## Assignment 3 Lab Report

# Regression:

For our regression testing, we used the Computer Hardware dataset from <a href="https://archive.ics.uci.edu/ml/datasets/Computer+Hardware">https://archive.ics.uci.edu/ml/datasets/Computer+Hardware</a>.

Number of Features	7
Names of the features	MYCT, MMIN, MMAX, CACH, CHMIN, CHMAX, PRP
Name of target	ERP
Number of samples	209
Description of data	(209, 8)

### First five rows of data:

Row#	MYCT	MMIN	MMAX	CACH	CHMIN	CHMAX	PRP
1	125	256	6000	256	16	128	198
2	29	8000	32000	32	8	32	269
3	29	8000	32000	32	8	32	220
4	29	8000	32000	32	8	32	172
5	29	8000	32000	32	8	16	132

### Correlation between the features:

	MYCT	MMIN	MMAX	CACH	CHMIN	CHMAX	PRP
MYCT	1.000000	-0.335642	-0.378561	-0.321000	-0.301090	-0.250502	-0.307099
MMIN	-0.335642	1.000000	0.758157	0.534729	0.517189	0.266907	0.794931
MMAX	-0.378561	0.758157	1.000000	0.537990	0.560513	0.527246	0.863004
CACH	-0.321000	0.534729	0.537990	1.000000	0.582245	0.487846	0.662641
CHMIN	-0.301090	0.517189	0.560513	0.582245	1.000000	0.548281	0.608903
CHMAX	-0.250502	0.266907	0.527246	0.487846	0.548281	1.000000	0.605209
PRP	-0.307099	0.794931	0.863004	0.662641	0.608903	0.605209	1.000000

### Best Parameters:

Using the GridSearchCV function of sklearn, we able to automatically try multiple parameters of alpha. We were able to deduce the optimal alpha for both Lasso and Ridge Regression.

# Random\_state = 43

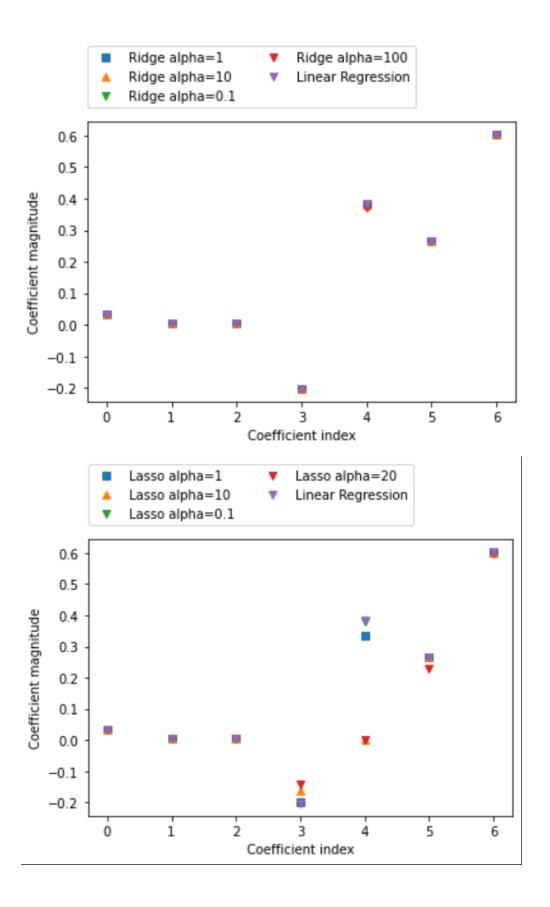
Ridge	100
Lasso	20

After finding the optimal parameter alpha, we used that parameter to predict the target data. We used  $R^2$  and RMSE as our scoring methods to measure the accuracy of our methods.

Method	$R^2$	RMSE
Lasso	0.8821	5.2389
Ridge	0.8821	5.2389
Linear	0.8820	5.2399

Comparison of Ridge, Lasso, and Linear Regression Models:

Using the optimal parameter that we found earlier, we made a linear model with calculated coefficients. The following figures show those coefficients and how they change with the parameter alpha.



# Classification

For our classification testing, we used Haberman's Survival dataset from <a href="https://archive.ics.uci.edu/ml/datasets/Haberman%27s+Survival">https://archive.ics.uci.edu/ml/datasets/Haberman%27s+Survival</a>.

Number of Features	3
Names of the features	Age. Year of Operation, Nodes
Name of target	Survival Status
Number of samples	306
Description of data	(306, 4)

#### Correlation between features:

	age	year_of_operation	nodes
age	1.000000	0.089529	-0.063176
year_of_operation	0.089529	1.000000	-0.003764
nodes	-0.063176	-0.003764	1.000000

After training our Logistic Regression model, we calculated the accuracy of our model to be 75.8%.

Accuracy of the Logistic Regression: 0.75806

With our model, we ran a prediction of the test target data using the test data. With the predictions, we were able to create a confusion matrix and solve for precision, recall, sensitivity, and accuracy, as seen below.

#### Model Performance:

Precision	0.9347
Recall	0.7818
Sensitivity	0.7818
Accuracy	0.7580

#### **ROC Curve:**

To determine the effectiveness of the model, we plotted a ROC Curve. From our data, we can see our model is not performing the best it could. There could be a model better suited to this

data.

