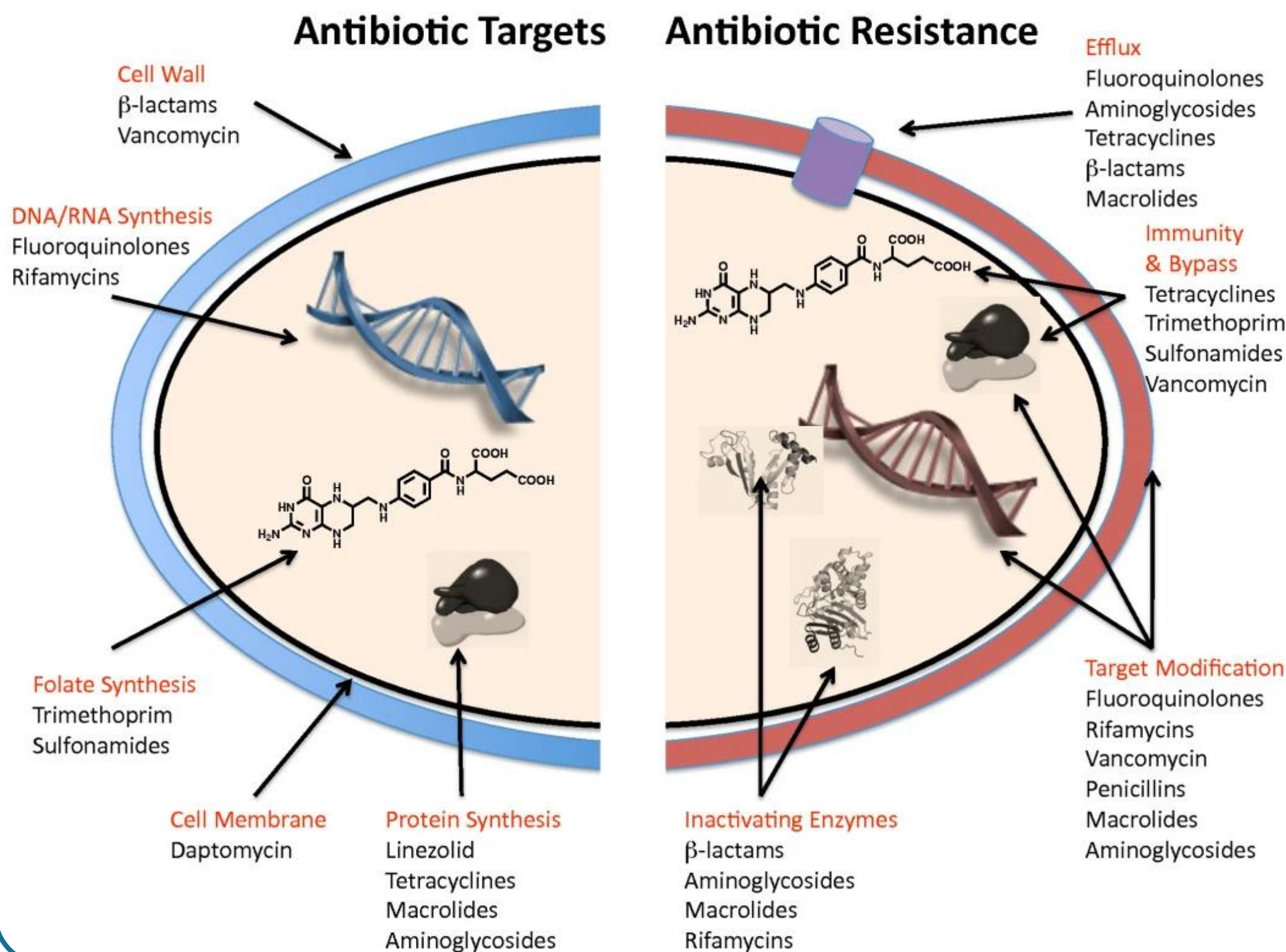


# ARSENAL: Antimicrobial ReSistance prEdiction by a mAchine Learning method

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## Background

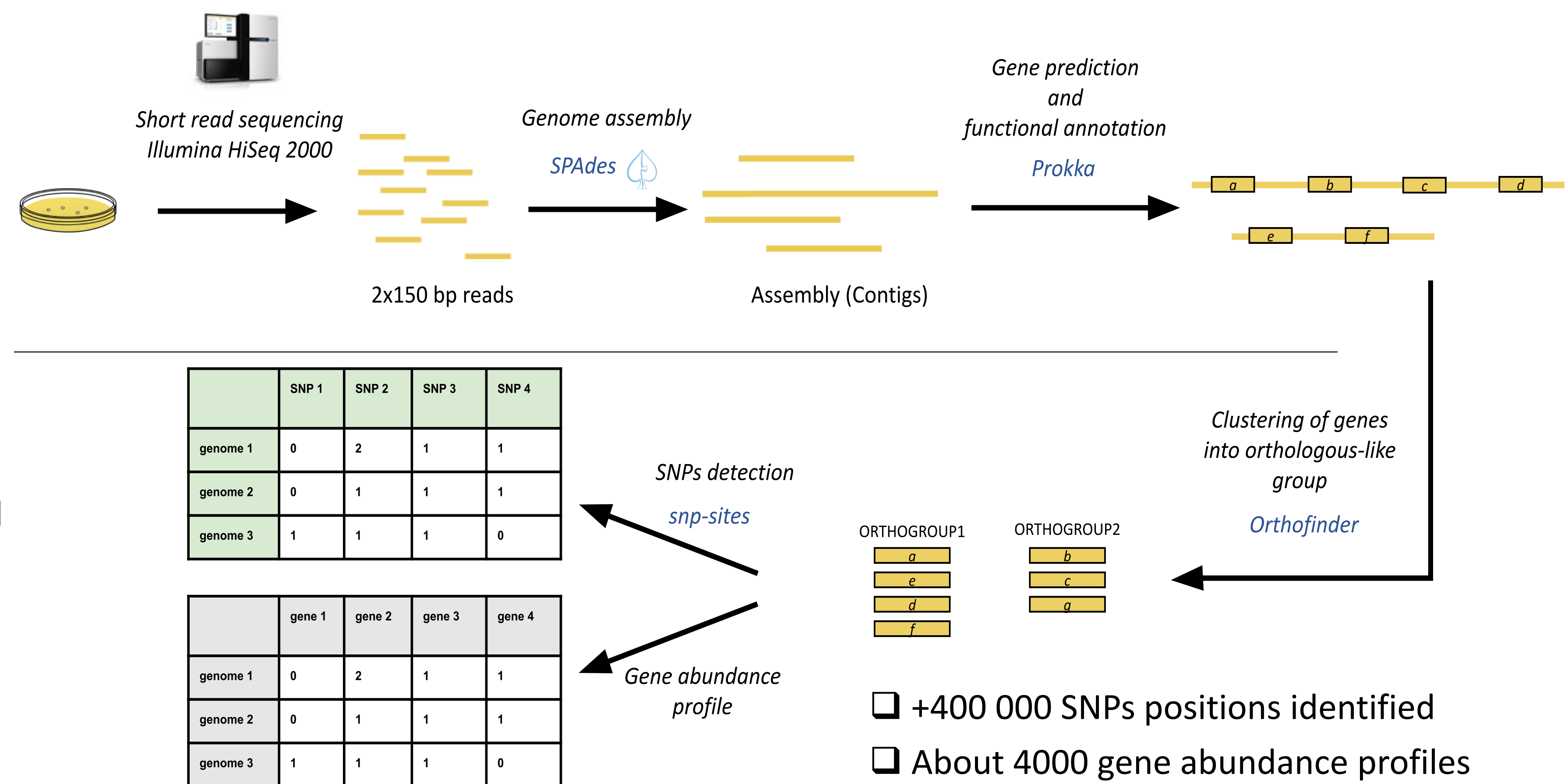


- Antimicrobial resistance (AMR) is a growing health threat responsible for an estimated 700 000 deaths per year and this number is projected to reach 10 million by 2050.
- Appropriate antibiotic therapy improves patient healing outcomes and is a key factor in preventing the emergence of antibiotic resistance.
