

# Breast Cancer Prediction

IST 718 Final Project Report

Group 1

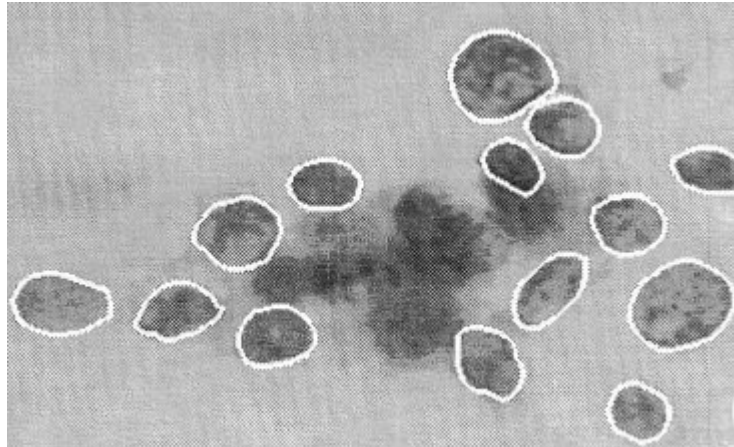
Mike DeMaria, Lu Guo, Haotian Shen, Casey Walsh

# Introduction

When a tumor is detected, it needs to be assessed if it's benign or malignant.

Nowadays, Machine Learning has been widely used to help detect tumors.

We want to know that given the output of the image analysis, can we predict malignancy?



# Data Exploration

We obtain breast cancer data from Kaggle:

<https://www.kaggle.com/datasets/uciml/breast-cancer-wisconsin-data/data>

569 rows

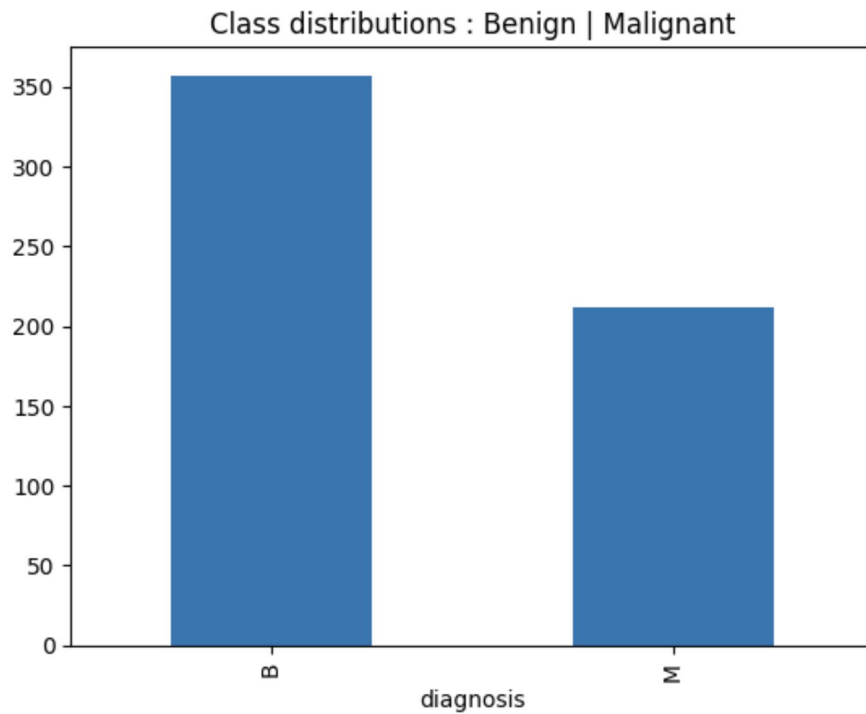
32 columns:

- patient ID
- 1 binary output variable (benign or malignant)
- 30 numerical variables of cancer cells

