

#### Mechanisms of Action (MoA) Prediction

Can you improve the algorithm that classifies drugs based on their biological activity?

JIE LIU, PH.D.

## The Project

- A Kaggle competition organized by MIT and Harvard
- ► The dataset combines gene expression and cell viability data as measurements of human celllular responses to drug treatment
- The task is to use this dataset to develop a machine learning model that automatically labels any new drug as one or more MoA types

## My Objective

build all kinds of classification models and compare their performances

# The Challenges

- A multi-label classification, 206 labels in total
- Custom evaluation metric required, logarithmic loss function, formula provided

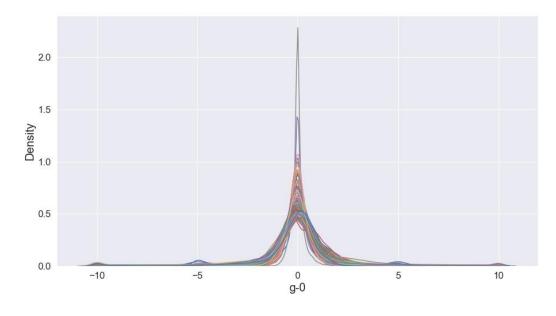
$$ext{score} = -rac{1}{M} \sum_{m=1}^{M} rac{1}{N} \sum_{i=1}^{N} \left[ y_{i,m} \log(\hat{y}_{i,m}) + (1-y_{i,m}) \log(1-\hat{y}_{i,m}) 
ight]$$

Features are highly correlated to each other

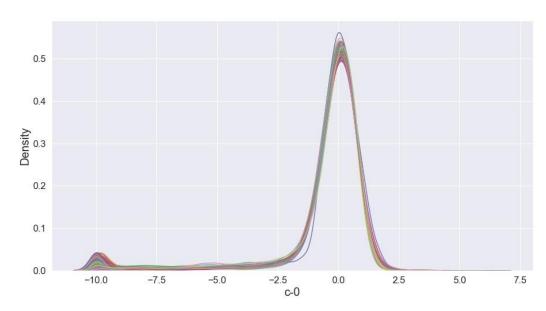
#### Gene Expression G-0 to G-771

# Exploratory Data Analysis

- ► 17860 drug samples, 875 features including gene expression and cell viability patterns in response to drug treatment, 206 labels
- ▶ 872 numerical features
- Data pre-normalized, following normal-like distributions

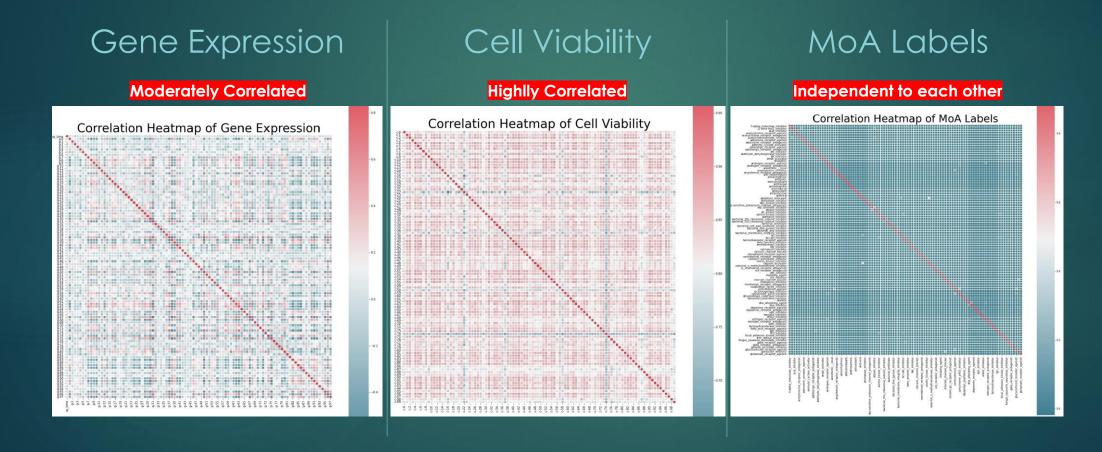


Cell Viability C-0 to C-99



# Exploratory Data Analysis (cont.)

High correlation suggests PCA dimensionality reduction



## Data Processing

- No need to normalize data again
- ▶ Data split into 60% train, 20% validation and 20% holdout