- Antibiotic susceptibility testing (AST) from bacterial culture is the current clinical practice for assessing drug resistance by the determination of the minimum inhibitory concentration (MIC) corresponding to the lowest concentration of a specific antibiotic that inhibits bacterial growth. This method, fastidious and long (between 24h and 72h), is only applicable to cultivable bacteria, which excludes analysis of the emergence and spread of antimicrobial resistance in diverse and complex microbial communities with large fractions of currently uncultured bacteria.
- Our method allows:
  - ✓ Fast screening of antibiotics to determine the most effective antibiotic and the right dose against a specific bacterial infection
  - ✓ Highlight new antibiotic-resistance genes and biomarkers

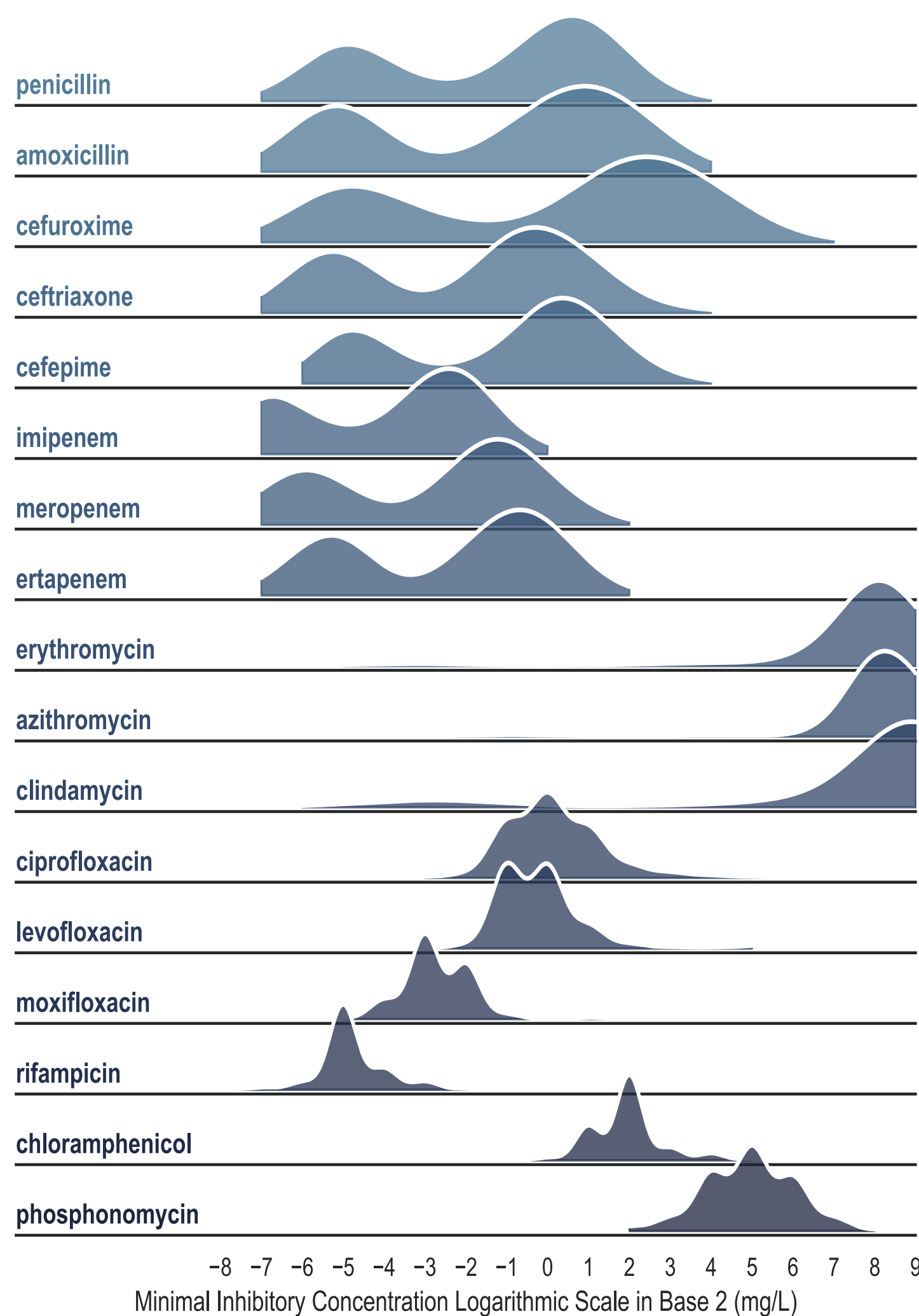
Resistance mechanisms are mainly acquired by genome modifications, e.g., mutations and horizontal gene transfers.