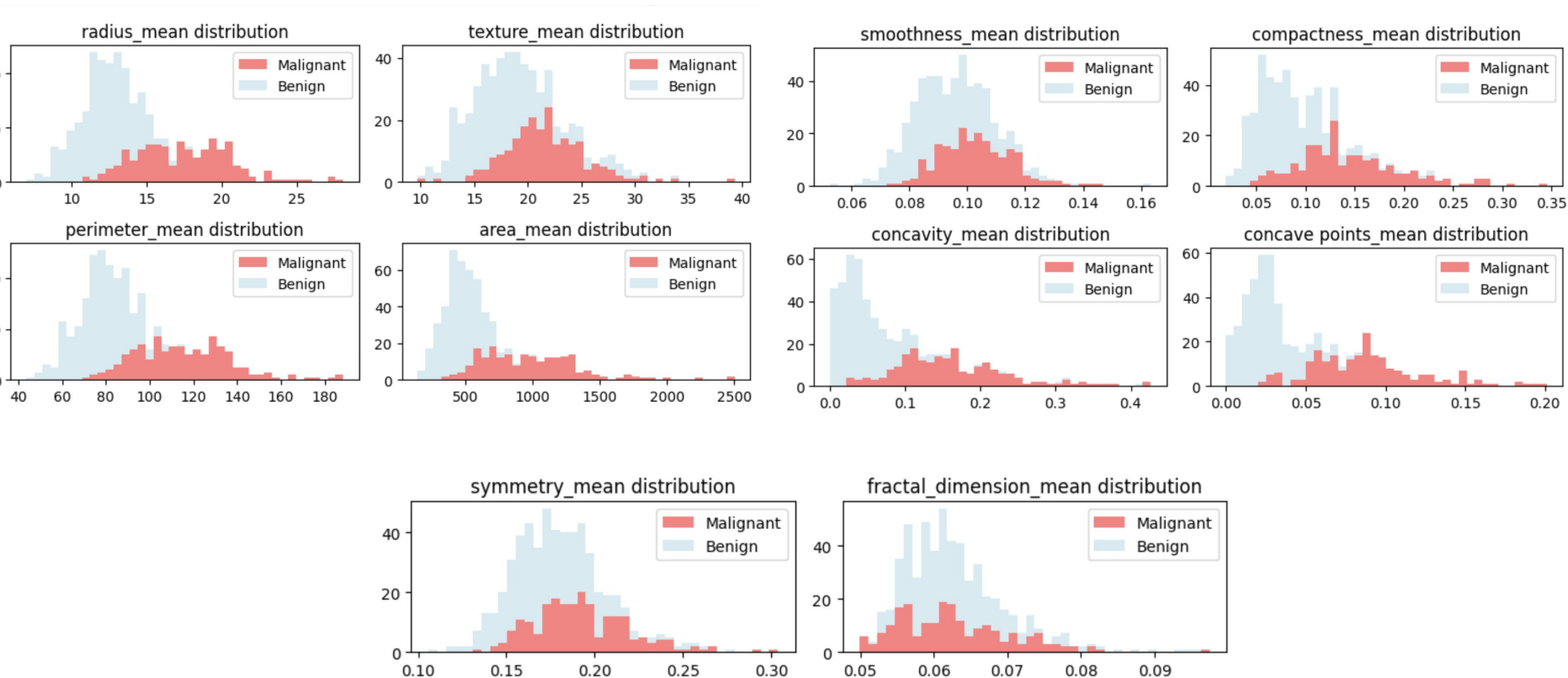
10 categories: radius, texture, perimeter, area, smoothness, compactness, concavity, concave points, symmetry, and fractal dimension.

