

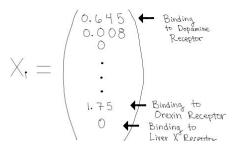


## Overview

- We construct a deep neural network to predict the binding affinity to serotonin receptors.
- Understanding the binding behavior of these receptors is important for a diverse range of future mental health drug research.
- We show representing small molecules as vectors with binding constants for other targets can provide predictive information.
- Our neural network approach outperforms a simple linear regression model as well as a more complex matrix completion method. The metric used is the  $F_1$  score.

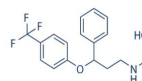
## Data Representation

- To represent the small molecule we construct a vector in which the  $i^{th}$  entry gives the binding constant,  $K_B$ , of the small molecule to the  $i^{th}$  chosen target (939 other targets aside from serotonin).
- A large  $K_B$  value indicates that the small molecule binds well to the target.

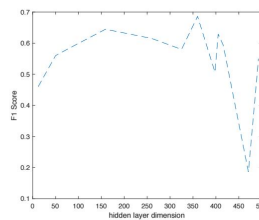
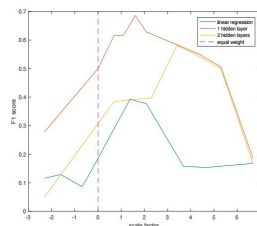


## Experimental Results

Comparison Methods: Sparse Matrix Completion, Linear Regression  
Metric of performance:  $F_1$  score



$$F_1 = \frac{2 * \text{true positive}}{2 * \text{true positive} + \text{false negative} + \text{false positive}}$$



Method	$F_1$ Score	Positive Example Accuracy	Negative Example Accuracy
Linear Regression	0.423	0.46/0.84	0.98/0.38
Matrix Completion	0.571	<b>0.80</b>	0.88
Neural Network	<b>0.686</b>	0.66/0.77	<b>0.98/0.89</b>

## References

Han Altae-Tran, Bharath Ramsundar, Aneesh S Pappu, and Vijay Pande. Low data drug discover with one-shot learning. *ACS central science*, 3(4):283-293, 2017.

David E Nichols and Charles D Nichols. Serotonin receptors. *Chemical reviews*, 108(5):1614-1641,2008.

The binding database.

## Model

Loss function:

$$L(p) = -\frac{1}{N} \sum_{i=1}^N s(y_i \log p_i + (1 - y_i) \log(1 - p_i)).$$

- $s$  allows us to give greater weight to correctly identifying positive examples. We include this to deal with the sparsity of positive examples in the dataset.
- We explore the effect of  $s$  as well as the number of layers of our neural network on the performance (as measured by F1 score). For a  $k$ -layer net we have,

$$A_0 = X,$$

$$A_j = f(W_j A_{j-1} + b_j),$$

$$p = \sigma(W_k A_{k-1} + b_k).$$

Where  $f()$  is a ReLU or a sigmoid.

- Prediction is done by:

$$\hat{y}_i = p > 0.5$$

## Discussion

- A shallow neural network architecture shows promising performance in predicting the binding affinity to serotonin receptors.
- Future research must implement more comparison methods to get a better sense of the kind of  $F_1$  score that would be helpful.
- Unlike matrix completion methods, constructing a loss function with scaling factor  $s$  allows freedom in tuning the specificity/sensitivity ratio
- To increase speed of training, dimensionality reduction methods should be explored in the vector representation of the small molecule.

**Thank you!!**