

Using Machine Learning for Predicting Antimicrobial Resistance



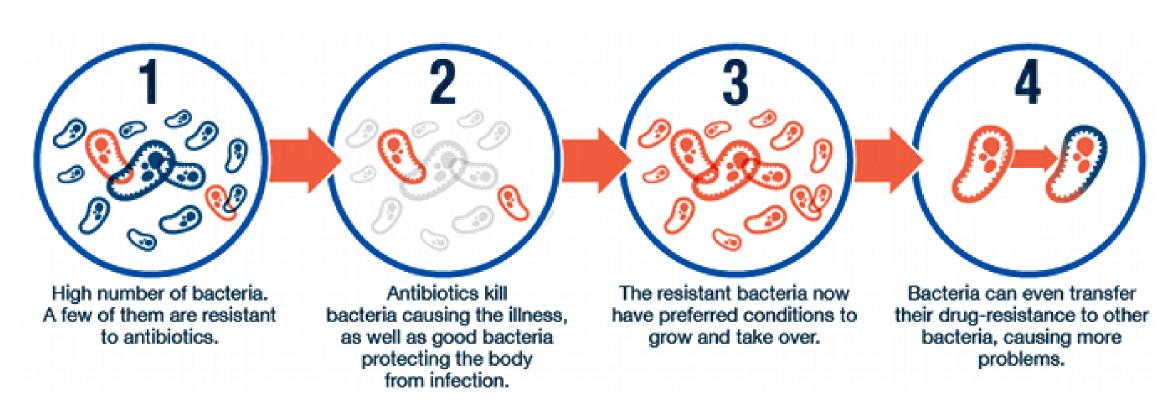
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Objectives

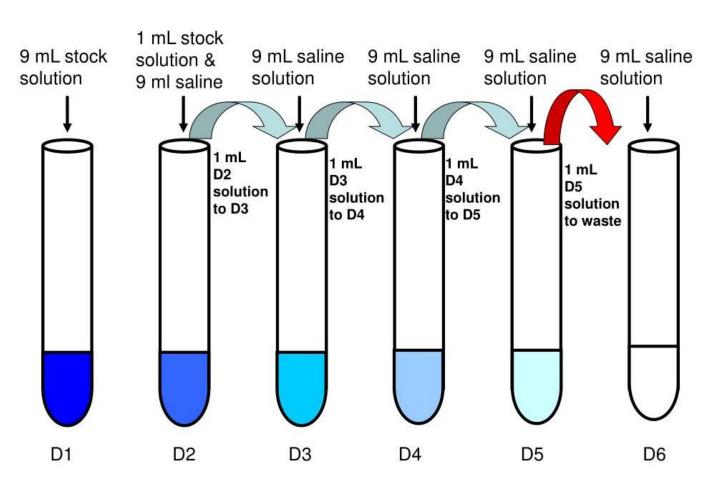
- To predict Minimum Inhibitory Concentration with a good accuracy, low cost and fast speed from the genome of a microbial strain
- To find the genetics determinants of antimicrobial resistance

Introduction

• Anti-Microbial Resistance (AMR) occurs when bacteria change in response to the use of medicines.

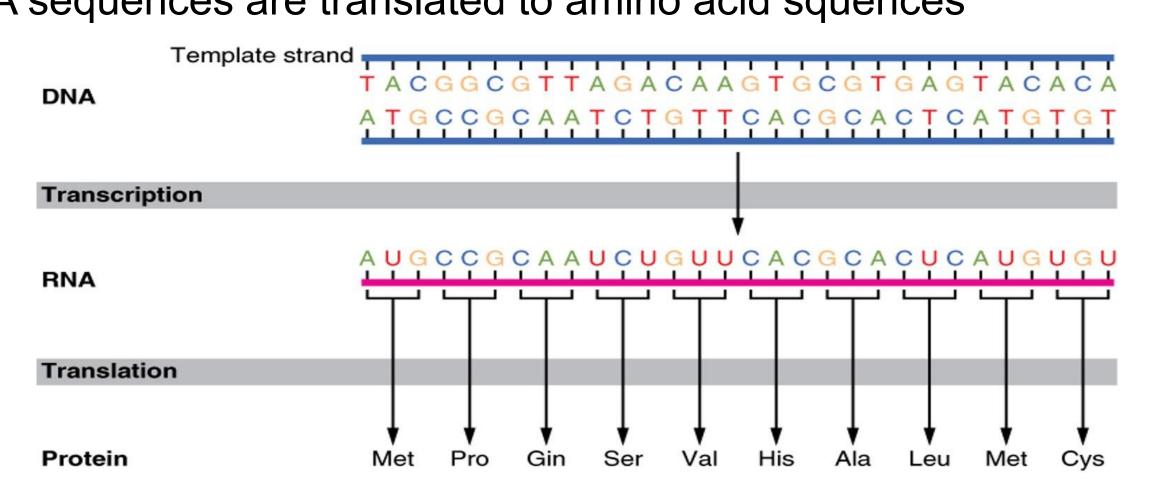


- The antibiotics should not be overprescribed. Only the minimum concentration must be used to treat the infection.
- MIC: Minimum inhibitory concentration: The lowest concentration of a chemical, usually a drug, which prevents visible growth of a bacterium