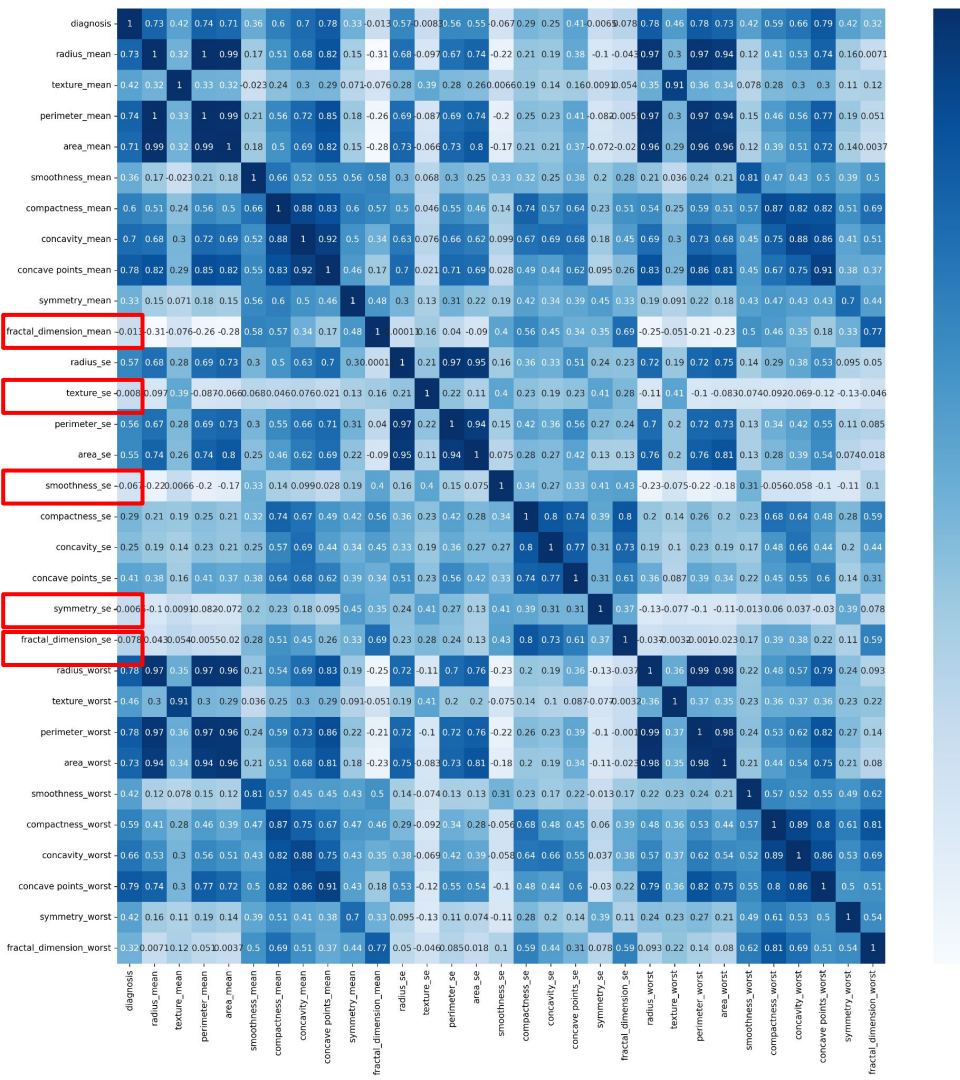
3 values for each feature: the **mean**, the **standard error**, and the mean of the three **worst** (largest) cells.

