSeq, as well as in the closely related genus *Shigella*.

– Theory

The aim of the following sections is to provide a brief explanation of how fluoroquinolones work, and what their targets are, as well as to explain the different types of sequence alignments used in the pipeline, and some relevant information regarding genome assembly.

2.1 Fluoroquinolones and Their Targets

Fluoroquinolones are a class of synthetic antibiotics that inhibit the replication and transcription of bacterial DNA, by acting against DNA gyrase and topoisomerase IV. At high concentrations, this leads to cell death. They are broad-spectrum, and important in health-care. Fluoroquinolones are categorised into generations based on the improvements that have been made to, among other things, their half life and range of different kinds of bacteria that they are active against. What they have in common is that they are synthetic fluorinated analogues of nalidixic acid, and tend to contain a 4-pyridone-3-carboxylic acid with a ring connecting to positions 5 and 6, as shown in figure 2.1 [11].



Figure 2.1: The required pharmacophore of fluoroquinolones. Image modelled after figure from [11].

The targets of fluoroquinolones, DNA gyrase and topoisomerase IV, are both topoisomerases, meaning they participate in the supercoiling of DNA. DNA gyrase is a tetrameric enzyme that is composed of two GyrA and two GyrB subunits. It is responsible for introducing negative supercoils into DNA, which is necessary for DNA replication [12]. Topoisomerase IV is homologous to DNA gyrase, and is composed of two ParC and two ParE subunits. It is involved in the separation of chromosomes during DNA replication [4]. Both enzymes cut DNA molecules in order to be able to change their coiling, and ligate them back together after having done so. The antibiotic mechanism of fluoroquinolones is that they bind to the DNA gyrase/topoisomerase IV-DNA complex, thereby stabilising it and preventing ligation [4]. This stops DNA synthesis, causing the cells to stop growing. There are several hypotheses for what causes the bactericidal effect that fluoroquinolones have. A common and likely one states that the release of DNA ends from the enzyme-DNA-drug complex triggers apoptosis [12].

The most clinically relevant cause of fluoroquinolone resistance is chromosomal mutations to the genes encoding DNA gyrase and topoisomerase IV. These mutations lower susceptibility to fluoroquinolones by reducing the binding affinity of the drug to the enzyme-DNA complex. The current knowledge of such mutations has been reviewed by Hopkins et al. [4], and that review is the basis for the resistance mutations in ARM-find's database. Though all those mutations have been associated with increased fluoroquinolone resistance, having one does not necessarily decrease a bacterium's susceptibility to the drug. As an example, all resistance mutations that have been found in topoisomerase IV have been accompanied by gyrA mutations. The reason for this is believed to be that the susceptibility of GyrA to fluoroquinolones must be decreased for mutations to the topoisomerase IV genes to even make a difference to survivability. Fluoroquinolone resistance can also be achieved through means other than mutations. One alternative is decreased uptake of the drug through increased amounts of efflux pumps or decreased amounts of porins, and another is mobile genes such as the plasmid-mediated qnr [4].

2.2 DNA and Protein Sequence Alignments

Both local and global alignments of nucleotide and amino acid sequences are used in this pipeline. Global alignments are only viable when both sequences are of roughly equal length, which of course is not the case when aligning a gene against a genome. A local alignment is suitable for aligning a shorter sequence against a longer one, as it does not attempt to match the entire sequences against each other in the way that a global alignment does. Instead, it finds where in the longer sequence the shorter sequence, or pieces of it, will fit. The choice between global and local alignments is also affected by sequence similarity, but for the purposes of the pipeline, sequence length differences are the deciding factor, as the sequences to be aligned are assumed to be highly similar. ARM-find uses BLAST (Basic Local Alignment Search Tool) [13] for local alignments, and MAFFT (Multiple Alignment using Fast Fourier Transform) [14] for global alignments.

BLAST is a heuristic local alignment tool. It splits the query sequence into short "words" (sequence snippets), and creates permutations of these words that are sufficiently similar to the original word, according to a scoring matrix. These words are then matched against the target sequence, and the matching positions are used as seeds for alignment. Every such seeded alignment is extended in both directions until its alignment score decreases beyond a certain threshold compared to the maximum value during extension. After termination, the alignment is rolled back to its maximum score. The resulting alignments' scores are then evaluated for statistical significance by comparison to random sequences. Alignments scoring above a certain threshold are kept and presented to the user [15].

MAFFT is a multiple sequence alignment tool that is capable of both local and global alignments. As the name implies, it is based on the fast Fourier transform, which is used to rapidly detect homologous segments [14]. It comes with options for several alignment strategies, including progressive methods, structural alignment methods for RNAs, and iterative refinement methods [16]. One such alignment option, the iterative refinement method G-INS-i, is used in this pipeline for global alignment. G-INS-i requires that the entire region can be aligned and attempts global alignment using the Needleman-Wunsch algorithm [17].

2.3 Genome Sequencing and Assembly

Since DNA sequencing can only generate reads of limited length, sequencing genomes necessitates the generation of many, many reads. To generate a full genomic DNA sequence, the genome has to be assembled using these reads. Assembled genomes can be in different states of completeness; genomes that have been fully assembled are called complete assemblies, whereas partially assembled genomes are called draft assemblies. In a draft assembly, the genetic sequence is typically split over many contigs, and the concept of scaffolding is important. A contig is a contiguous sequence of nucleotides known with a high degree of certainty. A scaffold can contain several contigs, between which the distance is known (or estimated), but the DNA sequence is not. The distance between contigs can be found through methods such as mate pair sequencing. The result is sequences that can include large stretches of N's, designating unknown bases. The number of N's between two contigs may not match the actual number of bases separating them exactly.

2. Theory

Methods

The pipeline was coded in Python (version 3), and is built to run on a Linux system with the BLAST, MAFFT, and EMBOSS command line interface programs installed. It was designed to find resistance mutations in a FASTA format input file containing the genomic DNA sequence of an organism. The DNA sequence may be split across many FASTA sequences in the input file. The pipeline is meant to be used with assembled genomes, and has therefore not been tested with raw sequence data. However, it was written to work with genomes in various stages of assembly, and can thus handle draft genomes. The workflow of the pipeline is explained in section 3.1, and illustrated by means of a flowchart in figure 3.1. The code is presented in appendix B.

ARM-find was built to find antibiotic resistance mutations, but the word *mutation* implies change, which can be a bit misleading. What the pipeline actually does is identify nucleotides and amino acids that are different in the input genome compared to a reference sequence. When mutations or substitutions are mentioned in this report, it would be more appropriate to say "difference between input genome and reference sequence". Working this distinction into every sentence would make for cumbersome reading, however, which is why I am clarifying the nomenclature now.

3.1 Pipeline Workflow

When searching for antibiotic resistance mutations, the pipeline considers one target sequence (e.g. the DNA sequence encoding an enzyme subunit) at a time. For each target sequence, the pipeline goes through the following main steps:

- 1. Extracting the appropriate region of DNA, corresponding to a reference DNA sequence (most likely a gene) with known resistance mutations, using BLAST.
- 2. Global alignment of the extracted sequence (and its translated amino acid sequence if the target is a protein) against the reference sequence, using MAFFT.
- 3. Mutation calling, by comparing the extracted sequence to the reference sequence at each position.
- 4. Comparison of found mutations to a database of known resistance mutations.



Figure 3.1: Flowchart showing the workflow of the pipeline. Processes in green and blue are part of a loop that is repeated for each target sequence associated with resistance mutations for the chosen species.

3.1.1 Sequence Extraction

To identify antibiotic resistance mutations in a specific DNA sequence in a genome, that sequence must first be found in the genome, which can be accomplished through sequence alignment against a reference sequence. The reference sequences for gyrA, gyrB, parC, and parE are from *E. coli* K-12 MG1655, and were downloaded from the PATRIC database [18]. Since genes and genomes are obviously not similar in length, global alignment is not an option. For that reason, a local alignment tool is used to find the correct sequence in the genome in this pipeline - more specifically, BLAST [13], as it is the most widely used tool for local alignments, and contains all the functionality needed for this part of the pipeline. The **blastn** command is used in order to run Megablast, which is suitable for highly similar sequences (approximately 95 % sequence identity) [19]. All parameters are left at default values. The output format is set to XML (-outfmt 11) to facilitate parsing.

This first local alignment would in many cases be enough for mutation calling, as most nucleotides would be comparable between the sequences directly from the alignment. However, there are some cases which demand sequence extraction and global alignment before mutation calling. For instance, if there are mutations at the end of the sequence being aligned, those might end up not being included in the BLAST alignment. Similarly, if there is a large mismatching region, or large insertions or deletions, in the aligned sequences, BLAST will tend to split the alignment into two hits, one on each side of the mismatching region. Both of these cases would result in mutations being missed during mutation calling, and possibly even in missed frameshifts that result in a completely different amino acid sequence after translation. For these reasons, the BLAST alignment is used only to find the correct region in the genome, which is then extracted for later global alignment.

It is quite possible for the pipeline to extract more than one sequence from a BLAST alignment, for instance when a gene is split over two or more FASTA sequences in the genome. These sequences are treated separately, rather than merging them, as the final report generated by the pipeline includes what FASTA sequence in the genome any given mutation was found on. Naturally, this would not be practical if hits from different FASTA sequences were merged at this stage. Additionally, multiple sequences in the extraction can be a result of multiple copies of a gene, or of several hits in one gene, separated by a mismatching region. In the latter case, the hits are normally merged, as discussed in section 3.1.1.2, but there are scenarios where this is not desirable, as well as those where it simply fails.

3.1.1.1 BLAST Hit Extension

As the alignments made by BLAST may not extend all the way to the end of the reference sequence, extension is often necessary before extraction of the sequence. This is done simply by checking the first and last base number of the reference sequence in a BLAST hit, and whether those numbers correspond to the first and last base numbers of the entire reference sequence. If they do not, then the BLAST

hit does not cover the entire reference sequence, and may need to be extended, as shown in figure 3.2. The number of bases not covered on either side of the BLAST hit is calculated, and the pipeline attempts to retrieve those bases from the genome one by one, taking into account what strand the match was found on to determine direction in the genome and whether the base at a certain position should be taken, or its complement. BLAST hit extension will stop if the pipeline tries to access bases outside of the range of the FASTA sequence the hit was found on in the genome, or if a base in the genome is already part of another BLAST hit. The latter restraint is in place to avoid overlap between the BLAST hits with respect to the bases in the genome. This only applies to overlap in the genome, though. Overlap in the reference sequence between hits can be caused by multiple copies of a gene, for instance, and should therefore be included, so that the pipeline can report whether mutations to a given reference sequence position in the genome is the same in each instance of reference sequence overlap.



Figure 3.2: During hit extension, the number of reference bases missed on both sides is counted, and extension of that length is attempted. Should the end of the genomic FASTA sequence (or another BLAST hit) be encountered first, extension will stop.

Sometimes, BLAST will find erroneous hits that tend to match only a short part of the reference sequence. When these are extended to cover as much of it as possible, the result can be that thousands of incorrect bases are added. To combat this, the pipeline will discard all hits that are extended more than 20 % of the reference sequence length, unless they meet one of the following criteria:

- 1. The hit neighbours another hit in the genomic sequence, after all hits have been extended. This preserves hits that would otherwise be deleted for extending over large insertions, for example in scaffolds.
- 2. The hit is the result of a merge of several hits, as detailed below in section 3.1.1.2. In this case, a hit could seem to be extending too far, whereas it in reality is just extending to cover the hits it was merged with.

3.1.1.2 Merging BLAST Hits from the Same FASTA Sequence in the Genome

There are two cases that will make BLAST find two or more correct hits in one FASTA sequence from the genome. The first is if there are multiple copies of a gene, in which case the hits should remain separate. The second case is when hits are separated by a mismatching region. In the latter case, it is useful to merge the hits.

To illustrate the problems that can be caused by not merging such hits, imagine an alignment with two hits separated by a large insertion. During hit extension, the number of bases to extend is decided based on where on the reference sequence the hit is, and does not take into account the size of the insertion. In most cases, the hits will simply be extended until they come to a base covered by the other hit, but it is also possible to miss part of the insertion during extension, depending on its size and the positions of the hits on the reference sequence. A second problem arises when the insertion size is not evenly divisible by three, thereby causing frameshift. If the hits were merged, translation of the resulting sequence would lead to a drastically different amino acid sequence, as it should. If the hits aren't merged, the frameshift would only show in one of the hits, or it might be missed completely.

In order to tell the difference between hits that should be merged and hits that should not, there are a number of checks in place. For this explanation, let us call the distance between two adjacent hits in the genome genome distance, and the distance between the same hits in the reference sequence reference distance (illustrated in figure 3.3). If reference distance < 100, the hits will be merged. This allows for some leeway for deletions and substitutions, both of which cause increased reference distance. It also allows any insertion size, as insertions do not change the reference distance. In addition to the above criterion, any pair of adjacent hits that satisfy the condition |genome distance - reference distance| < 100 will be merged. This allows for combinations of substitutions and insertions to pass more freely.



Figure 3.3: Two BLAST hits mapping to different parts of the reference sequence, with reference distance and genome distance explained.

Even if the criteria above are met, there are a few more checks that must be cleared for hit merging to occur. In order to avoid merging hits from different copies of a gene, as well as to avoid other complications, the hits must:

- be on the same DNA strand in the genome.
- satisfy the condition genome distance \leq reference sequence length.
- not overlap in the genome.
- come in the expected order in both the genome and in the reference sequence, with respect to which strand they are on. Out of order hits are likely either bad hits or parts of different gene copies.

• not be separated by a region with a high percentage (>30 %) of N's. This stops merging of hits separated by regions of uncertain lengths in scaffolds, where merging is likely to cause erroneous frameshift.

If there are more than two hits in a FASTA sequence in the genome, each adjacent pair of hits will be considered for merging separately. Then, unbroken chains of pairs accepted for merging will be marked for merging into one hit, as shown in figure 3.4.



Figure 3.4: Example of merging multiple hits from one FASTA sequence. Each pair is evaluated for merging separately. The pairs are then added together until a pair that was not accepted for merging is encountered. In this case, all pairs were accepted for merging except 3-4, which results in two final hits, one with hits 1-3, and one with hits 4 and 5.

Finally, when all merging decisions are finished, all hits except the last one in each merge set are deleted, and the value of genome distance – reference distance is saved for each merge pair. The last hit can then be extended as explained in section 3.1.1.1, with the only change being that extension distance is increased by the saved distance values for every involved merge pair, as shown in figure 3.5. This compensates for the change in length due to insertions or deletions between the merge pairs. If the saved distance value is negative, the extension will be shortened instead of lengthened, which is necessary for hit merging when the hits are separated because of a deletion, rather than an insertion or substitutions.



Figure 3.5: Once a group of hits to be merged has been decided, only the last hit (here, hit 2) is kept. Extension length is modified by genome distance – reference distance. This difference is indicated in black. Regular extension would only be based on the distance to the end of the reference sequence, which for hit 2 is as long as the green and white segments of the reference sequence. Extra extension (black) is necessary when there is an insertion between the hits, as in this case.

3.1.1.3 Post-Extraction Sequence Modifications

Once the appropriate DNA sequences have been extracted from the genome, some processing is necessary in preparation for global sequence alignment. If the sequence extraction yielded hits that do not cover the entire reference sequence, then the length of those hits will be shorter than the reference, which is suboptimal for a global alignment. Apart from making mutation calling more arduous, it could also result in faulty alignments, with the hit being stretched to cover more of the sequence than it should. To solve this, all hit sequences that do not cover the entire reference sequence are padded with N's, signifying unknown bases. To calculate how many N's should be added to each side of the sequence, the number of bases added during hit extension is subtracted from the number of bases that the original BLAST hit missed from the reference sequence, for each side. With this information, the hit can be padded with the appropriate amount of N's on either side. It is still possible for the hit and the reference sequence to be of different lengths, but only because of indels, which is as it should be.

Another benefit of N padding is avoiding frameshift during translation. If a hit does not cover the beginning of a gene and is not padded, the likelihood of an erroneous frameshift is high. Padding with the correct amount of N's solves that problem. Hence, translation of hit sequences is done *after* padding, if they code for proteins.

