

# MIC Predictions for Four $\beta$ -lactam Agents for *Escherichia coli* and *Klebsiella pneumoniae* from a Large Surveillance Program Using Genomic Data and a Machine Learning Model

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## Objective

To predict the MIC values for four  $\beta$ -lactam agents for *E. coli* and *K. pneumoniae* isolates using MIC and genomic data from the SENTRY Antimicrobial Surveillance Program using the Random Forrest machine learning algorithm.

## Methods



A total of 3054 *E. coli* and 2940 KPN isolates from 2016 and 2017 were susceptibility tested using the CLSI reference broth microdilution method.



Isolates displaying  $\beta$ -lactam and/or aminoglycoside resistance were submitted to whole genome sequencing for the identification of genes encoding  $\beta$ -lactamases.



MIC and genetic results from 2016 and 2017 were used to train the Random Forrest machine learning algorithm.

- MIC results from 2016 to 2018 were predicted for ceftriaxone, ceftazidime, cefepime, and meropenem.
- Error rates were calculated according to CLSI M23 criteria upon comparison to BMD MICs.

# Results

**Table 1. Error rates for predicted versus tested MIC values for four  $\beta$ -lactam agents against *E. coli* and *K. pneumoniae* isolates**

Organism (no. of isolates) Breakpoints <sup>a</sup>	Error type	Ceftriaxone		Ceftazidime		Cefepime		Meropenem	
		No. of isolates	%						
<i>E. coli</i> (3054)									
CLSI	Minor	13	1.4	90	9.4	136	14.2	3	0.3
	Major	141	14.8	162	17.0	36	3.8	459	48.0
	Very Major	0	0	41	4.3	42	4.4	0	0
EUCAST	Minor	0	0	77	8.1	35	3.7	8	0.8
	Major	145	15.2	111	11.6	20	2.1	458	47.9
	Very Major	1	0.1	42	4.4	60	6.3	0	0
ECV	Minor	0	0	0	0	0	0	0	0
	Major	128	13.4	28	2.9	0	0	701	73.3
	Very Major	0	0	0	0	0	0	4	0.4
<i>K. pneumoniae</i> (2940)									
CLSI	Minor	18	1.8	50	4.9	72	7.1	142	13.9
	Major	28	2.8	41	4.0	45	4.4	414	40.6
	Very Major	8	0.8	28	2.8	18	1.8	4	0.4
EUCAST	Minor	0	0	13	1.3	31	3.0	59	5.8
	Major	30	2.9	25	2.5	32	3.1	419	41.1
	Very Major	9	0.9	31	3.0	25	2.5	9	0.9
ECV	Minor	0	0	0	0	0	0	0	0
	Major	19	1.9	6	0.6	0	0	482	47.3
	Very Major	4	0.4	12	1.2	0	0	9	0.9

<sup>a</sup> Criteria published by CLSI (2022), EUCAST (2022) and ECVs determined using the entire dataset according to CLSI M57

## Conclusions

- Among machine learning prediction methods, the Random Forrest algorithm is capable of learning complex data representations to make accurate predictions that generate random decision trees.
- The Random Forrest machine learning algorithm was able to predict MICs for ceftriaxone for *K. pneumoniae* with acceptable error rates.
- For other cephalosporins, ECVs generated acceptable error rates, but not clinical breakpoints.
- Meropenem MIC predictions had high error rates, potentially due to the small sample of carbapenem-resistant isolates in the studied population.
- Machine learning algorithms should be further explored to predict MIC results, but the use of breakpoints that are established from clinical outcomes and PK/PD outcome studies might not be ideal to interpret these predictions.

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