# Data Exploration



# Data Exploration





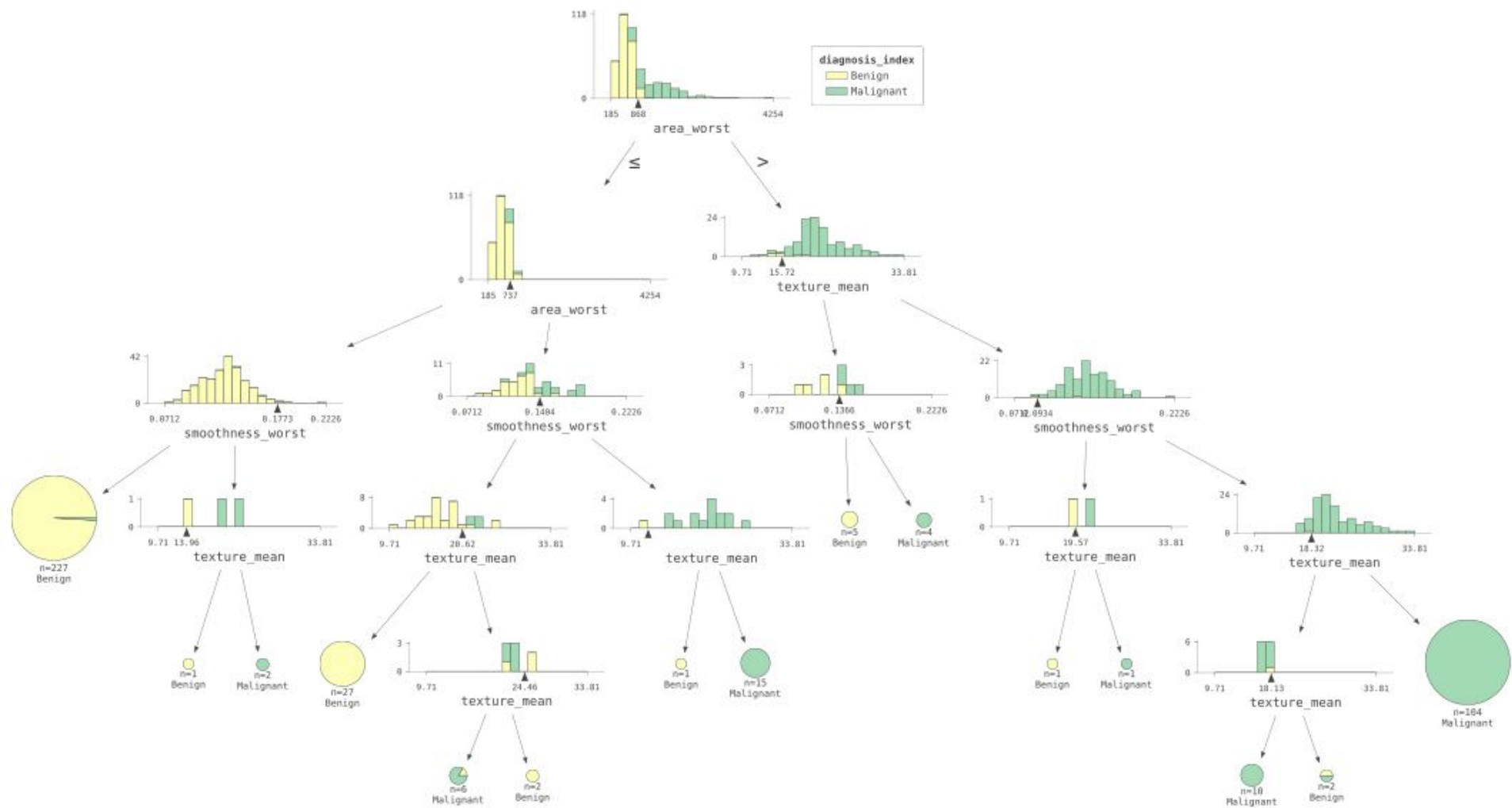
The below columns have very low correlation to diagnosis:

- fractal\_dimension\_mean,
- texture\_se,
- smoothness\_se,
- symmetry\_se,
- fractal\_dimension\_se.

When doing feature engineering, maybe we can try to remove these columns.