Translation is performed with the transeq command of the EMBOSS package [20]. All reference sequences that are to be translated start with a start codon and end with a stop codon in the pipeline's sequence database. The BLAST alignment, hit extension, and sequence padding discussed above make sure that the hits are correctly lined up against the reference sequence, starting with the start codon, and ending with the stop codon, which means that erroneous frameshift is unlikely to cause problems. Hence, transeq is run with the argument -snucleotide1 to limit translation to the first reading frame. Furthermore, the argument table=11 is used to set the translation table to that of bacteria.

While the pipeline database currently only contains protein-coding sequences, it is possible that future additions might include DNA sequences that do not code for proteins, such as promoters or sequences coding for rRNA. To distinguish between sequences that should be translated and sequences that should not, the pipeline checks the **mutations** database, to see if there are any resistance mutations for amino acids for a given target. If there are, then the sequence must code for a protein, and should be translated. If there are not, then it should not be translated. This holds true as long as no protein-coding sequences without resistance mutations for amino acids are added to the **mutations** database, but since there would be no point in adding such a sequence, this is not a problem.

Finally, the range of any overlaps between the BLAST hits are saved for later reporting. At this stage, only nucleotide sequence overlap is saved, as the BLAST alignment files provide sufficient information for finding said overlap, whereas the amino acid sequences need to be aligned before overlap can be calculated.

3.1.2 Global Alignment

For the purposes of this pipeline, there were only a few specific requirements for a global alignment tool. Since the sequence extraction can yield multiple hits, it would be much more practical to work with multiple alignment, rather than aligning every hit sequence against the reference sequence one by one. Additionally, to be able to call mutations that add or remove stop codons, it was necessary for the alignment tool to not only be able to accept nucleotide and amino acid codes, but special characters as well, such as the asterisk (*) used to demark a stop codon. MAFFT [14] meets these requirements, and was thus chosen for the role. MAFFT is run and adapted for global alignment through use of the ginsi command, with all parameters at default values, apart from the optional argument --anysymbol. This is necessary in order to include stop codon asterisks in the alignment. Global alignment is done for both the nucleotide sequence and, in case the target is a protein, the amino acid sequence, to enable mutation calling. After global alignment of the amino acid sequence, the ranges of potential overlaps between translated BLAST hits are saved for later reporting.

3.1.3 Mutation Calling and Identification of Resistance Mutations

During mutation calling, DNA and protein sequences are handled separately, and one hit at a time. For each hit, every position in the sequence is compared one by one. For each position in the reference sequence that is not a hyphen, signifying a gap, a counter is incremented. This is done to keep track of which number in the reference sequence a certain position in the alignment corresponds to, so that the correct position of mutations can be reported. During mutation calling, information on substitutions, insertions, and deletions is saved, but also information on what positions in the reference sequence are covered by the hits. This information is later used to generate reports on mutation coverage.

Once all mutations have been called for a given target sequence, they are compared with all resistance mutations in the pipeline's database for that target. If there are matches, the relevant antibiotic (or antibiotics) is associated with that mutation in the report. Since the **mutations** database could contain resistance mutations that the reference sequence itself has, all resistance-associated positions are saved during mutation calling. When comparing called mutations to the database, the resulting "non-mutations" (e.g. S83S) are skipped, unless they confer antibiotic resistance.

The database is split into two files, targets and mutations, in order to minimise repetition of data that could increase the risk of mistakes during data entry. The targets file simply associates every target DNA sequence with a unique identifier (target ID), and lists what species the sequence is from (see table 3.1).

Target ID	Species	Name
target0000001	E. coli	DNA gyrase - subunit GyrA
target0000002	$E. \ coli$	DNA gyrase - subunit GyrB
target0000003	E. coli	Topoisomerase IV - subunit ParC
target0000004	E. coli	Topoi somerase IV - subunit \mbox{ParE}

Table 3.1: The targets database lists all target sequences, along with with the associated species.

The mutations database contains the resistance mutations, where every entry is linked to the appropriate row in the targets database by its target ID. For every mutation in the database, there is information on what kind of antibiotics the conferred resistance applies to, its position in the target sequence, what nucleotide or amino acid occupies it in the reference sequence, and what the substituted nucleotide or amino acid is. A few example lines from the mutations database are shown in table 3.2.

Table 3.2: The mutations database lists all antibiotics resistance mutations, and is linked to the targets database by target ID.

Target ID	Antibiotics	Nucleotide/ protein	Position	Reference nucleotide/ amino acid	Mutation
target0000001	Quinolones	prot	84	A	P
target0000001	Quinolones	prot	84	A	V
target0000003	Quinolones	prot	57	S	T

3.1.4 Reporting

Three kinds of reports are created. The first one lists the mutations that were found, and whether or not these are associated with antibiotic resistance. The second report lists all known resistance mutation positions that the pipeline was unable to evaluate in the input genome. This can either be because the position was not covered by the BLAST hits, or because of an ambiguous base code. Both cases mean it is unknown whether a resistance mutation is present or not. The second report is included so that the user will know if the absence of a resistance mutation in the first report means that the mutation is not present, or if it just means that no information on it is available. Finally, the third report lists all target sequences that are not covered, in order to clearly show when a target sequence is not covered by any BLAST hits.

3.2 Pipeline Arguments

The behaviour of the pipeline can be controlled with several arguments, which are listed in table 3.3. The only required arguments are --infile and --species.

Table 3.3: Arguments of the pipeline. Mandatory arguments are underlined.

Argument	Short form	Explanation		
infile	-i	Path of the genome file to be analysed.		
out	-o	Takes path to output directory. Default: create output directory in current working directory. If existing path is specified, files in that directory may be removed if names conflict.		
species	-s	Takes abbreviated name of the species of the input genome (e.g. "E. coli" - quotation marks are necessary due to the space in the name).		
list_species		Lists all currently supported species and exits.		
verbose	-v	Sets output level to verbose.		
quiet	-q	Suppresses all printed output.		
logfile	-1	Redirects output (verbose) to log file. If no path is specified, it is placed in the output directory.		
ext_program_output		Includes external programs' output in the pipeline's output.		
keep_tmp_files		Prevents deletion of temporary files (alignments, translations etc.).		
BLAST_perc_identity		Takes a float ($0 \le x \le 100$) for "percent identity cut-off" in BLAST. Default is the same as Megablast default (95 %)		
report_all_coverage		Includes all missing positions in coverage report, not just those related to antibiotic resistance.		

3.3 Substitution Study in E. coli and Shigella

All RefSeq genomes for $E. \ coli$ and every species of the *Shigella* genus were downloaded from NCBI [21]. All RefSeq genomes are complete, but can be in varying stages of assembly, making them suitable for testing the pipeline. Knowledge of fluoroquinolone resistance mutations in $E. \ coli$ most likely applies to the genus *Shigella* as well, due to its close relationship to $E. \ coli$. Before pathogenic forms of $E. \ coli$ were discovered, *Shigella* was classified as a separate genus due to its clinical significance. Later research has clearly shown that they are one species [22]. Therefore, *Shigella* was included in this study.

The downloaded genomes were all analysed in the pipeline for fluoroquinolone resistance mutations, generating data on mutations in the DNA gyrase subunit genes gyrA and gyrB, as well as the topoisomerase IV subunit genes parC and parE. All subsequent analysis was done for *E. coli* and *Shigella* separately. The number of substitutions in every position in both the nucleotide sequence and the amino acid sequence for all four subunits was summed up across the genomes. All substitutions found after an indel in a genome were discarded in order to avoid counting substitutions caused by frameshift, as global alignment of a frameshifted amino acid sequence produces very erratic results. Additionally, since overlapping BLAST hits can result in multiple substitutions in the same reference position in one genome, the extra substitutions in every reference position with more than one substitution were counted. This was done in order to check if the percentage of genomes with substitutions would be overestimated, and if so, by how much. For the amino acid sequences, the most frequent substitution positions (>3 % of genomes) were also checked for which amino acids the substitutions were to. For the nucleotide sequences, analysis of the ratio of synonymous versus non-synonymous mutations was performed. This was done by checking whether an amino acid substitution resulted for each single nucleotide substitution. Analysis and graph creation was done in R and Python.

3. Methods

4

Results

In this section, two different kinds of results are presented. The first is an example of the output of the pipeline, and the second is the results of the substitution study.

4.1 Pipeline Output

G239C

G239C

T1372-

nucl

nucl

nucl

Quinolones

Quinolones

Quinolones

ParC

ParC

ParE

A few example lines from a report file that could be generated by the pipeline are shown in table 4.1. Target names have been abbreviated to fit the page, and would normally read "DNA gyrase - subunit GyrA" etc. "Involved antibiotics" shows which classes of antibiotics are present among the resistance mutations for a given target in the database. Called mutations are listed under "Mutation". The first letter indicates the base in the reference sequence, the following number its position in the reference sequence, and the remainder the mutation. Hyphens represent deletions, and multiple letters indicate insertions. For instance, "C231CT" means a T was inserted after the C at position 231. "Nucl/prot" simply indicates whether a given mutation is in the DNA or the amino acid sequence. Should a mutation (e.g. S83L in GyrA) confer resistance to an antibiotic, or antibiotics, they will be listed under "Resistance".

Target name	Involved antibiotics	Mutation	$\operatorname{Nucl}/\operatorname{prot}$	Resistance	$\begin{array}{l} {\rm FASTA} \\ {\rm sequence(s)} \end{array}$	Mutation prevalence	Sequence coverage
GyrA	Quinolones	C248T	nucl		'1'		Full
GyrA	Quinolones	S83L	prot	Quinolones	'1'		Full
GyrB	Quinolones	C231CT	nucl		'1'		Full
ParC	Quinolones	G239T	nucl		'2', '3', '3', '4'	0.25	Partial, w/ start

'3', '3', '2', '4' '3', '3', '2', '4'

'5'

0.5

0.5

Table 4.1: Example lines from a report file that could be generated by the pipeline.

"FASTA sequence(s)" lists the FASTA sequence in the input genome that the mutation was found on. If a reference sequence position is covered by two or more BLAST hits, the FASTA sequences they were found on will all be listed. For multicopy genes within one FASTA sequence, the same sequence can be listed more than once. For every mutation whose position is found on several FASTA sequences,

Partial, w/o start

Full

Full

the first sequence listed under "FASTA sequence(s)" will be the sequence where the mutation was found. This is illustrated by the three mutations in position 239 in ParC in table 4.1. Note that two were found on the same FASTA sequence, and that the position was covered four times. Since only three mutations are reported, there must be one place where the base is not mutated.

Whenever a reference sequence position is covered by more than one BLAST hit, the mutations at that position, or lack thereof, can be different from each other. The degree to which the mutations are the same can be read from the "Mutation prevalence" column, where

 $Mutation prevalence = \frac{Number of BLAST hits with same mutation}{Number of BLAST hits covering given position}$

Taking the example of position 239 in ParC again (table 4.1), G239C has a mutation prevalence of 0.5, since it occurs twice among four hits (including the non-mutated base), while the same value for G239T is 0.25 (one mutation in four hits).

Finally, "Sequence coverage" provides information on how much of the reference sequence was covered by the extended BLAST hit. The possible levels are "Full", "Partial w/ start", and "Partial w/o start". Whether the start of the sequence is included or not in a partial hit is of interest, as the reading frame cannot be guaranteed to be correct if the start of the sequence is missing in a hit. There may be indels that change it before the start of the hit.

Positions that are not covered by the extended BLAST hits are listed in the mutations not covered report file (see table 4.2). By default, only positions where mutations can confer antibiotic resistance are listed, but all non-covered positions can be included through the use of an optional argument (--report_all_coverage) to the pipeline. In case no positions in a target sequence are covered, that sequence's name will be listed in the targets not covered report (see table 4.3).

Table	4.2:	Excerp	t from	a mut	ations	not	covered	report i	file,	with	optional	argume	ent g	given to
include	e posi	itions no	ot relat	ed to	antibio	tics	resistance	mutatio	ons.	Targ	get names	s have b	been	abbre-
viated.														

Target name	Involved antibiotics	Not covered	Nucl/prot	Resistance
ParE ParE	Quinolones Quinolones	$458 \\ 459$	prot prot	Quinolones
ParE	Quinolones	460	prot	Quinolones

Table 4.3: Example of a	targets not	covered report file.
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Target name	Involved antibiotics
DNA gyrase - subunit GyrB	Quinolones
Topoisomerase IV - subunit ParC	Quinolones
Topoisomerase IV - subunit ParE	Quinolones

4.2 Substitution Study Results

Due to the similarity of the results between $E. \ coli$ and Shigella, and the large number of graphs, the results for $E. \ coli$ will be presented first, and in greater detail. A comparison between $E. \ coli$ and Shigella may be found in section 4.2.3.

Running all 4660 *E. coli* genomes through the pipeline sequentially took 5 hours, 41 minutes, and 21 seconds, meaning the average analysis time was 4.4 seconds per genome. This is by no means a controlled measurement, as no care was taken to limit other processes running on the system, and all the genome files were copied and unzipped during the run, but it serves as an indication of approximately how long it takes the pipeline to search a genome for fluoroquinolone resistance mutations.

4.2.1 Substitution Analysis

There was considerable variability in the nucleotide sequence of the analysed subunits, as shown in figures 4.1 (gyrA), 4.3 (gyrB), 4.5 (parC), and 4.7 (parE), where the substitution percentage is shown in grey. As an overlay on the graphs, the percentage of genomes with single nucleotide substitutions that cause amino acid substitutions is shown in red. These results clearly show that most substitutions are synonymous. Comparing the non-synonymous nucleotide substitutions to the amino acid substitutions - see figures 4.2 (GyrA), 4.4 (GyrB), 4.6 (ParC), and 4.8 (ParE) - it is clear that they match well.

Of all the *E. coli* genomes, 70 contained insertions in the topoisomerase genes. Many of these were checked to see if the insertion identification was correct, which resulted in two common insertion types being identified. The first was the insertion of a large amount of N's, typical of a scaffold where the sequence of bases between two contigs is unknown, and the distance is estimated. These cases did not cause frameshift, as they resulted in one BLAST hit on either side of the insertion. Those hits were protected from merging due to the high N content of the mismatching region. This is exactly the case that feature was designed for. In the second type of insertion, which caused frameshift, the insertion was part of a repeat sequence of one base, such as an extra A in a sequence of several A's, constituting a homopolymer.

Because some genomes (90 in the case of *E. coli*) contained overlapping BLAST hits, there were cases of multiple nucleotide substitutions in one reference position in the same genome. For the vast majority of these cases, there were only one or two extra substitutions in a reference position across all genomes. Ten reference positions had more than ten extra substitutions, with the highest amount being 35. That means that among all the nucleotide substitution graphs for *E. coli* (figures 4.1, 4.3, 4.5, and 4.7), there is one bar that is 0.75 percentage units (35/4660) too high due to overcounting of substitutions, and that some other bars are affected, but to a much lesser extent. Most of these extra substitutions (~89 % of affected positions) were synonymous mutations. For the amino acid sequences, only eight reference positions had extra substitutions, with the highest amount being two.



Figure 4.1: Substitution percentage for the DNA sequence of gyrA in *E. coli*. Red bars denote non-synonymous mutations.



Figure 4.2: Substitution percentage for the amino acid sequence of subunit GyrA in DNA gyrase in *E. coli*.



Figure 4.3: Substitution percentage for the DNA sequence of gyrB in *E. coli*. Red bars denote non-synonymous mutations.



Figure 4.4: Substitution percentage for the amino acid sequence of subunit GyrB in DNA gyrase in *E. coli*.



Figure 4.5: Substitution percentage for the DNA sequence of *parC* in *E. coli*. Red bars denote non-synonymous mutations.



Figure 4.6: Substitution percentage for the amino acid sequence of subunit ParC in topoisomerase IV in *E. coli*.



Figure 4.7: Substitution percentage for the DNA sequence of *parE* in *E. coli*. Red bars denote non-synonymous mutations.



Figure 4.8: Substitution percentage for the amino acid sequence of subunit ParE in topoisomerase IV in *E. coli*.