## Data Generation

- 1312 *Streptococcus pneumoniae* newly sequenced genomes of strains isolated from patients
  - ☐ Different sequence types (ST) and 64 distinct serotypes
  - ☐ Many strains are multidrug resistant
  - ☐ Diversity of geographical origin (isolated from 28 hospitals in 18 provinces of China between 2007 and 2020)
- MIC values (Minimum Inhibitory Concentration) of each strain measured by antibiotic susceptibility testing mainly for 6 Beta-lactam antibiotics :
  - ☐ Penicillin
  - ☐ Amoxicillin
  - ☐ Cefuroxime
  - ☐ Ceftriaxone
  - ☐ Cefepime
  - ☐ Imipenem



## Feature Selection



### The Challenge :

- ☐ The identification of resistance-related genes supports the relevance of the model.
- ☐ GWAS-based approach resulted in a highly correlated matrix with insufficient resolution for MIC prediction.
- ☐ Collinear features are a thread for Machine Learning efficiency.
- ☐ The initial number of SNPs does not allow us to compute the correlation matrix.
- ☐ Different population types required different approaches.

### The Results :

- ✓ We reduced the input data from 1,000,000 to 1,000 features
- ✓ Clusters contain all highly correlated features that allow us to study them afterward

### Low Variance Filtering

- SNPs with a variance lower than 0.05% are removed
- $\approx 0.95 > \text{prevalence} > 0.05$

### Per Gene SNP Clustering

- SNPs from the same orthogroup are clustered together according to their Spearman correlation (>95%)
- All features inside each cluster are represented by one single SNP

### Inter Gene Clustering

- Representative SNPs that share the same Spearman correlation to a random vector are clustered together (>95%)
- All features inside each cluster are represented by one single SNP

### Statistical Univariate Cluster Ranking

- Union of the  $n$  best clusters based on the KendallTau and Spearman correlations, Mutual information test, and ANOVA F-test
- $n$  will depend on the number of available samples

### Optimized Multivariate Feature Selection

- Union of the best feature combinations based on Random Forest, Linear Regressions and Relevance Vector Machine
- Feature numbers are chosen to optimize predictive model performance

## Conclusion and Perspectives

- ARSENAL is a new method that can predict a precise level of resistance in bacterial strains based on genomic data in a significantly shorter period of time but also pave the way for the discovery of new potential links between phenotype and genotype by identifying new genes and SNPs involved in resistance mechanisms
- We aim to identify some genes whose functions are related to antimicrobial resistance, but the function of many of the output genes of the model remains to be characterized (e.g., by site-directed mutagenesis) and may allow us to identify new resistance biomarkers.
- The method has been applied to strains of *Streptococcus pneumoniae* (gram-positive) and will be soon applied on several gram-negative Bacteria :
  - ☐ *Pseudomonas aeruginosa*
  - ☐ *Helicobacter pylori*
  - ☐ *Salmonella enterica*
  - ☐ *Klebsiella pneumoniae*
  - ☐ *Campylobacter jejuni*
  - ☐ *Escherichia coli*