# Original Model

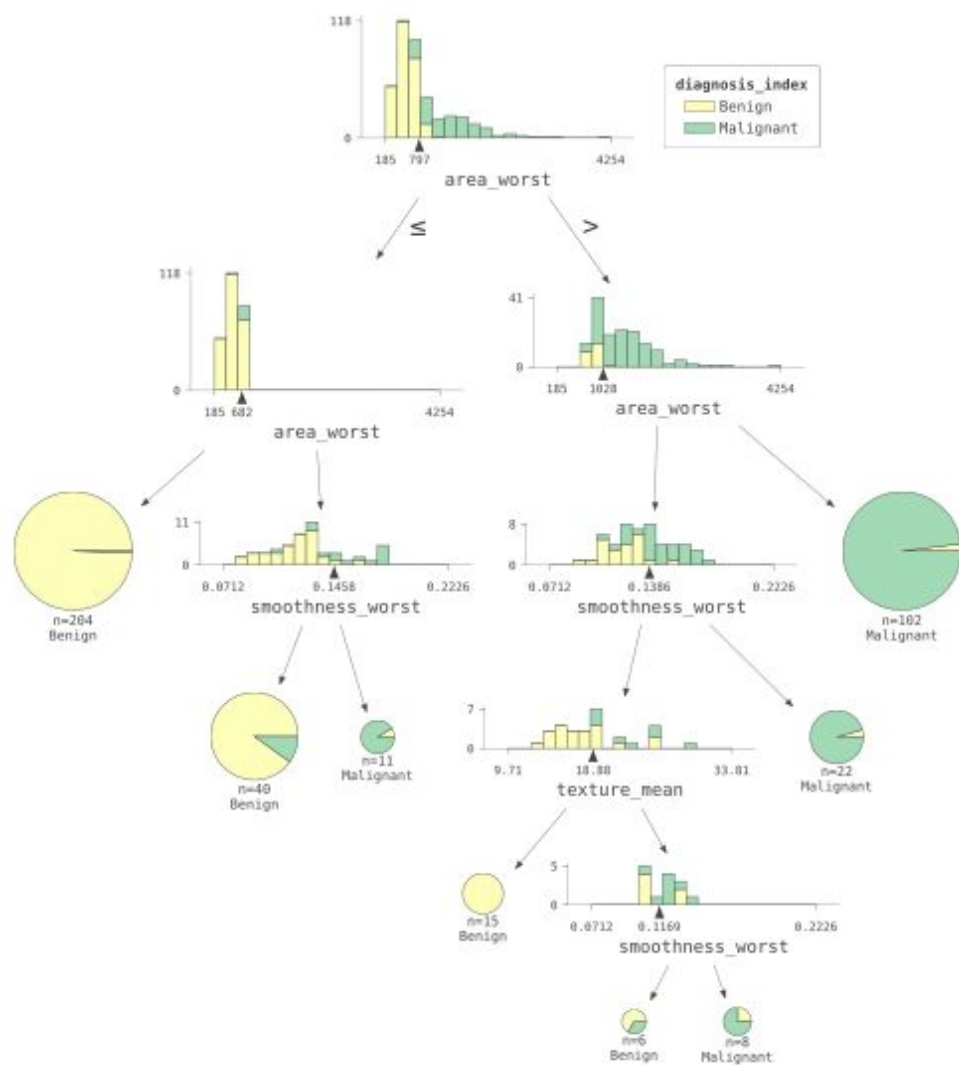
- Decision Tree Model
- Used 3 of the thirty available features
  - Mean Texture
  - Worst Area
  - Worst Smoothness
- Small subsets chosen for compute speed
- Original Model achieved 97% accuracy
- Attempt to replicate this model achieved 92% accuracy
  - Depth 6
  - 28 nodes



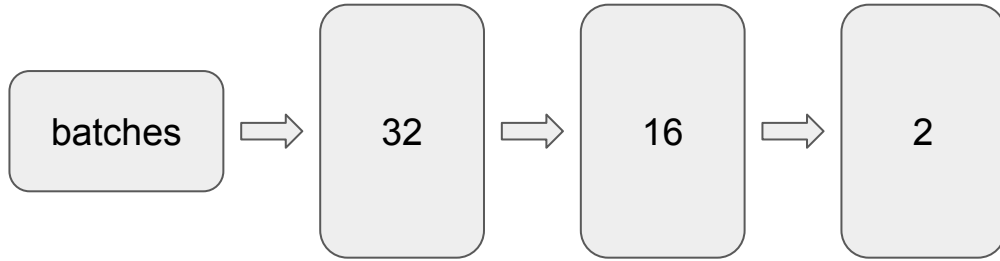


# Recreating the Original Model

- Decision Tree Model trained with the same three features
  - Trained in under 10 seconds
- Initial accuracy of 92%
- Retrained model with three features and hyperparameter grid search
  - Trained 120 models in approximately 10 minutes
  - Best model parameters:
    - Depth 5
    - 15 nodes
- Final accuracy of 96%



# Neural Network



Select all mean features

Confusion Matrix:

	Predict Benign	Predict Malignance
Actual Benign	99.0	3.0
Actual Malignance	8.0	49.0

