

# Predicting the Severity of Adverse Drug Reactions

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## Abstract

Clinicians use databases such as Lexi-Interact to determine the overall severity of side effects from prescribed drug combinations [1]. However, many drug combinations are not found within such databases, though adverse side effects of such combinations have been reported to the FDA's Adverse Event Reporting System. We used **multinomial classification methods**, SVM, Naive Bayes, Logistic Regression and Random Forests, to **predict drug-drug interaction severity values from the adverse drug reactions in the FDA's database**. SVM and Random Forests both had classification **accuracies of over 95%** though SVM had both higher overall accuracy and higher recall for the most severe severity labels.

## Models

### Logistic Regression

$$\text{Maximize: } LL(\theta) = \sum_{i=1}^n y^{(i)} \log(\sigma(\theta^T \cdot x^{(i)})) + (1 - y^{(i)}) \log[1 - \sigma(\theta^T \cdot x^{(i)})]$$

L2 regression, cross-entropy loss

### Naive Bayes

$$\text{Maximize: } \prod p(x^i | y^i) p(y^i)$$

### Random Forest

$$\text{Minimize: } \sum q_m \sum p_{mk} (1 - p_{mk})$$

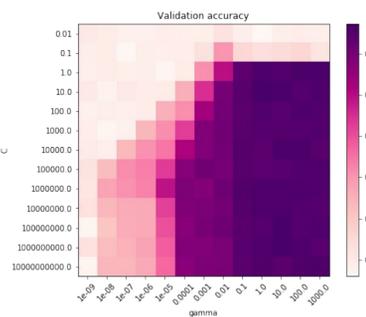
$q$  represents the ratio of samples in region  $m$  and  $p$  represents the ratio of class  $k$  in region  $m$

### SVM

Optimize:

$$\text{Hyperparameter: decision tree number } \min_{w,b} \frac{1}{2} \|w\|^2 \text{ given } y^{(i)}(w^T x^{(i)} + b) \geq 1 \text{ for all } i$$

RBF Kernel: tuned gamma, C



(Left) Figure 2. Hyperparameter tuning for SVM. Optimal C value is 10 with a gamma of  $10^{10}$ .

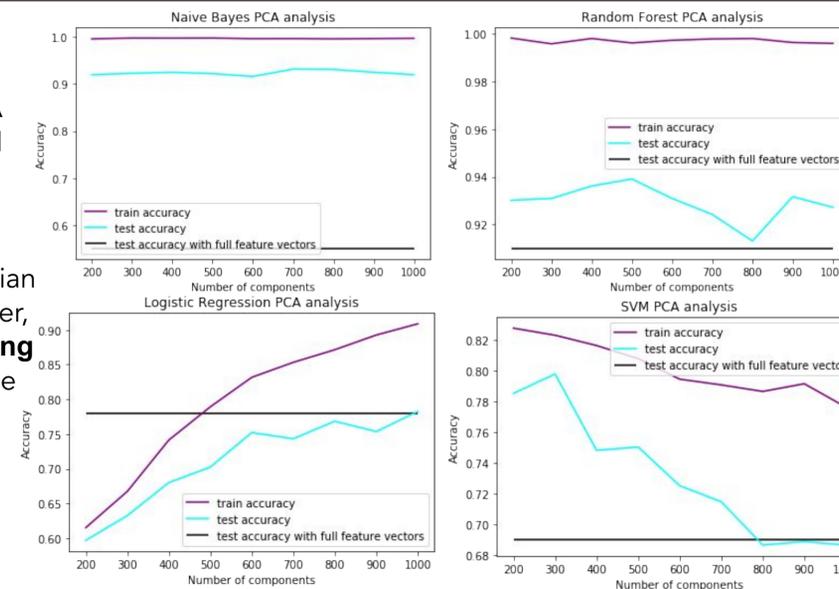
Gamma is a measure of how far a single example's influence can spread and C determines the relative weight of accuracy versus widening the decision boundary.