4.2.2 Amino Acid Breakdown of Substitutions

For most amino acid positions where the substitution percentage was above 3 % (lower percentage positions were not considered), there was only one amino acid that the substitutions were to, or one that composed the vast majority of the substitutions, as shown in figures 4.9 (GyrA), 4.10 (GyrB), 4.11 (ParC), and 4.12 (ParE). The amount of genomes with missing data was low for every position. Missing data can be a result of lack of coverage by BLAST hits, or ambiguous base codes in the triplet, both resulting in the unknown amino acid X during translation. It can also be caused by indels prior to the position in question, as all positions after an indel are disregarded to avoid counting substitutions caused by frameshift and the resulting bad global alignment.

In terms of fluoroquinolone resistance, many of the genomes had resistance mutations in positions 83 and 87 in GyrA (figure 4.9), 80 and 84 in ParC (figure 4.11), and 529 in ParE (figure 4.12). No resistance mutations were found in GyrB in any genome.



Figure 4.9: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for GyrA in *E. coli*, with amino acid breakdown. White diagonal lines indicate fluoroquinolone resistance. Right-hand bars show percentage of genomes with unknown amino acids.


Figure 4.10: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for GyrB in *E. coli*, with amino acid breakdown. None of the detected substitutions were associated with resistance. Right-hand bars show percentage of genomes with unknown amino acids.



Figure 4.11: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for ParC in *E. coli*, with amino acid breakdown. White diagonal lines indicate fluoroquinolone resistance. Right-hand bars show percentage of genomes with unknown amino acids.



Figure 4.12: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for ParE in *E. coli*, with amino acid breakdown. White diagonal lines indicate fluoroquinolone resistance. Right-hand bars show percentage of genomes with unknown amino acids.

4.2.3 Comparison between *E. coli* and *Shigella*

The same types of graphs were generated for *Shigella* as for *E. coli*. In the interest of readability, they may be found in appendix A.

In the Shigella genomes, fewer nucleotide positions had substitutions than in the $E.\ coli$ genomes, as evidenced by comparing figures 4.1, 4.3, 4.5, and 4.7 ($E.\ coli$) to figures A.1, A.3, A.5, and A.7 (Shigella). However, the substituted positions had a higher substitution percentage in general. Just like for $E.\ coli$, synonymous substitutions were much more common than non-synonymous ones. The profiles of non-synonymous mutations in figures A.1, A.3, A.5, and A.7 also matched the amino acid substitutions in figures A.2, A.4, A.6, and A.8 well, apart from one position. In gyrA, position 1270 has a non-synonymous mutation (C1270T) in approximately 15 % of the analysed genomes (figure A.1), whereas the corresponding amino acid (L423) is not substituted in any genomes (figure A.2). However, reviewing the raw mutation data (not shown), it became clear that every genome with a mutation in nucleotide 1270 also had a C1272G substitution, in the same codon. Together, these two mutations result in an unchanged amino acid, even though the C1270T mutation alone would have resulted in an amino acid substitution from leucine to phenylalanine, as shown in table 4.4.

The amino acid substitutions that are not related to fluoroquinolone resistance were mostly different between $E. \ coli$ (figures 4.9, 4.10, 4.11, and 4.12) and *Shigella*

Mutation	Codon	Amino acid
None (reference)	CTC	Leucine
C12701 C1272G	CTG	Leucine
C1270T and C1272G $$	TTG	Leucine

Table 4.4: The resulting codons and amino acids for all combinations of C1270T and C1272G mutations in *gyrA*.

(figures A.9, A.10, A.11, and A.12). In terms of resistance mutations, the same positions had substitutions in GyrA in *Shigella* as in *E. coli*, although with slightly lower substitution percentages in *Shigella*. Neither *E. coli* nor *Shigella* genomes had any resistance mutations in GyrB. For ParC, *Shigella* genomes had resistance mutations in position 84, but not in 80, unlike *E. coli*. Finally, there were no resistance mutations in ParE in *Shigella* genomes, whereas *E. coli* genomes had resistance-associated substitutions in position 529.

4. Results

5

Discussion

In this thesis, I have constructed a pipeline that automatically identifies resistance mutations in complete genomes, including draft genomes. I have created a database for resistance mutations that currently contains chromosomal mutations conferring resistance to fluoroquinolones in *E. coli*, but is easily extensible to cover more classes of antibiotics and other organisms. In order to test the pipeline, as well as to put it to good use, I have analysed every RefSeq genome available from NCBI for both *E. coli* and *Shigella*, and shown the prevalence of fluoroquinolone resistance mutations in those genomes. In the sections below, I will discuss pipeline design choices and their implications, as well as the results of the RefSeq genome analysis study.

5.1 Pipeline Design

Much work was put into making the pipeline able to handle difficult situations, including genes split over multiple FASTA sequences, multiple copies of genes, similar but unrelated sequence snippets, and intra-gene regions of sizeable dissimilarity. For many of these situations, test cases had to be constructed artificially, as they were not encountered at all during the large-scale genomic analysis of this thesis. While all tested scenarios worked well, real data is preferable for finding oversights. Additionally, since only four genes have been used during testing, it is possible that future additions of antibiotics could unearth new challenges. That being said, I have made an effort to keep the process of the pipeline as general as possible, so that it works for all manner of target sequences.

There are many cases of cut-off values in the pipeline that are used to distinguish between different situations, such as when BLAST hits should or should not be merged, or which hits should be kept and which should be discarded. These values were set to handle situations that were either thought likely to occur, or that were encountered during the large-scale genomic analysis. The values are ad hoc, in that they yield the correct results for the data analysed in this thesis, but have in no way been optimised. It is also important to point out that although the cut-off values work well for the current state of the pipeline, there is no guarantee that they will also work for any target sequences and species that might be added in the future. It is important to note that while the pipeline can find resistance mutations, it provides no information on what level of resistance can be expected from a certain mutation, or combination of mutations. That is left to the user to evaluate. Additionally, one should not assume that the list of resistance mutations for any antibiotic in the database is exhaustive. There may be other factors that contribute to antibiotic resistance as well, such as increased expression of efflux pumps, decreased expression of outer membrane porins, or mobile resistance genes.

The fact that this pipeline uses external programs for different tasks makes it susceptible to future parsing problems. If BLAST or MAFFT were to change the formats of their reports, for instance, there could be compatibility issues. The likelihood of that happening might be low, but it is an unavoidable risk when depending on other programs.

5.1.1 Sequence Extraction

Using Megablast for local alignment relies on the aligned sequences being highly similar (approximately 95 % sequence identity) [19]. This is a reasonable expectation, as sequences are unlikely to differ more in the same species. Additionally, local regions of high dissimilarity do not cause problems here, as BLAST will simply find hits that meet the sequence identity requirement on either side of such regions, which can then be analysed separately. Using the database entries for one species to analyse another could be problematic, but in those cases it would not be safe to assume that the resistance mutations are the same anyway, so the 95 % sequence identity issue should not be a matter of great concern. If the pipeline were to be used on genomes that were not yet assembled, for instance directly on raw sequencing data, the shorter sequence lengths might make the 95 % cut-off more limiting. However, that use case would most likely require extensive reworking of the pipeline, and the BLAST sequence identity cut-off would not be the biggest issue by far.

Another potential problem of the sequence extraction part of the pipeline concerns indels in the beginning and ends of genes. If BLAST does not match the ends of a target sequence, hit extension will retrieve the missed bases. However, if there is an insertion or a deletion in the region outside of the BLAST hit, then the number of retrieved bases will be off. This would result in not covering the first base in case of an insertion, or retrieving bases outside of the gene for deletions. The erroneous bases at the ends would then likely be misidentified as substitutions, rather than indels. However, the likelihood of indels in the beginning and end of genes was deemed to be low. Even so, it might be advisable to double-check mutations in the first and last bases of a sequence, should any be found.

Another challenge posed by indels comes from cases where two BLAST hits are separated by a region that contains an indel. If a hit extends to cover the indel, but does not reach the end of the gene, the number of N's added for padding will be off. Naturally, the worst possible outcome of this is frameshift. In most cases, hit merging will avert the issue completely through the extension distance compensation involved in the process. However, it is possible that the hits will not be merged, most notably if they are separated by a large amount of N's, which can occur in scaffold genomes. For these situations, the "sequence coverage" column of the main report file comes in handy, since the values "Full" and "Partial w/ start" will indicate that the reading frame is correct, as the start of the sequence is covered.

Deletion of bad BLAST hits is based on deleting hits that extend too far. This works well for short sequence matches outside the target sequence, as hit extension will be much longer than what is deemed acceptable by the pipeline, resulting in the deletion of the erroneous hit. However, if BLAST finds a hit just outside of a target sequence's position in the genome, it is theoretically possible for it to be protected from deletion by extending up to a legitimate hit in the correct place in the genome. In the worst case scenario, it might even claim bases in the end of the target sequence, if its hit extension is handled before the legitimate hit. A potential solution to this could be to check that all BLAST hits that are in close proximity to each other are arranged in a logical sequence with respect to what part of the reference sequence they match, and delete any that do not fit. However, as this problem was not encountered during this thesis, it was deemed to be too unlikely to merit the amount of work required to solve it.

5.1.2 Global Alignment

To get the correct reading frame during translation, the pipeline relies on alignment to the start codon, which will always be at the start of the stored target sequences. The problem with this is that the reading frame could be changed by the gene end indels described above. A possible solution to this could be alternative methods of identifying the reading frame, but the low likelihood of gene end indels makes this a most likely unnecessary addition.

When the reading frame is changed due to indels, the global alignment of amino acid sequences runs into problems, as the sequences to be aligned will tend to be very dissimilar. The report will then include a very large of amount of insertions, deletions, and substitutions in the amino acid sequence. Diagnosing this situation is easy when it occurs, though, as a nucleotide insertion or deletion will be immediately followed by all the mismatches in the amino acid sequence. However, any automated use of the pipeline will require checks for indels in the nucleotide sequence, so that frameshifted global alignment results may be handled, just as was done for the substitution study in this thesis.

5.2 Substitution Study

When analysing the substitution percentages in the four subunits, it is important to remember that "substitution percentage" refers to the percentage of genomes that have a different base or amino acid in any given position *compared to the reference sequence*. The reference is not the origin of all other strains, and therefore no conclusions can be drawn about which genomes are mutated; they can simply be compared. One cannot infer mutation likelihood from the graphs presented in this thesis. It should also be noted that the analysed genomes do not necessarily represent all *E. coli* and *Shigella* genomes well. It would not be unreasonable to suspect that clinical isolates are overrepresented among the RefSeq genomes, thereby possibly resulting in an overestimation of the prevalence of resistance mutations across all genomes. After all, the entire reason that *Shigella* is categorised as a genus rather than a subgroup of *E. coli* is its pathogenicity [22], highlighting our focus on effects on humans in our way of thinking of these organisms.

When the insertions that were found in some genomes were investigated, two common insertion types were identified. One of these was the frameshift-causing insertion of one base into a repeat sequence of the same base, such as an extra A inserted into a sequence of several A's. As repeating bases are more difficult to sequence and assemble correctly, these insertions are probably caused by sequencing or assembly problems. Since both DNA gyrase and topoisomerase IV are essential, any gene with frameshift would cause death, if it were the only copy of the gene. Though the genomes with this kind of insertion were not thoroughly checked for extra gene copies, sequencing or assembly problems seem much more likely than multiple gene copies, with one of them being completely useless because of frameshift.

For some reference positions, overlapping BLAST hits resulted in there being more than one substitution to a reference position in the same genome. However, the relatively low amount of extra substitutions means the resulting overestimation of substitution percentages is negligible. This is especially true for the amino acid sequences, where only very few reference positions had extra substitutions, and at most two per position. That is certainly few enough not to be visible in the presented graphs.

The non-synonymous nucleotide substitutions shown in figures 4.1, 4.3, 4.5, and 4.7, would not necessarily have had to match the amino acid substitutions in figures 4.2, 4.4, 4.6, and 4.8 (and likewise for the corresponding *Shigella* figures), as multiple mutations in a triplet could result in an exchanged amino acid where only one mutation would not have. However, as the profiles of the non-synonymous substitutions and the amino acid substitutions are so similar, one may conclude that the vast majority of amino acid substitutions were caused by single nucleotide mutations. The exception to this is the mutations in nucleotide 1270 in *gyrA* in *Shigella*, which by itself is a non-synonymous mutation, as indicated in figure A.1. However, the mutations in nucleotide 1272 negated the amino acid change, as shown by the lack of a substitution in amino acid 423 in figure A.2. The fact that synonymous mutations vastly outnumbered non-synonymous mutations was expected, as most non-synonymous mutations would likely reduce fitness. A synonymous mutation, on the other hand, can be expected to have much less effect on fitness, resulting in more mutations that remain.

The top substitution percentages were noticeably higher among the *Shigella* genomes than among the *E. coli* genomes. Since the *Shigella* genus is generally accepted to be more appropriately characterised as a part of the *E. coli* species [22], it seems reasonable that there would be more similarity among *Shigella* than the larger encompassing species of *E. coli*, resulting in some very high substitution percentages. If similar subgroups of *E. coli* were to be analysed separately from the species, similar results would likely emerge. However, it might also be possible that the apparent greater diversity of *E. coli* could be caused by the larger amount of genomes analysed - 4660 compared to *Shigella's* 812.

5.2.1 Fluoroquinolone Resistance Mutations

Previous studies have found that the most common resistance mutation in GyrA in $E.\ coli$ is in position S83 (especially the substitution S83L), followed by D87 [4]. This matches the results of this thesis exactly. For strains with single mutations in GyrA, S83 substitutions have been found to confer significantly higher fluoroquinolone resistance than D87 substitutions, providing a likely reason for the higher prevalence of the former [4]. Studies have also found that S83L results in higher resistance than S83A, and that D87N yields higher resistance than D87G and D87Y [4]. This provides a plausible explanation for the fact that S83L and D87N were much more common than the other mutations in the same positions in this study.

In spontaneous *in vitro E. coli* mutants, gyrB nucleotide substitutions have been shown to be approximately as common as gyrA nucleotide substitutions, while the latter dominated clinical isolates [4]. The substitution study showed no resistance mutations in gyrB at all, while gyrA resistance mutations were common. This could possibly be an indication of clinical isolates being overrepresented among RefSeq genomes.

In ParC, the most commonly reported resistance mutations are in position S80 (especially the substitution S80I), followed by E84 [4]. Again, this matches the results of the substitution study. Since ParC is homologous to GyrA, with S83 and D87 in GyrA corresponding to S80 and E84 in ParC [23], this similarity to the GyrA resistance mutations was expected. In previous studies, ParC and ParE mutations have not been found without GyrA mutations in *E. coli*, probably because the fluoroquinolone susceptibility of DNA gyrase needs to be reduced before topoisomerase IV mutations can affect resistance [4]. This could explain why there are fewer S80 and E84 substitutions in ParC than S83 and D87 substitutions in GyrA. It would have been interesting to check whether the GyrA mutation requirement holds true for all the *E. coli* RefSeq genomes, but time constraints prohibited it.

5. Discussion

6

Conclusion

I have constructed a pipeline (ARM-find) that works well for finding fluoroquinolone resistance mutations in both $E.\ coli$ and Shigella genomes, including draft genomes. Its resistance mutation database is easily extensible, allowing for the identification of resistance mutations for any antibiotic in any species, provided that it is known which mutations are relevant. The analysis time required to analyse a genome is short, making ARM-find suitable to run on consumer-grade computers. Through scripted use of the pipeline, I have been able to discover that a large portion of $E.\ coli$ and Shigella RefSeq genomes contain fluoroquinolone resistance mutations. The relative frequencies of those resistance mutations matched was has been previously reported, and the most common resistance mutations were the ones that lower susceptibility to fluoroquinolones the most.

6. Conclusion

7

Future Work

ARM-find functions as was intended from the start, but there are improvements that could be made. The most obvious improvement is of course the addition of more species and classes of antibiotics, to increase the usefulness of the pipeline for resistance mutation identification. This would provide the added benefit of allowing testing of situations that do not occur when just looking for fluoroquinolone resistance mutations in *E. coli*. For instance, future reference sequences might be more difficult to align, due to things like segments that are also found elsewhere in the genome. Testing situations like these would allow for the optimisation of all parameter values in the pipeline, which are currently chosen in an ad hoc manner that works for fluoroquinolone in *E. coli*.