DNA and amino acid

- DNA of the bacterium contains the AMR information and can be used for predicting MIC
- DNA sequences are translated to amino acid squences



Methods

Data Preprocessing

DNA	Antibiotic	MIC
AACTTATATACACAGGGTACACAT	Ampicillin	2 g/ liter
GGGTACCCTGATGTGAACCATGRA	Ampicillin	1 g/liter
CCCGGTAGCTGACTGCTAGTGGGG	Tetracycline	4 g/ liter
ATCGGGGTCATTGCAAATGACCGT	Ampicillin	1 g/liter

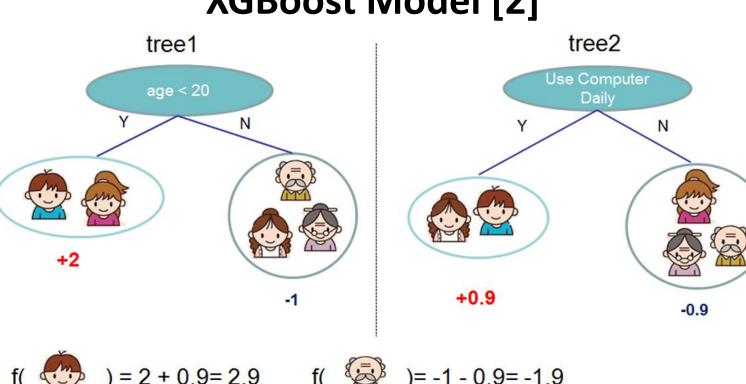
DNA is converted to k-mers.

AGTTGAG AGTTGAG AGTTGAG AGTTGAG AGTTGAG GTTGAG TTGAGT TGAGTT TGAGT TGAG

Double-Dilution Accuracy

FDA range of correct prediction [1-3]:

$$MIC_{predicted} \in \left[\frac{MIC_{target}}{2}, MIC_{target} * 2\right]$$
XGBoost Model [2]



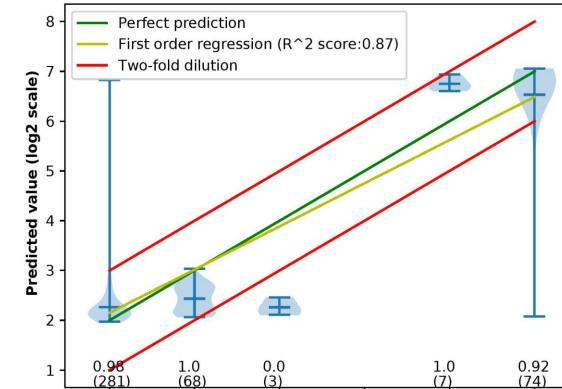
Results

Accuracy of prediction

Data	Mean accuracy	Standard Deviation
10-mers NT	0.935	0.012
3-mers AA	0.747	0.038
4-mers AA	0.941	0.012
5-mers AA	0.959	0.008