## References

[1] Debruyne, P. R., Pottel, L., Lycke, M., Boterberg, T., Ketelaars, L., Pottel, H., ... & Rottey, S. (2012). Experience with Lexicomp® Online Drug Database for medication review and drug-drug interaction analysis within a comprehensive geriatric assessment in elderly cancer patients. *Journal of Analytical Oncology*, 1(1), 32-41.  
 [2] Tatonetti, N. P., Patrick, P. Y., Daneshjoo, R., & Altman, R. B. (2012). Data-driven prediction of drug effects and interactions. *Science translational medicine*, 4(125), 125ra31-125ra31.  
 [3] Up-to-date [online]. Lexi-interact online. www.uptodate.com/crslq/interact/framest.jsp (accessed 3 June 2019).

## Data/Features

The presence or absence of **1,317 potential adverse drug reactions** recorded in the FDA database will be a sparse vector **transformed via PCA** to represent each of the ~63,000 drug-drug interactions [2]. The true labels for these interactions are one of 5 classes from Lexi-Interact, which were combined by physician action to help combat class imbalance. Further, **minority class upsampling within the training set** was used to eliminate class imbalance. The intersection of drug-drug interaction records from the FDA database with the drug-drug severity scores from Lexi-Interact resulted in **3,646 drug pairs**.



(Left) Figure 1. Accuracy across varying number of PCA components with an 80/20 train/test split and 10-fold cross-validation to determine the optimal number of PCA components for the feature vectors.

Table 1. Frequency of Labels

Label	Frequency
No Action	225
Consider Modification	2604
Action Required	817

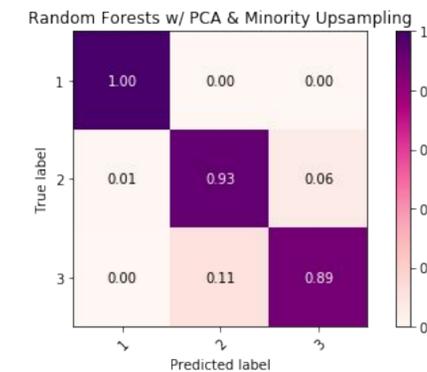
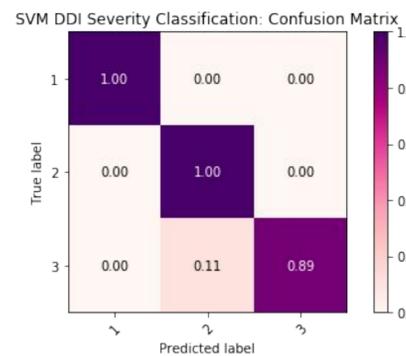
## Evaluation

Table 2. Classification Results.

Accuracy for each method with optimal hyperparameters and feature vectors with an 80/20 train/test split and 10-fold cross-validation.

Model	Train Accuracy	Test Accuracy
Logistic Regression	91.0%	79.7%
Naive Bayes	99.7%	93.1%
Random Forest	99.9%	95.2%
Multi-Class SVM	99.9%	96.5%

(Below) Figure 3. Normalized Confusion Matrix for the optimal SVM model. Represents 729 test examples. Labeled as:  
 (1) No Action  
 (2) Consider Modification  
 (3) Action Required



(Above) Figure 4. Normalized Confusion Matrix for the optimal Random Forest Model. Represents 729 test examples. Labeled as:  
 (1) No Action  
 (2) Consider Modification  
 (3) Action Required

Table 3. Recall/Precision by class for SVM and Random Forest Models.

	Random Forests (Precision/Recall)	SVM (Precision/Recall)
No Action	1.00/1.00	1.00/1.00
Consider Modification	0.91/0.97	0.91/1.00
Action Required	0.96/0.88	1.00/0.89

## Discussion

- Logistic Regression seems to have a problem with **over-fitting**
- PCA transformation** greatly improved Naive Bayes accuracy, probably as it eliminated sparsity
- SVM accuracy was greatly influenced by **hyperparameters**
- Low Recall for "Action Required" perhaps a function of **drug delivery method** as some drugs can be delivered multiple ways (orally, systemically, optically) which influences drug interaction severity

## Future

Validate the predicted severity of a drug pair not in Lexi-Interact against a panel of clinical pharmacists, individuals familiar with clinical outcomes of drug interactions. This is a crucial step as there may be a **selection bias** when only training on drug pairs found within Lexi-Interact.