Another possible feature that comes to mind is automatic species identification. Currently, the species of the input genome has to be set with the **--species** argument. The rationale for this was that the pipeline is meant to be used with assembled genomes, and it seems unlikely that the species would remain unknown after assembly. If the pipeline were to be adapted for use with raw sequencing data, automatic species identification would make much more sense, as it would be more plausible that the species would be unknown in that scenario. Allowing for input of raw sequencing data would be a big improvement for the pipeline, for several reasons. It would reduce the level of expertise required to use it, and it would probably reduce the amount of time from sequencing to resistance mutation identification, allowing for speedier determination of suitable treatment in hospitals. Reducing the time required for resistance identification is of course a big part of why this pipeline was made in the first place. However, this feature would probably require extensive reworking of the workflow. If the reads from the sequencing data are not sufficiently long, which they most likely would not be, some assembly might be required in the pipeline. That would not be an easy feature to implement, but it would certainly be useful. Another option could be to align the reads against the reference sequence, and modify reporting so that it would not produce one line for every mutation.

One feature that already exists, but could be improved upon, is handling of overlapping sequences. The most important addition would be identification of multiple gene copies. Currently, whether a mutation is covered elsewhere in the genome or not is made clear by the report through the FASTA sequence(s) and mutation prevalence columns. However, no information is provided on why the position is covered multiple times. Identification of multiple gene copies would be very useful, as a resistance mutation in one gene copy might not mean the same thing if there are non-mutated copies, as it would have if there were only one copy of the gene. That sort of information would aid in determination of whether the analysed bacteria are clinically resistant to a certain antibiotic or not.

I would like to slightly rework the **mutations** database to include information for every mutation on the references that support its inclusion. This would simplify future maintenance of the database, as well as lend it more credibility. The easiest way to implement this would probably be the addition of a column of comma-separated Digital Object Identifiers (DOIs) to the articles that the resistance mutations are referenced in.

These are just some of the features that would be useful in ARM-find. Other examples include making it work on non-Linux platforms, or even creating a website where analysis could be run on servers, instead of the user's computer. While certainly useful, those features are less likely as future additions than the ones discussed above.

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А

Shigella Results



Figure A.1: Substitution percentage for the DNA sequence of gyrA in *Shigella*. Red bars denote non-synonymous mutations.



Figure A.2: Substitution percentage for the amino acid sequence of subunit GyrA in DNA gyrase in *Shigella*.



Figure A.3: Substitution percentage for the DNA sequence of gyrB in Shigella. Red bars denote non-synonymous mutations.



Figure A.4: Substitution percentage for the amino acid sequence of subunit GyrB in DNA gyrase in *Shigella*.



Figure A.5: Substitution percentage for the DNA sequence of *parC* in *Shigella*. Red bars denote non-synonymous mutations.



Figure A.6: Substitution percentage for the amino acid sequence of subunit ParC in topoisomerase IV in *Shigella*.



Figure A.7: Substitution percentage for the DNA sequence of *parE* in *Shigella*. Red bars denote non-synonymous mutations.



Figure A.8: Substitution percentage for the amino acid sequence of subunit ParE in topoisomerase IV in *Shigella*.



Figure A.9: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for GyrA in *Shigella*, with amino acid breakdown. White diagonal lines indicate fluoroquinolone resistance. Right-hand bars show percentage of genomes with unknown amino acids.



Figure A.10: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for GyrB in *Shigella*, with amino acid breakdown. No resistance mutations were detected. Right-hand bars show percentage of genomes with unknown amino acids.



Figure A.11: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for ParC in *Shigella*, with amino acid breakdown. White diagonal lines indicate fluoroquinolone resistance. Right-hand bars show percentage of genomes with unknown amino acids.



Figure A.12: Substitution percentage for all amino acid positions with substitutions in >3 % of genomes for ParE in *Shigella*, with amino acid breakdown. No resistance mutations were detected. Right-hand bars show percentage of genomes with unknown amino acids.