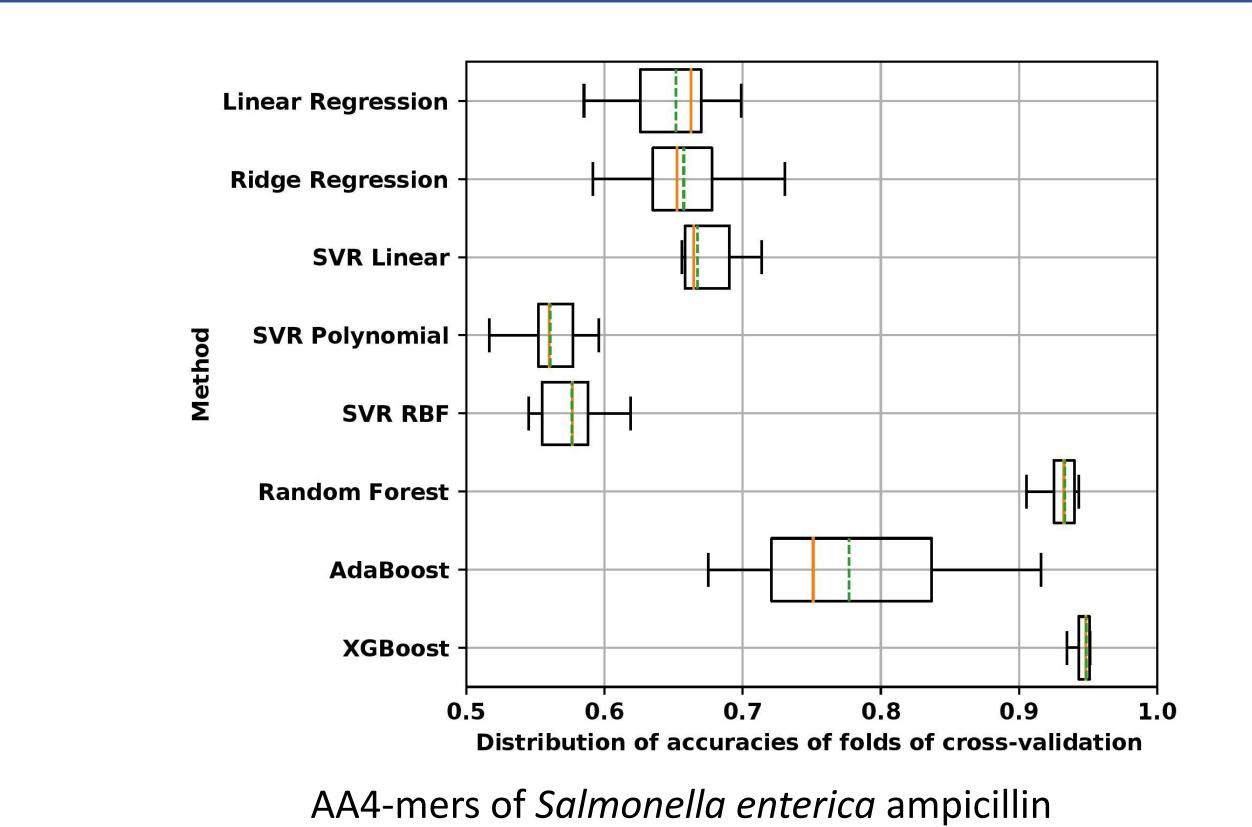
Accuracy of XGBoost for effect of Ampicillin on Salmonella enterica

Visualization of the predictions



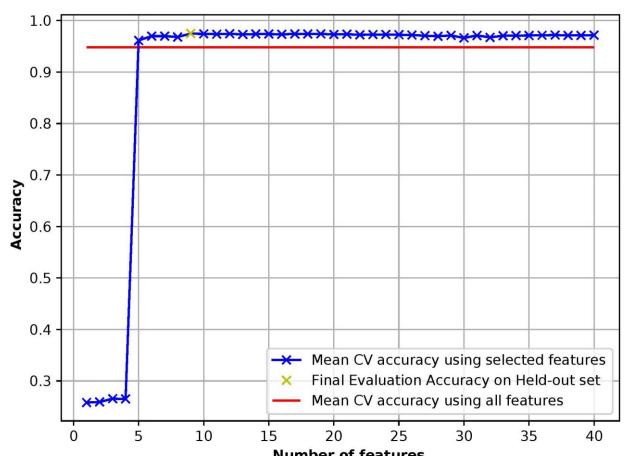
Distribution of prediction of values of MIC (effect of nalidixic acid on *Campilobacter Jejuni*. 4-mers of AA)

Other machine learning methods



Feature selection

- After training the model, the important features were extracted, and a new model was trained using only the selected features.
- Important genes were found by searching the important k-mers on NCBI database.





Left: Accuracy of the XG Boost model only using a few important features (effect of ampicillin on *Salmonella enterica* 4-mers of aminoacidic)
Right: Important genes that affect AMR for *Staphylococcus aureus* (extracted after training the model with 15-mers of nucleotide)

Conclusions

- Given enough number of genomes, machine learning can be used to predict MIC with a good accuracy
- Instead of nucleotide sequences, amino acid sequences can be used for increasing accuracy
- XG Boost outperforms other methods for MIC prediction from genome
- Feature selection can identify genes that are important to AMR

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References

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[2] T. Chen, C. Guestrin, XGBoost: A Scalable Tree Boosting System, (n.d.).

[3] M. Nguyen, S.W. Long, P.F. McDermott, R.J. Olsen, R. Olson, R.L. Stevens, G.H. Tyson, S. Zhao, J.J. Davis, Using Machine Learning To Predict Antimicrobial MICs and Associated Genomic Features for Nontyphoidal Salmonella, J. Clin. Microbiol. 57 (2018). doi:10.1128/JCM.01260-18.