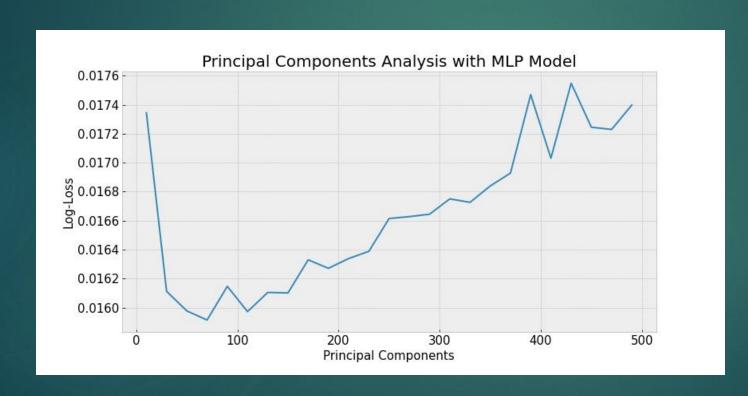
## First Two Neural Network Models

Baseline Model	Neural Network Structure	Parameters	Log-loss
Multilayer Perceptron	1 input, 1 hidden, 1 output layers	91,470	0.0183
1-D Convolutional NN	1 input, 2 hidden, 1 output layers	905,102	0.0194

MLP model yields smaller thus better log-loss.

## Principal Component Analysis (PCA)

#### Dimensionality Reduction



Original feature number: 875

n\_components scan:10 to 500

Best log-loss: at 70

Log-loss MLP: 0.0159 (down from 0.0183)

Log-loss 1D-CNN: 0.0180 (down from 0.0194)

## 15 Classification Models

#### MLP and All kinds of Convolutional Neural Networks

	MLP	1D-CNN	AlexNet	LeNet-5	VGG-16 Net	ResNet	Inception Net
Parameters	91,470	905,102	23,337,214	116,110	28,828,174	340,430	5,235,936
Log-loss	0.0159	0.0180	0.0199	0.0170	0.0176	0.0191	0.0189
	*Best			*			

#### Other Classification Models

	Random Forest	SVC	KNN	XGBoost	Adaboost	Logistic Regression	GaussianNB	Decision Tree
Log-loss	0.0165	0.0167	0.0168	0.0170	0.0172	0.0177	0.0434	0.0684
	*	*	*	*				

<sup>\*</sup> Six models selected for further optimization

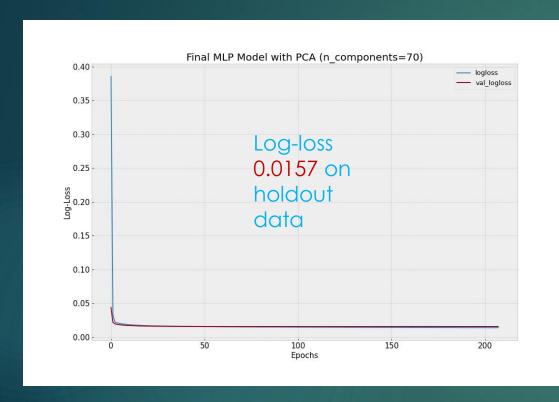
## Hyperparameter Tuning - Hyperopt

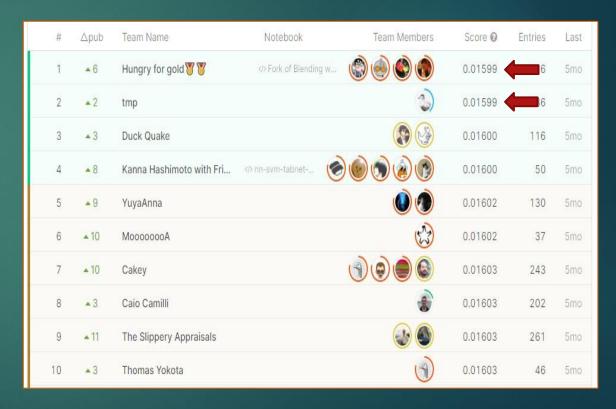
- A powerful python library for hyperparameter tuning, using Bayesian optimization algorithm.
- Allowing the optimization of hundreds of parameters efficiently.
- Highly recommended over Sklearn RandomizedSearch and GridSearch optimization.

	Log-loss Before Optimization	Log-loss After Optimization
Multilayer Perceptron (MLP)	0.0159	0.0156
LeNet-5	0.0170	0.0161
C-Support Vector Classification (SVC)	0.0167	0.0164
Gradient Boost XGBoost	0.0170	0.0166
RandomForest	0.0165	0.0166
K Nearest Neighbours (KNN)	0.0168	0.0168

## My Final MLP Model

1 Input layer, 2 Hidden Layers, 1 Output Layer, Batch Normalization and Dropout





Top 10 Winners of the Competition

### Conclusions

▶ The simple MLP neural network model works best for my project.

## Techniques Most Helpful

- Principle component analysis (PCA)
- Hyperparameter optimization with Hyperopt



## Jie Liu, Ph.D.

- ▶ jliu1999@gmail.com
- ▶ Tel: 917-306-8708
- ▶ github.com/jliu1999
- ▶ linkedin.com/in/jieliu1999





