В

Pipeline Code

```
1 #!/usr/bin/env python3
  # ARM-find pipeline
2
3
  #
4
  # Overview
  \# This pipeline takes a genomic DNA sequence in FASTA format and looks for
5
      antibiotic resistance mutations saved in its database.
6
  #
7
  #
8
  # Main Steps:
9
  #
10
  # Iteration through every target sequence in the pipeline's database:
11
      1. Reading reference mutations and checking whether sequence codes for a
  #
      protein.
12
  #
      2. Sequence extraction from genome FASTA file of gene that can contain
      antibiotics resistance mutations.
        - BLAST alignment of reference sequence against genome.
13
  #
        - Merging of BLAST hits from same FASTA sequence and same gene copy.
14 \#
        - Extension of BLAST hits to cover reference sequence fully.
15 \#
16
  #
        - Deletion of bad BLAST hits.
         - N padding of BLAST hits that don't cover reference sequence fully.
17 \#
18 \#
      3. Global alignment of nucleotide sequence (as well as translated sequence if
      it is a protein) against the reference sequence.
      4. Mutation calling
19
  #
20 \#
      5. Comparison with antibiotics resistance mutations in database.
21
  #
        - Report creation
  #
22
23 \#
24 # Dependencies:
25 \# 1. NCBI BLAST
26
  # 2. MAFFT
27
  # 3. EMBOSS
28
  #
29
  30
  # IMPORTS
31
  32
33 import sys
34
   import os.path
  import shutil
35
36
  import subprocess
37
   import argparse
  import random
38
39
   import datetime
40
   import errno
   import re
41
42
  import collections
43
44
# Main data flow
46
47
  ______
48
49
50
   def main():
```

```
51
      \# find directory of this script
52
       global pipeline dir
53
       pipeline_dir = os.path.dirname(os.path.realpath(___file__))
54
       # Parse arguments
       parse_parameters()
55
       # Initialise output/working directory
56
       create_dir(args.out)
57
58
59
       # Saving report path variables
       mutations report path = args.out + "/report.txt"
60
       mutations_not_covered_path = args.out + "/mutations_not_covered.txt"
targets_not_covered_path = args.out + "/targets_not_covered.txt"
61
62
63
64
       # Remove old output files in output directory
       remove_old_out_files([mutations_report_path, mutations_not_covered_path,
65
         targets_not_covered_path])
66
67
       # Read input file to variable
68
       global genome_contents
69
       with open(args.infile.name, "r") as myfile:
70
         genome_contents = myfile.read()
71
 72
       # Read 'mutations' and 'proteins' databases
73
       mutations_db = read_database("mutations")
       targets_db = read_database("targets")
74
 75
76
       # Check if species argument is supported. Exit if not.
77
       supported_species = get_supported_species(targets_db)
78
       if args.species not in supported_species:
         print ('ERROR: "' + args. species + '"' + ' was not recognised as a supported
79
         species. Supported species are: ')
         for species in supported_species:
80
81
           print(species)
82
         print()
83
         sys.exit()
84
85
       # Keep only targets for the correct species
86
       targets_db = get_targets_for_species(targets_db)
87
88
      # Initiate report lists
89
       mutation_report = []
 90
       mutation_coverage_report = []
91
       target_ID_coverage_report = []
92
93
       # Make BLAST database from genome file
94
       BLAST_database_path = make_BLAST_db(args.infile.name, args.out + "/tmp")
95
       L_print("Iterating through target sequences in targets database.")
96
97
       \# Iterate through each target ID to handle one sequence at a time
98
       for target_ID in targets_db:
         L_print("Working with target '{0}.".format(targets_db[target_ID][1]))
99
100
101
         # Saving path to reference sequence and working directory for current target ID
         refseq_path = pipeline_dir + "/reference_sequences/" + target_ID + ".fasta"
out_dir = args.out + "/tmp/" + target_ID
102
103
104
105
         # Check that the reference sequence file is available
         if check_file(refseq_path) is False:
106
         print ("ERROR: Path to reference sequence ({0}) could not be resolved. Check integrity of targets database. Shutting down.".format(refseq_path))
107
108
           sys.exit()
109
110
         # Initiating tmp directory for target ID
         create_dir(out_dir)
111
112
         113
         # READING REFERENCE MUTATIONS
114
115
116
         # Reading reference mutations
117
         ref_mutations = read_ref_mutations(target_ID, args.species, mutations_db,
```

```
targets_db)
118
        # Check if "target ID" sequence is a protein.
119
120
        if len(ref_mutations.prot) > 0:
121
          translation\_required = True
122
        else:
123
          translation_required = False
124
125
        # SEQUENCE EXTRACTION
126
127
128
        # BLAST alignment
129
        BLAST_alignment_path = BLAST_align(args.infile.name, refseq_path, out_dir,
        BLAST_database_path)
130
131
        # BLAST parsing
132
        match_dict, query, no_of_hits = parse_BLAST_alignment(BLAST_alignment_path,
        refseq_path)
133
134
        if no_of_hits > 0:
135
          # Checking for multiple hits from any one FASTA sequence and merging these if
         appropriate
136
          match_dict = merge_same_fseq_hits(match_dict, query.length)
137
          # Extending BLAST hits to cover reference sequence, and deleting bad hits.
138
          match_dict, no_of_hits = extend_BLAST_hits(match_dict, query.length)
139
140
141
          if no_of_hits > 0:
            # Finding overlap between BLAST hits
142
            nucl_overlaps = find_nucl_overlap(match_dict)
143
144
145
            # Saving reference sequence and (padded) hit sequences
146
            nucl_sequences = get_padded_BLAST_sequences(match_dict, query.name, query.
        sequence)
147
148
            ****
149
            # TRANSLATION AND SEQUENCE ALIGNMENT
150
151
            # Writing nucleotide sequences to file for MAFFT
152
            extracted_sequences_path = write_fasta(nucl_sequences, out_dir + "/
        extracted_sequences.fasta")
153
            \# MAFFT alignment and parsing for nucleotide sequences
154
155
            MAFFT\_nucl\_alignment\_path = MAFFT\_align(extracted\_sequences\_path, out\_dir + Contracted\_sequences\_path)
         "/MAFFT nucl alignment")
156
            MAFFT_aligned_nucl_sequences = parse_MAFFT_alignment(
        MAFFT_nucl_alignment_path)
157
158
            if translation_required is True:
159
              # Translation to amino acid sequence and parsing translation
              translated_path = translate(extracted_sequences_path)
160
161
              remove_translation_suffixes(translated_path)
162
              # MAFFT alignment and parsing for amino acid sequences
163
164
              MAFFT_prot_alignment_path = MAFFT_align(translated_path, out_dir + "/
        MAFFT_prot_alignment")
165
              MAFFT_aligned_prot_sequences = parse_MAFFT_alignment(
        MAFFT_prot_alignment_path)
166
167
              # Finding overlaps between the amino acid sequences
              prot_overlaps = find_protein_overlap(MAFFT_aligned_prot_sequences.ref,
168
        MAFFT\_aligned\_prot\_sequences.hits)
169
170
            171
            # MUTATION CALLING
172
173
            # Calling mutations
            nucl_mutations, bases_covered = call_mutations(MAFFT_aligned_nucl_sequences
174
        .ref, MAFFT_aligned_nucl_sequences.hits, ref_mutations, "nucl")
175
            if translation_required is True:
```

```
176
              prot_mutations, amino_acids_covered = call_mutations(
        MAFFT aligned prot sequences ref, MAFFT aligned prot sequences hits,
                       "prot")
        ref_mutations,
177
178
            # Compare called mutations to reference sequence mutations and append
        results to the mutation report
179
            mutation_report = append_mutation_report(mutation_report,
        MAFFT_aligned_nucl_sequences, nucl_overlaps, nucl_mutations, ref_mutations, "
        nucl")
180
            if translation required is True:
              mutation_report = append_mutation_report(mutation_report,
181
        MAFFT_aligned_prot_sequences, prot_overlaps, prot_mutations, ref_mutations, "
        prot")
182
183
        # Creating empty sets for append_coverage_report(), in case no hits were
        available
184
        if no_of_hits = 0:
185
          bases\_covered = set()
186
          amino_acids_covered = set()
187
188
        # MUTATION COMPARISON AND REPORT
189
190
191
        # Check which bases/amino acid weren't covered in the sequence and append to
        coverage reports.
192
        mutation_coverage_report , target_ID_coverage_report = append_coverage_report (
        mutation_coverage_report ,
193
                                               target_ID_coverage_report ,
194
                                               bases covered,
195
                                               ref_mutations,
196
                                               targets_db,
197
                                               target ID,
198
                                               query.length,
199
                                                nucl")
200
        if translation_required is True:
201
          mutation_coverage_report, target_ID_coverage_report = append_coverage_report (
        mutation_coverage_report,
202
                                                 target_ID_coverage_report ,
203
                                                 amino_acids_covered,
204
                                                 ref_mutations,
205
                                                 targets_db,
                                                 target_ID,
206
207
                                                 int(query.length / 3),
208
                                                  "prot")
209
        \#\ {\rm Remove\ target\ ID\ tmp\ directory\ when\ it\ is\ no\ longer\ needed.\ Not\ necessary\,,}
210
        but keeps the tmp directory from growing in size unnecessarily.
211
        if args.keep_tmp_files is False:
212
          shutil.rmtree(out_dir)
213
214
        L_print("Finished with target '{0}'.".format(targets_db[target_ID][1]), 2)
215
216
      # Remove the entire tmp directory
217
      if args.keep_tmp_files is False:
218
        shutil.rmtree(args.out + "/tmp")
219
220
      L_print("Finished with all target sequences.")
221
222
      # Write mutation report to file
      header = "Target name\tInvolved antibiotics\tMutation\tNucl/prot\tResistance\
223
        tFASTA sequence(s)\tMutation prevalence\tSequence coverage
224
      write_report_file(mutation_report, header, mutations_report_path)
225
226
      # Write mutation coverage report to file
227
      228
      write_report_file(mutation_coverage_report, header, mutations_not_covered_path)
229
      # Write target ID coverage report to file
230
231
      header = "Target name\tInvolved antibiotics"
      write_report_file(target_ID_coverage_report, header, targets_not_covered_path)
232
```

```
233
234
      235
236
      # Exit
      L_print("Done. Exiting")
237
238
      sys.exit()
239
240
241
    242
    # Basic input and parameter handling
243
    ****
244
245
    \# Prints an error message along with how to call —help for the script
246
247
    def error_msg(msg):
248
      print(msg)
      print("Try \"python {0} ---help\" for more information.".format(sys.argv[0]))
249
250
251
252
    \#\ {\rm Defining\ arguments}\,, and processing where direct access to the global args.
        INPUT PARAMETER is not sufficient
253
    # (e.g. assigning printlevels for --verbose and --quiet)
254
    def parse_parameters():
255
      # Defining arguments
256
      parser = argparse.ArgumentParser()
      parser.add_argument("-i", "--infile", required="True", type=argparse.FileType('r'
257
        ), help="Input genome file to be analysed - fasta format.")
       parser.add_argument("-o", "--out", help="Output and working directory. Choose a
258
        non-existing directory to remove all risk of conflicting file names.")
      parser.add_argument("-s", "--species", required="True", help="Specify the species
of the input genome - abbreviated name (e.g. E. coli).")
parser.add_argument("--list_species", action="store_true", help="Lists the
259
260
      species that are supported, and exits.")
parser.add_argument("-1", "-logfile", nargs='?', const="standard_path", help="
261
        Directs all output (verbose) to a log file.")
       parser.add_argument("-v", "--verbose", action="store_true", help="Prints all
262
        messages.")
       parser.add_argument("-q", "--quiet", action="store_true", help="Suppresses all
263
        output.")
264
       parser.add_argument("--ext_program_output", action="store_true", default=False,
        help="Includes external program output in the pipeline's output.")
      parser.add_argument("--keep_tmp_files", action="store_true", default=False, help=
    "Does not delete any temporary files.")
265
       parser.add_argument("-BLAST_perc_identity", type=float, help="Sets percent
266
        identity cutoff for BLAST (0-100). Default is 95 %.")
267
       parser.add_argument("--report_all_coverage", action="store_true", default=False,
        , regardless of whether it corresponds to a possible resistance mutation or not .")
268
       global args # Declaring global inside parse_parameters, so that both other
269
        functions AND this function can access it.
270
      args = parser.parse_args()
271
272
      # Processing arguments
273
       if args.list_species:
274
        targets_db = read_database("proteins")
        supported_species = get_supported_species(targets_db)
275
         print("\nSupported species:")
276
277
         for species in supported_species:
278
          print(species)
279
         print()
280
         sys.exit()
281
282
       if args.BLAST_perc_identity is not None:
283
         if args.BLAST_perc_identity <= 0 or args.BLAST_perc_identity >= 100:
          error_msg("ERROR: Specified BLAST percent identity cutoff is not a float
284
        between 0 and 100.")
285
          sys.exit()
286
       else:
```

```
287
         args.BLAST_perc_identity = 95
288
289
       if args.quiet:
290
         args.printlevel = 0
291
       elif args.verbose:
292
         args.printlevel = 2
293
       else:
294
         args.printlevel = 1
295
296
       if args.out is not None:
297
         if args.out.endswith("/"):
298
           args.out = args.out.rstrip("/")
299
       else:
300
         # Creates a directory named with today's date and some random numbers if no
         directory was specified, in order not to overwrite anything
         date = datetime.date.today().isoformat()
301
302
         randbits = random.getrandbits(32)
303
         \operatorname{args.out} = [0]_{1}_{2}".format("pipeline_out", date, randbits)
304
305
       if args.logfile == "standard_path":
306
         args.logfile = args.out + "/log.txt"
307
308
       if args.logfile is not None:
309
         init_logfile()
310
311
312 # Custom print function that takes levels 0, 1, and 2 to designate in which
         operation modes
313 # it will be printed or suppressed (normal, quiet, verbose).
314 # Writes to logfile if one is specified.
315
     def L_print(msg, level=1):
316
       if args.logfile is None:
317
         if level <= args.printlevel:
           print (msg)
318
319
       else:
320
         of = open(args.logfile, "a")
321
         of.write (msg + " \setminus n")
322
         of.close()
323
324
    \# Executes shell commands, and uses L_print to either display the output, suppress
325
         it, or write it to the logfile.
    \# Note that suppressOut=True can be set if output should be excluded from L_print (
326
     e.g. for MAFFT, whose output is all alignments)
def L_execute(cmd, level=1, header="", suppress_out=False):
327
       proc = subprocess.Popen(cmd, stdout=subprocess.PIPE, stderr=subprocess.PIPE,
328
         shell=True)
329
       (out, err) = proc.communicate()
330
       if args.ext\_program\_output is True:
         # Print header if it's defined - used for external programs with printed output if header != "":
331
332
           L_{print}(" n" + "-" * 80 + " n" + header + " n" + "-" * 80, level)
333
334
         # For shell commands whose output needs to be saved, but not L printed, the
         following conditional helps
335
         if suppress_out is False:
           # Empty output should not show up as empty lines in the log/output
if out.decode('ascii') != "":
336
337
338
              L_print(out.decode('ascii'), level)
         if err.decode('ascii') != "":
L_print(err.decode('ascii'), level)
339
340
          if header != "":
341
           L_{print}("-" * 80 + " \ n", level)
342
343
       return out
344
345
    # Removes files in file_list, as well as tmp directory, in output directory.
346
     def remove_old_out_files(file_list):
347
348
       tmp_dir = args.out + "/tmp"
349
       if os.path.exists(tmp dir) and os.path.isdir(tmp dir):
350
         shutil.rmtree(tmp_dir)
```

```
351
      for file in file_list:
352
         silent remove (args.out + "/" + file)
353
354
355
    # Initialises logfile. If file exists: exit if not an old logfile, overwrite if it
        is
356
     def init_logfile():
357
      # Creates log file directory if it doesn't exist
       if args.logfile.endswith("/"):
358
         error_msg("Specified logfile path ended with a '/'. Please specify the desired
359
         path of the logfile, not its parent directory.")
360
         sys.exit()
361
       log_dir = args.logfile[:args.logfile.rindex("/")]
362
       create_dir(log_dir, suppress_print=True)
363
364
      # Checks if file exists, and stops script if it is something else than an old log
          file
365
       logfile_header = "#" * 80 + "\n
                                              Logfile of ARM-find\ln^{+} + \# * 80 + \ln^{+}
       if os. path. is file (args. log file):
366
367
         with open(args.logfile, "r") as myfile:
368
           contents = myfile.read()
         if re.match(logfile_header, contents) is None:
369
370
           error_msg("Logfile error: the specified file already exists, but is not an
         old logfile. Exiting.")
371
          sys.exit()
       if os.path.isdir(args.logfile):
372
373
         error_msg("Logfile error: the specified path is a directory. Please specify the
          desired path of the logfile. Exiting.")
374
         sys.exit()
375
      # Removes old logfile and writes header to new file
376
       silent_remove(args.logfile) # If args.logfile is the path to an old logfile,
377
         removes it
378
       of = open(args.logfile, "a")
379
       of.write(logfile_header)
380
       of.close()
381
382
383
    # Creates directories recursively
384
    def create_dir(path, suppress_print=False):
       parts = path.rstrip("/").split("/")
385
386
387
       for i in range(len(parts)):
         tmppath = "/" . join(parts[:(i + 1)])
# This check makes sure paths starting with "/" don't cause crashes.
388
389
         if tmppath !{=} "":
390
391
           if os.path.exists(tmppath) is False:
392
             if suppress_print is False:
               L_print("Creating {0}".format(tmppath), 2)
393
394
             os.makedirs(tmppath)
395
396
397
    # File removal
398
    def silent_remove(filename):
399
      \mathbf{try}:
400
         os.remove(filename)
       except OSError as e:
401
         if e.errno != errno.ENOENT: # errno.ENOENT = no such file or directory
402
403
           raise # re-raise exception if a different error occured
404
405
    \# Checks for a file
406
407
    def check_file(filename):
408
      if os.path.exists(filename) is False or os.path.getsize(filename) == 0:
409
         L_print("Cannot find {0}.".format(filename), 2)
410
         return False
       else:
411
         return True
412
413
414
```

```
415
    416
    # BLAST functions
417
    418
    # Makes a BLAST database from the genome file
419
    def make_BLAST_db(infile , out_dir):
420
      # Making a BLAST database for alignment
421
      L_print("Making a BLAST database from {0}".format(infile), 2)
BLAST_database_path = "{0}/BLAST_database".format(out_dir)
422
423
      L execute("makeblastdb -dbtype nucl -in {0} -out {1}".format(infile,
424
        BLAST_database_path), level=2, header="makeblastdb output:
425
       return BLAST_database_path
426
427
    # Does a BLAST alignment of a reference sequence (ref_seq_path) against a genome
428
         file (infile), and generates output files.
    def BLAST_align(infile, ref_seq_path, out_dir, BLAST_database_path):
429
430
      L_print("Making a Megablast alignment of {0} against {1}".format(ref_seq_path,
        infile), 2)
431
      \# Megablast alignment. BLAST archive chosen as output format to enable later
      conversion with blast_formatter
BLAST_result_path = "{0}/megablast_alignment.asn".format(out_dir)
432
      L_execute("blastn -db {0} -query {1} -outfmt 11 -out {2} -perc_identity {3}".
format(BLAST_database_path, ref_seq_path, BLAST_result_path, str(args.
433
        BLAST_perc_identity)), level=2)
      L_print("BLAST alignment complete\n", 2)
434
435
      return BLAST_result_path
436
437
438
    \# Takes a BLAST alignment in output format 11 (BLAST archive) and parses it.
439
    #
      Also takes the path to the reference sequence used as the query in the BLAST
        alignment,
440
    # in order to be able to return that sequence as part of a list for MAFFT alignment
441 \# Parsing BLAST in tabular format (5) rather than the current xml (11) has some
        advantages, but since sequence names with multiple adjacent spaces are
        truncated badly
442 # in tabular format, we would have to create a temporary genome file with sequence
        names without spaces, and then replace the names there with the correct names
        in the report.
    def parse_BLAST_alignment(BLAST_alignment, ref_seq_path):
443
      L_print("Parsing BLAST alignment file: {0}".format(BLAST_alignment), 2)
444
      # Conversion to BLAST XML (for parsing) and pairwise (for readability) output
445
        formats
446
      dir path = os.path.dirname(BLAST alignment)
      name = os.path.basename(BLAST_alignment)
447
448
       if name.endswith('.asn'):
        name = name[:-4]
449
      xml_file_path = "{0}/{1}.xml".format(dir_path, name)
450
451
      readable_file_path = "{0}/{1}_pairwise".format(dir_path, name)
452
453
      \# The tabular format is probably the best for computationally low-demand parsing,
         except for the fact that all names with spaces are truncated in fasta files,
        which is unacceptable.
      L_print("Converting BLAST archive format to xml ({0})".format(xml_file_path), 2)
454
      L_execute("blast_formatter -archive {0} -outfmt 5 -out {1}".format(
455
        BLAST_alignment, xml_file_path), level=2)
456
      # There is no point in creating human-readable files if all tmp folders are
457
        deleted.
458
       if args.keep tmp files is True:
        L_{print}("Converting BLAST archive format to human-readable format ({0})".format
459
        (readable_file_path), 2)
460
        L_execute("blast_formatter -archive {0} -outfmt 0 -out {1}".format(
        BLAST_alignment, readable_file_path), 2)
461
      L_print("Retrieving alignment information from file", 2)
462
463
      # Read xml file contents to variable
      with open(xml file path, "r") as myfile:
464
        xml_contents = myfile.read()
465
```

```
466
      \# Reading the contents of the reference sequence file
       with open(ref seq path, "r") as myfile:
467
468
         ref_seq_contents = myfile.read()
469
470
      # Number of alignments made
471
       no_of_hits = xml_contents.count("<Hsp_num>")
472
473
      \# Name of query sequence in fasta file (after >).
       query_name = re.search(r"<BlastOutput_query-def>(.*?)</BlastOutput_query-def>",
474
         xml contents).group(1)
475
476
      \# Replace any XML codes for characters with the actual characters.
477
      query_name = replace_XML_codes(query_name)
478
      \# Sequence name correction (lines starting with > in fasta files). BLAST XML
479
        output format removes all but
480
      \# one space whenever there are two or more consecutive spaces in sequence names,
        but only for the query sequence.
       query_name = correct_BLAST_name_spaces(query_name, ref_seq_contents)
481
482
       query_seq = get_fseq(query_name, ref_seq_contents).upper()
483
484
       query\_length = len(query\_seq)
485
486
       Query = collections.namedtuple('Query', 'name, sequence, length')
487
       query = Query(query_name, query_seq, query_length)
488
489
      \# The function has to be exited if no hits are found, with the same kind of data
         structure returned as if it had been successful.
490
       # Further handling of this case in main()
       if no_of_hits == 0:
491
492
         L print("WARNING! No BLAST hits were found for {0}. Mutation calling not
         possible.".format(query_name))
493
         return None, query, no_of_hits
494
495
      \# Create dictionary where each key will be a property of the BLAST alignment (e.g
          sequence length).
496
      \# The values will be lists, where the indices correspond to BLAST hits. For
         example, with 2 hits, one would get
497
        the sequence length of the second hit by accessing the list belonging to the
         key for sequence length, at the second index.
498
       match_dict = \{\}
499
      # Save all relevant attributes of the hits to the dictionary.
match_dict['db_strand'] = re.findall(r"<Hsp_hit-frame>(.*?)</Hsp_hit-frame>",
500
501
        xml_contents)
502
       match_dict['query_strand'] = re.findall(r"<Hsp_query-frame>(.*?)</Hsp_query-frame
        >", xml_contents)
503
       match_dict['db_hit_seq'] = re.findall(r"<Hsp_hseq>(*?)</Hsp_hseq>", xml_contents
         )
504
       match_dict['query_hit_seq'] = re.findall(r"<Hsp_qseq>(.*?)</Hsp_qseq>",
        xml_contents)
505
       match_dict['db_hit_from'] = re.findall(r"<Hsp_hit-from>(.*?)</Hsp_hit-from>",
        xml_contents)
506
       match_dict['db_hit_to'] = re.findall(r"<Hsp_hit-to>(.*?)</Hsp_hit-to>",
         xml_contents)
507
       match_dict['query_hit_from'] = re findall(r"<Hsp_query-from>(*?)</Hsp_query-from</pre>
        >", xml_contents)
       match_dict['query_hit_to'] = re.findall(r"<Hsp_query-to>(.*?)</Hsp_query-to>",
508
         xml_contents)
509
      \# As several hits can be in one FASTA sequence, and all are within one <Hit></Hit
510
         >, each FASTA sequence name has
511
      # to be counted the right amount of times.
512
       matching_fseqs = re.findall(r"<Hit>(.*?)</Hit>", xml_contents, re.DOTALL)
       match\_dict['fseq'] = []
513
       for fseq in matching_fseqs:
514
         fseq_name = re.search(r"<Hit_def>(.*?)</Hit_def>", fseq).group(1)
515
516
         hits_in_fseq = fseq.count("<Hsp>")
         for i in range(hits_in_fseq):
    match_dict['fseq'].append(replace_XML_codes(fseq_name))
517
518
```

```
519
520
       # Type conversions
       match_dict['db_hit_from'] = [int(x) for x in match_dict['db_hit_from']]
521
       match_dict['db_hit_to'] = [int(x) for x in match_dict['db_hit_to']]
522
       match_dict['query_hit_from'] = [int(x) for x in match_dict['query_hit_from']]
match_dict['query_hit_to'] = [int(x) for x in match_dict['query_hit_to']]
523
524
525
526
       # Create match_dict entries for query/db_hit_from/to + extension, and initialise
         lists in them.
       match dict ['query hit from extended'] = [None] * no of hits
527
       match_dict['query_hit_to_extended'] = [None] * no_of_hits
match_dict['db_hit_from_extended'] = [None] * no_of_hits
528
529
       match_dict['db_hit_to_extended'] = [None] * no_of_hits
530
531
532
       # Order hits by query_hit_from to get later reporting in a sensible order from
         the perspective of the reference sequence
533
       sorted_indices = sorted(range(no_of_hits), key=lambda k: match_dict['
         query_hit_from '][k])
534
       for key in match dict:
535
         match_dict[key] = [match_dict[key][i] for i in sorted_indices]
536
537
       return match_dict, query, no_of_hits
538
539
540
    # Sequence name correction (lines starting with > in FASTA files). BLAST XML output
         format
    \# removes all but one space whenever there are two or more consecutive spaces in
541
         sequence names.
542
    def correct_BLAST_name_spaces(sequence_name, fasta_file_contents):
       pattern = "'
543
544
       for char in sequence_name:
545
         if char == " ":
           pattern += r"\s+"
546
547
         else:
548
           pattern += re.escape(char)
549
       sequence_name = re.search(pattern, fasta_file_contents).group(0)
550
       return sequence name
551
552
553
    # Replaces any XML codes for characters with the actual characters.
554
    def replace_XML_codes(string):
       codes = { "" ": `, "& ": "&", "' ": ", "< ": "<", "&gt; ": ">" }
555
       for code, char in codes.items():
556
557
         string = string.replace(code, char)
558
       return string
559
560
    \# Merges non-overlapping BLAST hits that are on the same FASTA sequence, and are
561
         not different copies of a gene.
562
     def merge_same_fseq_hits(match_dict, query_length):
      \# Check for more than one match per FASTA sequence and LPrint if so.
563
       # Note that this assumes no two FASTA sequences will have the same name in one
564
         genome file.
565
       multiple_hits_in_fseq = False
566
       no_of_hits = len(match_dict['fseq'])
567
       if no_of_hits > 1:
         L_print("Checking for multiple hits from any one genomic FASTA sequence.", 2)
568
         multimatch_fseq = {}
569
         multimatch_fseq_hit_indices = {}
570
571
         for i in range(no_of_hits):
           fseq = match_dict['fseq'][i]
572
           count = match_dict['fseq'].count(fseq)
573
574
           if count > 1:
575
             multimatch_fseq[fseq] = count
576
             multiple_hits_in_fseq = True
              if fseq not in multimatch_fseq_hit_indices:
577
578
               multimatch_fseq_hit_indices[fseq] = [i]
             else:
579
580
               multimatch fseq hit indices [fseq].append(i)
581
```

```
582
         if len(multimatch_fseq) > 0:
583
           L print ("WARNING: multiple matches from one genomic FASTA sequence found.")
584
         else:
585
          L_print("No genomic FASTA sequence gave rise to more than one hit sequence.",
          2)
586
         for fseq , count in multimatch_fseq.items():
587
           L_print("WARNING: genomic FASTA sequence named {0} gave rise to {1}
588
         alignments.".format(fseq, count))
589
590
      # Looking for hits on the same FASTA sequence that should be merged.
591
      \# This solution assumes no overlap and ascending order of query_hit_from. This
        should be true, as we've sorted above,
592
      \# and BLAST should not find multiple matches that overlap in one FASTA sequence.
593
594
      # Initialise list for remembering how many bases to add if hits are merged
       match_dict['merged_hits_extra_bases'] = [0] * no_of_hits
595
      # Initialise list for remembering how many bases to add if hits are NOT merged.
match_dict['no_merge_extra_bases'] = [0] * no_of_hits
596
597
598
      # Initialise list for remembering which hits are involved in merges.
599
       match_dict['merged_hit'] = [False] * no_of_hits
600
601
       if multiple_hits_in_fseq is True:
602
         hits_to_merge = []
603
604
         for fseq , indices in multimatch_fseq_hit_indices.items():
           L_print("Checking if multiple hits from genomic FASTA sequence '\{0\}' can be
605
         merged.".format(fseq), 2)
606
          \# Indices must be sorted by strand, and then by db_hit_from (ascending for
607
         strand 1, descending for strand -1).
608
          # This is in order to make sure that hits that are next to each other in the
         genome are next to each other here.
          # If there are multiple copies of a gene on one FASTA sequence, only using
609
         query_hit_from sorting could cause problems.
           indices_by_strand = \{1: [], -1: []\}
610
611
           for i in indices:
             if match_dict['db_strand'][i] == "1":
612
613
               indices_by_strand [1].append(i)
614
             else:
615
               indices_by_strand[-1].append(i)
616
           indices\_by\_strand[1].sort(key=lambda~k:~match\_dict['db\_hit\_from'][k])
617
           indices_by_strand[-1].sort(key=lambda k: match_dict['db_hit_from'][k],
618
         reverse=True)
619
620
           indices = indices_by_strand[1] + indices_by_strand[-1]
621
622
           \# i can't be the last index, since we're comparing with i + 1.
623
           for i in range(len(indices) - 1):
             \# Distance between the matching query sequence parts of the two hits
624
625
             query_hit_from = match_dict['query_hit_from'][indices[i]], match_dict['
         query_hit_from '][indices[i + 1]]
             query_hit_to = match_dict['query_hit_to'][indices[i]], match_dict['
626
         query_hit_to'][indices[i + 1]]
627
628
             query_distance = (max(query_hit_from) - min(query_hit_to) - 1)
629
             # Distance between the two hits on the FASTA sequence
630
             db_hit_from = match_dict['db_hit_from'][indices[i]], match_dict['
631
         db_hit_from '][indices[i + 1]]
             db_hit_to = match_dict['db_hit_to'][indices[i]], match_dict['db_hit_to'][
632
         indices[i + 1]]
633
             \# Hit distance calculation depends on which strand the hits are on.
             if match_dict['db_strand'][indices[i]] == "1":
634
635
               hit_distance = max(db_hit_from) - min(db_hit_to) - 1
636
             else:
637
               hit_distance = max(db_hit_to) - min(db_hit_from) - 1
638
             # For hit overlap (on FASTA sequence) check
639
```

640	overlap = range(max(match_dict['db_hit_from'][indices[i]], match_dict['
	db_hit_from'][indices[i + 1]]),
641	min(match_dict['db_hit_to'] indices[i]], match_dict['db_hit_to']
0.40	$\operatorname{indices}\left[1 + 1\right] + 1))$
642	
643	# If the query distance is 0, or close to 0, we have an insertion, and the bits should be marged
644	# If the query distance isn't 0, but the difference between the hit and
	query distances is small (some room for indels),
645	# the we have a mismatching region, and the hits should still be merged.
646	# Also check that the matches are on the same strand – though problems with this are unlikely.
647	if (
648	$(query_distance < 100 \text{ or } abs(hit_distance - query_distance) < 100)$ and
649	# Make sure hits are on the same strand
650	<pre>match_dict['db_strand'][indices[i]] == match_dict['db_strand'][indices[</pre>
	i + 1]] and
651	# Don't merge hits that are from different copies of a gene
652	hit_distance <= query_length and
653	# Checks to make sure that the hits don't overlap in the genome file,
	as this could cause problems. I don't think they ever will, but I'm not 100 $\%$ sure.
654	len(overlap) = 0
655	
656	
657	# If there are multiple copies of a gene on the same FASTA sequence, it
658	is possible to erroneously try to merge the end of one copy with the start # of another. This check prevents that by making sure that the order of
000	the hits in the genome matches the order in the query
659	bit from distances – query bit from $[1] - query bit from [0] db bit from$
005	[1] _ db hit from [0]
660	if (
661	\mathcal{H} If db strand is 1 shock that the sign of the differences between
001	$\#$ if ub_stand is i, there into its sign of the differences between
669	$(n_{\text{the statis}})$ of the fits is the same (of that effective differences is o) and $(n_{\text{the statis}})$ and $(n_{\text{the statis}})$
002	match digt (db, strond) [indigas [ii]] - "1") or
669	match_dict[db_strand][indices[i]] — 1) of
005	$\#$ II dD_strand is -1 check that the sign of the differences between
CC A	the starts of the first different (of that either/both differences is o)
004	$(\min(\operatorname{ntc_irom_uistances}) \le 0 \le \max(\operatorname{ntc_irom_uistances})$ and
0.0 F	$matcn_act[ab_strand][indices [1]] = -1)$
665):
666	
667	# Get FASIA sequence sequence to check whether interval between hits
	contains too many N's for merging
668	<pre>tseq_sequence = get_tseq(tseq, genome_contents)</pre>
669	# Get range of interval between hits, regardless of whether the strand
C T O	
670	space_trom, space_to = sorted $(db_htt_trom + db_htt_to)[1:-1]$
671	# Accept merge if amount of N's in interval between hits is below 30 %.
672	<pre>it tseq_sequence[space_trom - 1:space_to].upper().count("N") < (</pre>
c 7 9	space_to $+ 1 - \text{space}_{\text{trom}} * 0.3$:
673	# Append adjacent hits that are to be merged, and remember that they
	re involved in merging.
674	hits_to_merge.append({'indices': [indices[i], indices[i + 1]], '
	insertion_size': hit_distance - query_distance})
675	for index in $[indices[i], indices[i + 1]]$:
676	match_dict['merged_hit'][index] = True
677	
678	# If two adjacent hits are not merged due to N content problems, but
	there is an indel between them, then that needs to be compensated for during
679	$\#$ hit extension for the hit with lower query_hit_from.
680	else:
681	<pre>match_dict['no_merge_extra_bases'][min(indices[i], indices[i + 1])] =</pre>
	hit_distance - query_distance
682	
683	# If there are hits to merge, compare adjacent sets of hits to merge with each
	other. If all hits should
684	# be merged, add the last hit to the first set, and add the additional hit
	distance. The second set is then disregarded.
685	if hits_to_merge != []:
```
686
           # Since info on each merge is collected in one list, other lists of the same
         merge are disregarded.
687
           skip_index = set()
688
           # indices in hits_to_merge that will end up containing information on merging
            merge_info_indices = set()
689
            for i in range(len(hits_to_merge)):
690
              if i not in skip_index:
691
692
                merge_info_indices.add(i)
                if len(hits_to_merge) > 1:
693
                  for j in range(i + 1, len(hits_to_merge)):
    if hits_to_merge[i]['indices'][-1] == hits_to_merge[j]['indices'][0]:
        hits_to_merge[i]['indices'].append(hits_to_merge[j]['indices'][1])
694
695
696
                       hits_to_merge[i]['insertion_size'] += hits_to_merge[j]['
697
         insertion_size']
698
                       skip_index.add(j)
699
700
           # Initiate list with indices to delete across all merges
701
           all\_indices\_to\_delete = []
702
           # Going through sets of hits to merge
703
            for i in merge_info_indices:
              # Saving indices of hits that should be deleted for current merge
704
705
              indices_to_delete = hits_to_merge[i]['indices'][:-1]
706
              \# Saving the extra bases necessary for the hit that will be extended to
         cover the others
              match_dict['merged_hits_extra_bases'][hits_to_merge[i]['indices'][-1]] =
707
         hits_to_merge[i]['insertion_size']
708
709
              L_print("{0} hits (hit numbers: {1}) were merged into one.".format(len(
         indices_to_delete) + 1, ", ".join([str(hits_to_merge[i]['indices'][j] + 1) for
         j in range(len(hits_to_merge[i]['indices']))]), 2)
710
711
              \# The hits that are to be deleted can contain indels, which will change how
          long the remaining sequence should be extended - this is taken care of here.
712
              for j in indices_to_delete:
                indel_extra = abs(match_dict['db_hit_to'][j] - match_dict['db_hit_from'][
713
         j]) - abs(match_dict['query_hit_to'][j] - match_dict['query_hit_from'][j])
match_dict['merged_hits_extra_bases'][hits_to_merge[i]['indices'][-1]] +=
714
          indel extra
715
716
              all_indices_to_delete += indices_to_delete
717
           \# Delete hits in reverse order, so that the things that should be deleted don
718
          't change index in the loop
719
           for j in all_indices_to_delete[::-1]:
              no_of_hits -= 1
720
721
              for key in match dict:
722
                del match_dict[key][j]
723
         else:
724
           L_print("No hits from genomic FASTA sequence {0} could be merged.".format(
         fseq), 2)
725
726
       return match dict
727
728
729
    # Extends BLAST hits to cover as much of query sequence as possible
730
     def extend_BLAST_hits(match_dict, query_length):
       # Looping through the BLAST hits and adding extra bases to sequences as needed to
731
           fill as much of the query sequence as possible
732
       L_print("Checking if BLAST hits need to be extended to cover query sequence.", 2)
733
       \# Save all covered position INDICES (first position is 0) for each FASTA sequence
734
         , so that extension can avoid covering a base twice, thereby avoiding false
         overlap.
735
       fseq\_coverage = \{\}
736
       for i in range(len(match_dict['fseq'])):
737
         if match_dict['fseq'][i] not in fseq_coverage.keys():
           fseq_coverage[match_dict['fseq'][i]] = set()
738
         # We have to sort hits in order not to get 0 length range on one strand.
sorted_db_hits = sorted ([match_dict['db_hit_from'][i], match_dict['db_hit_to'][
739
740
```

XXI

```
i ]])
         for j in range(sorted_db_hits[0], sorted_db_hits[1] + 1):
    fseq_coverage[match_dict['fseq'][i]].add(j - 1)
741
742
743
744
      no_of_hits = len(match_dict['fseq'])
745
       for i in range(no_of_hits):
        L_print("Working with hit \#\{0\} found on genomic FASTA sequence named: \{1\}".
746
        format(i + 1, match_dict['fseq'][i]), 2)
        \# Save number of extra bases needed from FASTA sequence (upstream and
747
        downstream) to cover as much of the query sequence as possible.
        \# Upstream and downstream are in the perspective of the query sequence. "
748
        merged_hits_extra_bases" are added upstream in case
        \# several hits from one FASTA sequence are to be merged (the distance between
749
        these hits must be taken into account).
        # 'no_merge_extra_bases' are added downstream in case two adjacent hits weren't
750
         merged due to too many N's, to compensate for indels.
         upstream_extra = match_dict['query_hit_from'][i] - 1 + match_dict['
751
        merged_hits_extra_bases '][i]
752
        downstream_extra = query_length - match_dict['query_hit_to'][i] + match_dict['
        no_merge_extra_bases'][i]
753
        \# Initialise extension counters to keep track of how much a match sequence has
        been extended.
754
        \# Necessary in order to keep track of what query index corresponds to the ends
        of a match.
755
        added\_upstream = 0
756
        added_downstream = 0
757
758
        # If the FASTA sequence doesn't span the entire query:
759
         if upstream extra > 0 or downstream extra > 0:
760
761
          # Save DNA sequence of the matched FASTA sequence
762
          fseq_seq = get_fseq(match_dict['fseq'][i], genome_contents)
763
764
           if upstream\_extra > 0:
765
             \# If BLAST hit is on the strand presented in the FASTA sequence:
             if match_dict['db_strand'][i] == "1":
766
767
              # Save indices of needed upstream bases in order of proximity to hit
         sequence
768
              indices = range(match_dict['db_hit_from'][i] - 2, match_dict['db_hit_from
         [i] - upstream\_extra - 2, -1)
769
               for j in indices:
                # If index is in range of FASTA sequence string, add base at that index
770
         to beginning of hit sequence
771
                 if j \ge 0:
                  \# Check that the base isn't already part of another hit on the FASTA
772
        773
774
                     break
775
                   else:
776
                     match_dict['db_hit_seq'][i] = fseq_seq[j] + match_dict['db_hit_seq'
        ][i]
777
                     added_upstream += 1
778
                     fseq_coverage[match_dict['fseq'][i]].add(j)
779
780
             \# Else (if BLAST hit is on the complementary strand to the FASTA sequence):
781
             else:
782
              # Save indices of needed upstream bases in order of proximity to hit
        sequence.
783
               indices = range(match_dict['db_hit_from'][i], match_dict['db_hit_from'][i
        ] + upstream_extra)
784
               for j in indices:
                \# If index is in range of FASTA sequence string, add complement of base
785
         at that index to beginning of hit sequence.
786
                \# Not REVERSE complement, since adding bases one by one in the order in
         indices already takes care of the order.
787
                 if j < len(fseq_seq):
                   \# Check that the base isn't already part of another hit on the FASTA
788
        sequence to avoid false overlap. Stop loop if it is.
                   if j in fseq_coverage[match_dict['fseq'][i]]:
789
790
                     break
```

791 792	<pre>else: match_dict['db_hit_seq'][i] = complement(fseq_seq[j]) + match_dict['db_hit_seq'][i]</pre>
793 794 795	added_upstream += 1 fseq_coverage[match_dict['fseq'][i]].add(j)
796 797 798 799	<pre>if downstream_extra > 0: # If BLAST hit is on the strand presented in the FASTA sequence: if match_dict['db_strand'][i] == "1": # Save indices of needed downstream bases in order of proximity to hit</pre>
800	<pre>sequence indices = range(match_dict['db_hit_to'][i], match_dict['db_hit_to'][i] +</pre>
801 802	<pre>downstream_extra) for j in indices: # If index is in range of FASTA sequence string, add base at that index to end of hit operators</pre>
803 804	<pre>if j < len(fseq_seq): # Check that the base isn't already part of another hit on the FASTA sequence to avoid false overlap. Stop loop if it is.</pre>
805 806 807	<pre>if j in fseq_coverage[match_dict['fseq'][i]]:</pre>
808 809	<pre>match_dict['db_hit_seq'][i] += fseq_seq[j] added_downstream += 1</pre>
810 811	<pre>fseq_coverage[match_dict['fseq'][i]].add(j)</pre>
812 813	# Else (if BLAST hit is on the complementary strand to the FASTA sequence): else:
814	<pre>indices = range(match_dict['db_hit_to'][i] - 2, match_dict['db_hit_to'][i]] - downstream extra - 2, -1)</pre>
815 816	for j in indices: # If index is in range of FASTA sequence string, add complement of base
817	at that index to end of hit sequence. # Not REVERSE complement, since adding bases one by one in the order in indices already takes care of the order.
818 819	$\begin{array}{l} \mbox{if $j>=0$:} \\ \mbox{ $\#$ Check that the base isn't already part of another hit on the FASTA} \\ \mbox{sequence to avoid false overlap. Stop loop if it is.} \end{array}$
820 821 822	<pre>if j in fseq_coverage[match_dict['fseq'][i]]:</pre>
823 804	<pre>match_dict['db_hit_seq'][i] += complement(fseq_seq[j])</pre>
824 825 826	added_downstream += 1 fseq_coverage[match_dict['fseq'][i]].add(j)
827	$message = ($ "Added {0} bases upstream of hit sequence, and {1} bases
828	"Note that upstream and downstream refer to reverse complement if match was found on complement strand.")
829 830	$L_{print}(message.format(added_upstream, added_downstream), 2)$
831	# Save the range of the reference sequence covered by the hit sequence after extension
832	# 'merged_hits_extra_bases' only applies upstream, since only the highest query hit from hit is kept during merging.
833	<pre>match_dict['query_hit_from_extended'][i] = match_dict['query_hit_from'][i] - added_upstream + match_dict['merged_hits_extra_bases'][i]</pre>
834	# 'no_merge_extra_bases' only applies downstream. If hits aren't merged, extra (or less) extension is handled downstream, to avoid frameshift for a whole hit in case of unknown regions (many N's).
835	<pre>match_dict['query_hit_to_extended'][i] = match_dict['query_hit_to'][i] + added_downstream - match_dict['no_merge_extra_bases'][i]</pre>
836 837	# Save the range of the genome sequence covered by the hit sequence after extension
838 839	<pre>if match_dict['db_strand'][i] == "1": match_dict['db_hit_from_extended'][i] = match_dict['db_hit_from'][i] - added upstream</pre>
840	<pre>match_dict['db_hit_to_extended'][i] = match_dict['db_hit_to'][i] + added_downstream</pre>

```
841
         else:
            match dict['db hit from extended'][i] = match dict['db hit from'][i] +
842
         added_upstream
           match_dict['db_hit_to_extended'][i] = match_dict['db_hit_to'][i] -
843
         added downstream
844
       # Remove lone hits that have to be extended too far.
845
       match_dict, no_of_hits = remove_lone_short_hits(match_dict, query_length,
846
         fseq_coverage)
847
848
       return match_dict , no_of_hits
849
850
    # Remove lone hits that have to be extended too far, unless part of a merge.
def remove_lone_short_hits(match_dict, query_length, fseq_coverage):
851
852
       L_print("Checking if any BLAST hits need to be deleted."
853
                                                                      , 2)
854
855
       no_of_hits = len(match_dict['fseq'])
856
       indices_to_delete = []
857
858
       for i in range(no_of_hits):
         L_{print} ("Working with hit #{0}, found on genomic FASTA sequence named: {1}".
859
         format(i + 1, match_dict['fseq'][i]), 2)
         # Don't delete merged hits.
if match_dict['merged_hit'][i] is True:
860
861
862
           continue
         # Don't delete if hit extends up to another hit. Min/max solves strand issues.
Note that 'db_hit_to_extended' contains a base number (first number is 1),
863
         while fseq coverage contains indices (first index is 0).
864
         \#\ \mathrm{OBS!} This is problematic if the short incorrect hit is close to the actual
         gene. It might extend up to a base that's already added and be safe from
         deletion.
865
         if min(match_dict['db_hit_from_extended'][i], match_dict['db_hit_to_extended'][
         i]) - 2 in fseq_coverage[match_dict['fseq'][i]]:
866
           continue
         if max(match_dict['db_hit_from_extended'][i], match_dict['db_hit_to_extended'][
867
         i]) in fseq_coverage[match_dict['fseq'][i]]:
868
           continue
         # Delete hits that did not meet the criteria above, and that were extended too
869
         far.
         extension = abs(match_dict['query_hit_from_extended'][i] - match_dict['
870
         query_hit_from '][i]) + abs(match_dict['query_hit_to_extended'][i] - match_dict[
         'query_hit_to'][i])
if extension / query_length > 0.2:
871
872
           L_print("Hit #{0} was deleted due to long extension (likely erroneous hit).".
         format(i + 1), 2)
873
            indices_to_delete.append(i)
874
875
       \# Delete hits in reverse order, so that the things that should be deleted don't
         change index in the loop
876
       for i in indices_to_delete [:: -1]:
         no_of_hits -= 1
877
878
         for key in match dict:
879
           del match_dict[key][i]
880
881
       if len(indices_to_delete) == 0:
         L\_print("No hits were deleted.", 2)
882
883
884
       return match_dict, no_of_hits
885
886
887
     # Return overlapping ranges in nucleotide sequence
888
     def find_nucl_overlap(match_dict):
889
       # No overlap until it's found
890
       overlap\_exists = False
891
       no_of_hits = len(match_dict['fseq'])
892
893
       # Initialise list of lists for overlap info
894
       overlaps = [[] for i in range(no_of_hits)]
895
```

```
897
      # Check for several hits from one FASTA sequence, and hit sequence overlap, if
        more than one BLAST hit was found.
      if no_of_hits > 1:
898
899
        L_print("Checking for overlap between hit sequences.", 2)
        # Check for overlap between hit sequences
900
901
        for i in range(no_of_hits):
902
          for j in range(no_of_hits):
903
            # Disregard overlap with self
             if j != i:
904
905
              \#\ {\rm Range}\ {\rm object}\ {\rm of}\ {\rm overlap}\ {\rm sequence.}\ {\rm Length}\ 0 if no overlap
              overlap = range(max(match_dict['query_hit_from_extended'][i], match_dict[
906
         'query_hit_from_extended'][j]),
                      min(match_dict['query_hit_to_extended'][i], match_dict['
907
        query_hit_to_extended '][j]) + 1)
              # If there is overlap:
908
               if len(overlap) > 0:
909
910
                overlap_exists = True
                \# Save info on overlap for hit sequence i: overlapping sequence index (
911
        j\,)\,, and query sequence numbers for the overlap (from, to)
912
                overlaps[i].append([j, min(overlap), max(overlap)])
913
        if overlap_exists is True:
   L_print("WARNING: Overlap between sequences found.")
914
915
          for i in range(no_of_hits):
916
             if overlaps[i] != []:
917
918
               for j in range(len(overlaps[i])):
919
                 if overlaps [i][j][0] > i:
                  L_print("Genomic FASTA sequences '{0}' and '{1}' overlap in the
920
        reference sequence region of {2}-{3}".format(match_dict['fseq'][i], match_dict[
         fseq' = 0 overlaps [i][j][0], overlaps [i][j][1], overlaps [i][j][2], 2)
921
        else:
922
          L_print("No overlap between sequences was found.", 2)
923
924
      return overlaps
925
926
    \# Return reference and hit sequences, padding the latter if necessary
927
928
    def get_padded_BLAST_sequences(match_dict, query_name, query_seq):
929
        Adding sequences to a list of lists for passing to write_fasta. Starting here
        with the reference sequence
930
      # and later appending padded_hit_seqs.
931
      seq_list = [[query_name, query_seq]]
932
933
      query length = len(query seq)
934
935
      no of hits = len(match dict['fseq'])
936
      \# Add appropriate n padding to the hits and save them to seq_list for return
937
        statement
      n_seq = "n" * query_length
938
939
      for i in range(no_of_hits):
940
        # Add padding to the hit sequence.
        padded_hit_seq = n_seq[:match_dict['query_hit_from_extended'][i] - 1] +
941
        match_dict['db_hit_seq'][i] + n_seq[match_dict['query_hit_to_extended'][i]:]
942
        # Gap hyphens need to be removed.
        padded_hit_seq = padded_hit_seq.replace("-", "")
943
944
        # Append padded hit sequences to seq_list for passing to write_fasta
        "seq_list.append([match_dict['fseq'][i], padded_hit_seq.upper()])
945
946
      L print ("BLAST parsing of '\{0\}'' complete \n\n". format (query name), 2)
947
948
949
      return seq_list
950
951
    ****
952
953
    # Alignment
954
    ****
955
    # Does a MAFFT alignment of sequences in infile and returns a string with the
956
```

896

```
output path
def MAFFT_align(infile, out_path):
   L_print("Making a MAFFT alignment of sequences in {0}".format(infile), 2)
957
958
959
       \# Without — any symbol, MAFFT removes * (stop codons) before alignment. Note that
          this results in case-sensitivity.
        L_execute("ginsi --anysymbol {0} > {1}".format(infile, out_path), level=2, header
960
         ="MAFFT output:")
961
        L_print("MAFFT alignment complete", 2)
962
        return out_path
963
964
965
     # Parse MAFFT alignment
     def parse MAFFT alignment (MAFFT alignment path):
966
        L_print("Parsing MAFFT alignment file: {0}".format(MAFFT_alignment_path), 2)
967
968
969
        with open(MAFFT_alignment_path) as myfile:
970
          MAFFT_alignment = myfile.read()
971
972
       \# Go through MAFFT alignment and save sequences and their names as lists in the
          hit_seqs list.
        \mathrm{hits} = []
973
        lines = MAFFT_alignment.splitlines()
974
975
        for line in lines:
976
          if line.startswith(">"):
            hits.append([line[1:], ""])
977
978
          else:
979
            hits [-1][1] += "".join(line.split())
980
981
       # Save the reference sequence list in hit_seqs to ref_seq, and delete it from
          hit_seqs
982
        ref = hits[0]
        del hits [0]
983
984
        AlignedSequences = collections.namedtuple('AlignedSequences', 'ref, hits')
985
        aligned_sequences = AlignedSequences(ref, hits)
986
987
988
        return aligned sequences
989
990
991
     # Returns positions of overlaps in terms of reference base numbers for all protein
          hit sequences.
     def find_protein_overlap(ref, hits):
992
993
        no_of_hits = len(hits)
994
995
        ref seq = ref[1]
996
        hit_seqs = [hits[i][1] for i in range(no_of_hits)]
997
998
        prot_start_index = [None] * no_of_hits
999
        prot_end_index = [None] * no_of_hits
1000
1001
        overlaps = [[] for i in range(no_of_hits)]
1002
1003
        if no of hits > 1:
1004
          \# Save index of first and last amino acid that isn't an X or a - for each hit,
          and then change the indices to
1005
          # reference sequence index by subtracting any gap hyphens in the reference
          sequence up to the relevant point.
1006
          for i in range(no_of_hits):
             \begin{array}{l} prot\_start\_index[i] = re.search(r"^[X-]*.", hit\_seqs[i]).end() - 1 \\ prot\_start\_index[i] = ref\_seq.count("-", 0, prot\_start\_index[i] + 1) \end{array} \end{array} 
1007
1008
1009
            1010
1011
1012
1013
          \# Saving overlap with other hit sequences for each hit sequence.
1014
          for i in range(no_of_hits):
1015
            for j in range(no_of_hits):
1016
              # No overlap with self
1017
              if j != i:
                # Range object of overlap sequence. Length 0 if no overlap
1018
```

```
1019
               overlap = range(max(prot_start_index[i], prot_start_index[j]), min(
         prot_end_index[i], prot_end_index[j]) + 1)
               if len(overlap) > 0:
1020
1021
                 # Reporting amino acid numbers, not indices, so +1.
1022
                 overlaps[i].append([j, min(overlap) + 1, max(overlap) + 1])
1023
1024
       return overlaps
1025
1026
1027
     1028
     \# Mutation calling , comparison , and report
1029
     1030
1031
     # Call mutations in nucleotide or amino sequence (sequence_type = "nucl"/"prot").
1032
     \# Also checks which bases in the reference sequence are covered, that is not only
         by N/X.
     def call_mutations(ref, hits, ref_mutations, sequence_type):
    if sequence_type == "nucl":
1033
1034
         message = "DNA"
1035
1036
         ref_mutations = ref_mutations.nucl
1037
       elif sequence_type == "prot":
         message = "protein"
1038
1039
         ref_mutations = ref_mutations.prot
1040
       L_print("Calling mutations in {0} sequence".format(message), 2)
1041
1042
       \# Get positions of potential resistance mutations. Necessary for cases where the
         reference sequence has a resistance mutation,
1043
       # since only called mutations are checked for resistance.
1044
       ref_mutation_positions = set()
       if ref_mutations != []:
1045
1046
         for ref_mutation in ref_mutations:
           ref_mutation_positions.add(int(ref_mutation[0]))
1047
1048
1049
       no_of_hits = len(hits)
1050
1051
       \# Initialise lists with mutation dicts
       subs = [collections.OrderedDict() for i in range(no_of_hits)]
ins = [collections.OrderedDict() for i in range(no_of_hits)]
1052
1053
1054
       dels = [collections.OrderedDict() for i in range(no_of_hits)]
1055
1056
       # Initialise set to save all bases/amino acids that are covered (not N/X),
1057
       # to facilitate checking if a resistance mutation is covered.
1058
       ref_numbers_covered = set()
1059
1060
       # Save sequence info to separate variables, in order to avoid index mayhem.
1061
       ref\_seq = ref[1]
1062
       hit_seqs = [hits[i][1] for i in range(no_of_hits)]
1063
       # Saving mutations
1064
       for hit in range(no_of_hits):
1065
          Initialise base number counter for reference sequence - for handling gaps.
         The first base will be numbered 1.
1066
         ref_number = 0
1067
         for i in range(len(ref_seq)):
          # To report mutation positions in relation to reference sequence, gaps have
1068
         to be handled
1069
           if ref_seq[i] != "-":
1070
             ref_number += 1
1071
           \# If base is the same in hit and ref, remember that this base is covered
1072
           if ref_seq[i] == hit_seqs[hit][i]:
1073
             ref_numbers_covered.add(ref_number)
1074
             # Check if ref number is a resistance-associated position. If so, add like
         normal sub.
1075
             if ref_number in ref_mutation_positions:
1076
               subs[hit][ref_number] = [ref_seq[i].upper(), hit_seqs[hit][i].upper(), i]
1077
           # Else (if bases differ), check how
1078
           else:
             \#\ {\rm If} hit sequence base is a gap, save reference base number, reference base
1079
          and index in the hit
             if hit_seqs[hit][i] == "-":
1080
               dels[hit][ref_number] = [ref_seq[i], i]
1081
```

```
1082
                 ref_numbers_covered.add(ref_number)
          \# If reference sequence base is a gap, save reference base number before gap, inserted hit sequence base, and index in the hit.
1083
1084
              \# If there is an extension of a gap, append the extra base instead. We only
           need to save the index of the first gap (not extra for extensions).
               elif ref_seq[i] == "-":
1085
                 if ref_number in ins[hit]:
1086
                   ins [ hit ] [ ref_number ] [ 1 ] += hit_seqs [ hit ] [ i ]
1087
1088
                 else:
1089
                   ins[hit][ref_number] = [ref_seq[i - 1], hit_seqs[hit][i], i]
1090
              \#\ {\rm If} normal substitution, save reference base number, reference sequence
          base, hit sequence base, and index in the hit.
              \# Only normal substitution , and unknown base (N) remain , so check for "not
1091
          Ν".
1092
               elif ((sequence_type == "nucl" and hit_seqs[hit][i].upper() != "N") or
                   (sequence_type == "prot" and hit_seqs[hit][i].upper() != "X")):
1093
                 subs[hit][ref_number] = [ref_seq[i].upper(), hit_seqs[hit][i].upper(), i]
1094
1095
                 ref_numbers_covered.add(ref_number)
1096
1097
        CalledMutations = collections.namedtuple('CalledMutations', 'subs, ins, dels')
        called_mutations = CalledMutations(subs, ins, dels)
1098
1099
        return called_mutations, ref_numbers_covered
1100
1101
     # Appends called mutations to report, while making a comparison with known
1102
          resistance mutations.
1103
      {\tt def append\_mutation\_report(mutation\_report, MAFFT\_alignment, overlaps, }
        hit_mutations, ref_mutations, seq_type):
if seq_type == "nucl":
1104
          message = "DNA"
1105
        elif seq_type == "prot":
1106
          message = "protein"
1107
        L\_print("Comparing called mutations in {0} sequence to resistance mutations, and
1108
          adding to report.".format(message), 2)
1109
1110
        antibiotics = ", ".join(get_involved_antibiotics(ref_mutations))
1111
        \# Set variables for nucleotid/amino acid so the same name can be used.
1112
        if seq_type == "nucl":
1113
1114
          ref\_resistance\_mutations = ref\_mutations.nucl
1115
          unknown = "N"
        elif seq_type == "prot":
1116
1117
          ref\_resistance\_mutations = ref\_mutations.prot
1118
          unknown = "X"
1119
        ref = MAFFT_alignment.ref
1120
1121
        hits = MAFFT alignment.hits
1122
1123
        no_of_hits = len(hits)
1124
        report_list = []
1125
1126
        # Loop through hit sequences
1127
        for i in range(no_of_hits):
1128
          # Loop through substitutions for the given hit sequence
1129
          for mutation_type in ['subs', 'dels', 'ins']:
1130
1131
             for ref_number, mutation in getattr(hit_mutations, mutation_type)[i].items():
1132
              \# No resistance unless found
               resistance = ""
1133
1134
               # Loop through resistance mutations
1135
               for ref_mutation in ref_resistance_mutations:
                 \# Set resistance to the the correct \operatorname{antibiotic}\,(\,s\,) if the mutation matches
1136
           any on record for resistance
1137
                 if ((mutation_type = 'subs' and int(ref_mutation[0]) = ref_number and
          ref_mutation [2] == mutation [1]) or

(mutation_type == 'dels' and int(ref_mutation [0]) == ref_number and

ref_mutation [2] == "-")):
1138
                   if resistance != "":
1139
                     resistance += ",
1140
                   resistance += ref_mutation[3]
1141
```

```
1143
              # If reference base/aa is the same as mutated base/aa, and that mutation is
           not associated with resistance, skip adding it to report
1144
              \# mutation [0] == mutation [1] occurs because all resistance-associated
          positions are called as mutations, in case the reference sequence has a
          resistance mutation.
1145
              # This check removes all cases like R45R, except if that confers resistance
1146
              if mutation [0] == mutation [1] and resistance == "":
1147
                 continue
1148
1149
              \# Initiate variable for counting overlapping FASTA sequences at this
          mutation, and how many of them have the same mutation.
1150
              overlapping_fseqs = 0
              same_mutation = 0
1151
1152
              \# Initiate variable containing the FASTA sequences a mutation was found on
               fseqs = "'{0}''.format(hits[i][0])
1153
1154
               for overlap in overlaps [i]:
                \# Have to check that there are overlaps. If there aren't, overlaps will
1155
          contain *no_of_hits* empty lists
1156
                 if len(overlap) > 0:
                  \# Check if base/amino acid at ref_number overlaps with any other FASTA
1157
          sequence
1158
                   if overlap [1] <= ref_number and overlap [2] >= ref_number:
1159
                     overlapping_fseqs += 1
                     \# Appends the FASTA sequence name which overlaps the current one
1160
1161
                     fseqs += ", " + " '{0}'.format(hits[overlap[0]][0])
                     # Check if the substituted/deleted base/amino acid in FASTA sequence
1162
          is the same as the base/amino acid in the
                     \# overlapping one. (overlap[0] is the index of an overlapping FASTA
1163
          sequence in hits, 1 refers to the sequence (not name)
1164
                     \# \ {\rm and} \ {\rm mutation} \left[ -1 \right] is the index of the base/amino acid in that
          sequence). If same mutation, increment variable
           if ((mutation_type == 'subs' and hits[overlap[0]][1][mutation[-1]] == mutation[1]) or
1165
                       (mutation\_type = 'dels' and hits[overlap[0]][1][mutation[-1]] = "
1166
          -")):
1167
                       same_mutation += 1
1168
                     # If mutation_type is 'ins':
                     elif mutation_type == 'ins':
1169
                       # Same check as above, but taking potentially longer-than-one
1170
          sequence into account.
                       if hits [overlap [0]][1][mutation [-1]:mutation [-1] + len(mutation [1])
1171
           + 1] == mutation [1]:
1172
                         same mutation += 1
1173
1174
              # Adjusting report depending on type of mutation
1175
               if mutation_type == 'subs':
                mutation\_text = mutation[0] + str(ref\_number) + mutation[1]
1176
1177
               elif mutation_type == 'dels
                mutation\_text = mutation[0] + str(ref\_number) + "-"
1178
1179
               elif mutation_type == 'ins'
1180
                mutation text = mutation [0] + str(ref number) + mutation [0] + mutation [1]
1181
1182
              # Only write mutation prevalence column value if there are overlapping
          FASTA sequences
1183
               if overlapping_fseqs > 0:
                mutation\_prevalence = (same\_mutation + 1) / (overlapping\_fseqs + 1)
1184
1185
               else:
                 mutation_prevalence = ""
1186
1187
              # Info on how well the hit covers the reference sequence
if hits[i][1][0] != unknown and hits[i][1][-1] != unknown:
1188
1189
1190
                coverage = "Full"
               elif hits[i][1][0] != unknown:
    coverage = "Partial w/ start"
1191
1192
1193
               else:
                coverage = "Partial w/o start"
1194
1195
              \# Append an item to internal report list with information on the mutation.
1196
```

1142

```
1197
                         report_list_entry = [ref[0], antibiotics, mutation_text, seq_type,
                 resistance, fseqs, mutation_prevalence, coverage]
report_list.append([report_list_entry, ref_number])
1198
1199
1200
             # Sort internal report list by ref_number, instead of by ref_number for subs, ins
                 , and dels separately
1201
              sorted_indices = sorted(range(len(report_list)), key=lambda k: report_list[k][1])
1202
             report_list = [report_list[i] for i in sorted_indices]
1203
             # Append internal report list (nucl/prot) to class instance
1204
1205
              for item in report_list:
1206
                 mutation_report.append(item[0])
1207
1208
             return mutation_report
1209
1210
1211
         # Appends resistance mutations that are not covered by the hit sequences to
                 mutation coverage report,
1212
         \# and appends target sequences where no mutations were covered to target coverage
                 report.
         {\tt def append\_coverage\_report(mutation\_coverage\_report, target\_ID\_coverage\_report, target\_ID\_coverag
1213
                 positions_covered , ref_mutations , targets_db , target_ID , seq_length , seq_type):
1214
             L_print("Checking if any resistance mutations weren't covered in hit sequences,
                 and adding to coverage report", 2)
1215
1216
              antibiotics = ", ".join(get_involved_antibiotics(ref_mutations))
1217
1218
             \# Set variables for nucleotide/amino acid so the same name can be used.
              if seq type == "nucl":
1219
1220
                 ref_resistance_mutations = ref_mutations.nucl
1221
                 any\_coverage = False
              elif seq_type == "prot":
1222
1223
                 ref_resistance_mutations = ref_mutations.prot
1224
1225
             \# Save reference mutations list to dict, with the mutation position as the key.
1226
             ref_mutations_dict = {}
1227
             for mutation in ref_resistance_mutations:
1228
                 ref_mutations_dict[int(mutation[0])] = mutation[1:]
1229
1230
             for i in range (1, \text{ seq\_length} + 1):
1231
                 if i not in positions_covered:
                    \# If the current position isn't covered and corresponds to a reference
1232
                 mutation, add item to coverage report, including resistance type of the
                 mutation.
1233
                     if i in ref mutations dict.keys():
                        mutation_coverage_report.append([targets_db[target_ID][1], antibiotics, i,
1234
                 seq_type, ref_mutations_dict[i][-1]])
1235
                    \# If the current position isn't covered and doesn't correspond to a reference
1236
                  mutation, add item to coverage report, but only if the -- report_all_coverage
                 argument is given.
1237
                     elif args.report_all_coverage is True:
1238
                        mutation_coverage_report.append([targets_db[target_ID][1], antibiotics, i,
                 seq_type , ""])
1239
1240
                 elif i in positions_covered and seq_type == "nucl":
1241
                    any\_coverage = True
1242
             # If no base is covered, add target ID to target_ID_not_covered report.
if seq_type == "nucl" and any_coverage is False:
1243
1244
1245
                 target_ID_coverage_report.append([targets_db[target_ID][1], antibiotics])
1246
1247
              return mutation_coverage_report, target_ID_coverage_report
1248
1249
         \# Write list of lists as tab-separated values on separate lines in file.
1250
          def write_report_file(report, header, file_path):
1251
             L_print("Writing report: {0}".format(file_path), 2)
1252
1253
              file = open(file path, 'w')
              file.write(header + "\n")
1254
```

```
1255
1256
       for row in report:
1257
         file.write("t".join([str(x) for x in row]) + "n")
1258
1259
       file.close()
1260
1261
1262
     1263
     # Database handling
1264
     1265
     # Read database file ("mutations" or "targets") and return dict
1266
     \# with keys being all unique values at index 0 in each line, and the corresponding
1267
1268
     \# values being lists with the remaining values of each line in list format.
1269
     def read_database(database_type):
       L_print("Reading {0} database.".format(database_type), 2)
1270
1271
       # Set file path according to database type
1272
       if database_type == "mutations":
       file_path = pipeline_dir + "/mutation_databases/mutations"
elif_database_type == "targets":
1273
1274
         file_path = pipeline_dir + "/mutation_databases/targets"
1275
1276
1277
       if check_file(file_path) is False:
1278
         print("ERROR: Could not find {0} database. Check if correct file is present
         with path '{1}'. Exiting.".format(database_type, file_path))
1279
         sys.exit()
1280
1281
       with open(file_path, 'r') as myfile:
1282
         contents = myfile.read()
1283
1284
       lines = contents.splitlines()
1285
       db dict = collections.OrderedDict()
1286
1287
       for i in range(1, len(lines)):
         lines [i] = lines [i]. split ("\t")
1288
1289
         \# Since there is only one entry for each target ID, each key in db_dict will
1290
         # contain one list without sublists.
         if database_type == "targets":
1291
1292
           db_dict[lines[i][0]] = lines[i][1:]
         # Since there almost certainly is more than one entry for each antibiotic, each # key in db_dict will contain a list of lists.
1293
1294
1295
         else:
           if lines[i][0] in db_dict:
db_dict[lines[i][0]].append(lines[i][1:])
1296
1297
1298
           else:
             db_dict[lines[i][0]] = [lines[i][1:]]
1299
1300
1301
       return db_dict
1302
1303
     \# Takes the full targets database dict and returns a dict with only the targets
1304
         corresponding to the correct species.
1305
     def get targets for species (targets db):
1306
       filtered_targets_db = collections.OrderedDict()
1307
       for target_ID, target_data in targets_db.items():
1308
         if target_data [0] == args.species:
1309
           filtered_targets_db[target_ID] = target_data
1310
       return filtered_targets_db
1311
1312
1313
     # Searches proteins database and returns species supported by the pipeline
     def get_supported_species(targets_db):
1314
1315
       supported\_species = set()
1316
1317
       for target_ID, data in targets_db.items():
1318
         if data [0] not in supported_species:
           supported_species.add(data[0])
1319
1320
1321
       return sorted (supported species)
1322
```

```
1323
1324
     # Return reference sequence mutations (both nucleotide and protein) corresponding
         to
1325
     # a certain target ID, and species.
1326
     {\tt def \ read\_ref\_mutations(target\_ID, \ species, \ mutations\_db, \ targets\_db):}
       L_print("Reading resistance mutations.", 2)
# Save list corresponding to the correct antibiotic from the mutation database
1327
1328
1329
       ref_mutations = mutations_db[target_ID]
1330
1331
       # Initialise lists for saving nucleotid/protein mutations
       nucl_mutations = []
1332
1333
       prot_mutations = []
1334
1335
       \# Save mutations corresponding to the correct species and protein
1336
       for i in range(len(ref_mutations)):
          if targets_db[target_ID][0] = species:
1337
1338
            # Separate nucleotide and protein mutations
1339
            if ref_mutations[i][1] == "nucl":
             nucl_mutations.append(ref_mutations[i][2:] + [ref_mutations[i][0]])
1340
1341
            else:
1342
             prot_mutations.append(ref_mutations[i][2:] + [ref_mutations[i][0]])
1343
1344
       RefMutations = collections.namedtuple('RefMutations', 'nucl, prot')
1345
       ref_mutations = RefMutations(nucl_mutations, prot_mutations)
1346
1347
       return ref_mutations
1348
1349
     \# Save the antibiotics classes that are involved in the resistance mutation search
1350
         for the current target ID.
1351
     def get_involved_antibiotics(ref_mutations):
       antibiotics = set()
1352
1353
       for ref_mutation in ref_mutations.nucl + ref_mutations.prot:
1354
          if ref_mutation[3] not in antibiotics: # The conditional is not really
         necessary here, but can improve speed
           antibiotics.add(ref_mutation[3])
1355
1356
       return antibiotics
1357
1358
1359
     1360
     \# General bioinformatics functions
     1361
1362
     # Translates DNA sequences in infile. Table 11 is for bacterial translation (Use
1363
         transeq -help for other organisms)
     def translate(infile, outfile=None, table=11):
1364
1365
       if outfile is None:
1366
         outfile = infile + "_translated"
       \# -snucleotide1 means the input file contains nucleotide sequences,
1367
1368
       L_print("Translating '{0}'' and saving output to '{1}'".format(infile, outfile),
         2)
       L\_execute("transeq -sequence \{0\} -snucleotide1 -table \{1\} -outseq \{2\}".format(
1369
         infile, table, outfile), level=2, header="transeq output:")
1370
       return outfile
1371
1372
     \# Removes reading frame suffix added to sequence names by EMBOSS's transeq
1373
     def remove_translation_suffixes(infile):
1374
       with open(infile, 'r') as myfile:
1375
1376
         translation = myfile.read()
1377
1378
       \# Go through translation and save sequences and their names (including >) as
         lists in the sequences list.
1379
       sequences = []
1380
       lines = translation.splitlines()
1381
        for line in lines:
          if line.startswith(">"):
1382
           \#\ {\rm Transeq}\ {\rm modifies}\ {\rm the}\ {\rm names}\ {\rm in}\ {\rm fasta}\ {\rm files}\ {\rm with}\ {\rm a}\ {\rm suffix}\ {\rm to}\ {\rm indicate}
1383
         translation frame.
           # For a frame one translation, "_1" is added after the first space, or at the
1384
```

```
end of the name
            \# if there are no spaces. The following code removes that suffix. if " " in line:
1385
1386
              corrected_name = re.sub(r"(^.*?)(_1)()(.*)", r"\1\3\4", line)
1387
1388
            else:
              corrected_name = re.sub(r"(^.*)(_1$)", r"\1", line)
1389
1390
1391
            sequences.append([corrected_name, ""])
1392
          else:
            sequences [-1][1] += "".join(line.split())
1393
1394
1395
        # Overwrite old file.
        with open(infile, 'w') as myfile:
1396
          for i in range (len (sequences)):
1397
            myfile.write(sequences[i][0] + "\n")
myfile.write(sequences[i][1] + "\n")
1398
1399
1400
1401
     \# Complement, including ambiguous base codes (except those that are their own
1402
          complement)
      def complement(sequence):
1403
        trantab = str.maketrans("ATCGKMRYBVHDatcgkmrybvhd", "TAGCMKYRVBDHtagcmkyrvbdh")
1404
1405
        sequence = sequence.translate(trantab)
1406
        return sequence
1407
1408
     \#~{\rm Gets} the DNA sequence between ">fseq" and the next ">" or end of string (
1409
          whichever is shortest)
1410
      {\tt def get\_fseq(fseq, genome):}
        pattern = r">" + re.escape(fseq) + r"(.*?)(>|$)"
1411
1412
        match = re.search(pattern, genome, re.DOTALL).group(1)
        match = "".join(match.split())
1413
1414
        return match
1415
1416
     \# Writes a fasta file from a list of lists , the latter containing fasta names (
1417
         after ">") and sequences
      def write_fasta(fasta_list, outfile):
1418
1419
        of = open(outfile, 'w')
1420
        for i in range(len(fasta_list)):
          of.write(">" + fasta_list[i][0] + "\n" + fasta_list[i][1] + "\n")
1421
1422
        of.close()
1423
        return outfile
1424
1425
1426
     \# Runs the main() function
1427
      if ____name____
                        main
1428
       main()
```