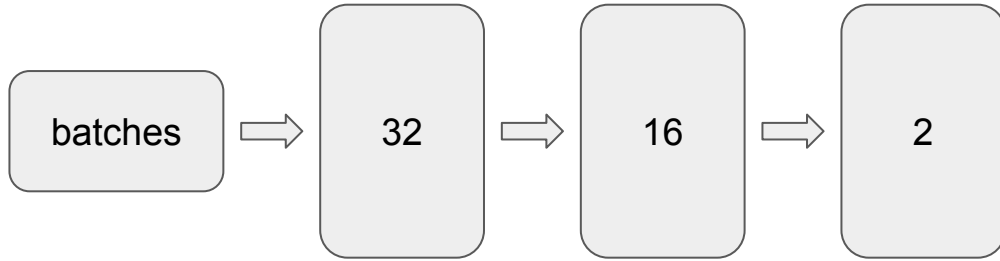
=====  
Accuracy: 0.9308176100628931

Precision: 0.9313545297939585

Recall: 0.9308176100628931

F1-score: 0.9300583985496023

# Neural Network



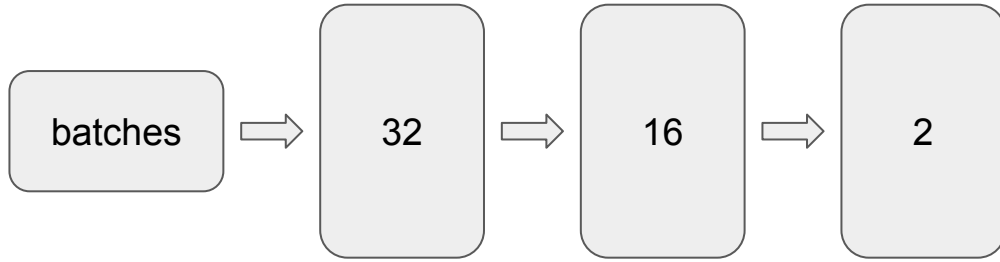
Select all standard error features

Confusion Matrix:

	Predict Benign	Predict Malignance
Actual Benign	99.0	3.0
Actual Malignance	12.0	45.0

=====  
Accuracy: 0.9056603773584906  
Precision: 0.9082419683834777  
Recall: 0.9056603773584906  
F1-score: 0.9036103412930414

# Neural Network



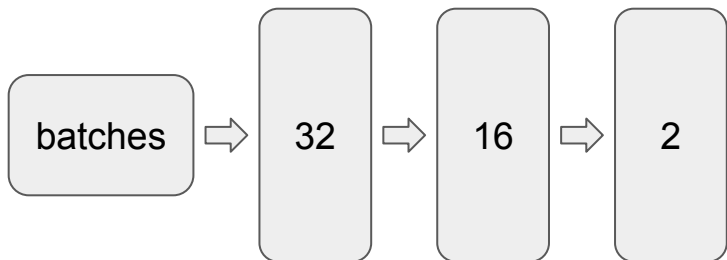
Select all worst features

Confusion Matrix:

	Predict Benign	Predict Malignance
Actual Benign	97.0	5.0
Actual Malignance	5.0	52.0

=====  
Accuracy: 0.9371069182389937  
Precision: 0.9371069182389937  
Recall: 0.9371069182389937  
F1-score: 0.9371069182389937

# Neural Network

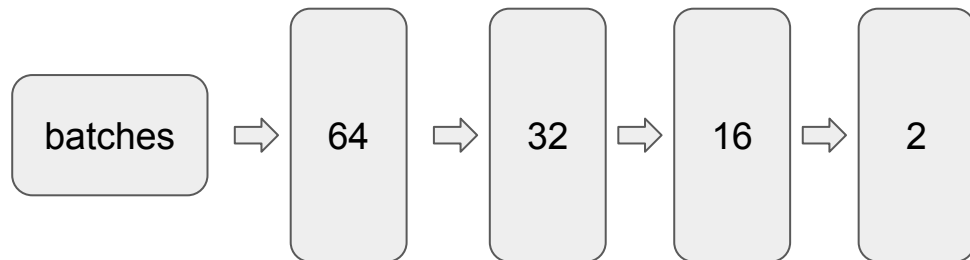


Confusion Matrix:

	Predict Benign	Predict Malignance
Actual Benign	84.0	9.0
Actual Malignance	8.0	60.0

=====

Accuracy: 0.8944099378881988  
Precision: 0.8946799891979476  
Recall: 0.8944099378881987  
F1-score: 0.8945099245321397



Confusion Matrix:

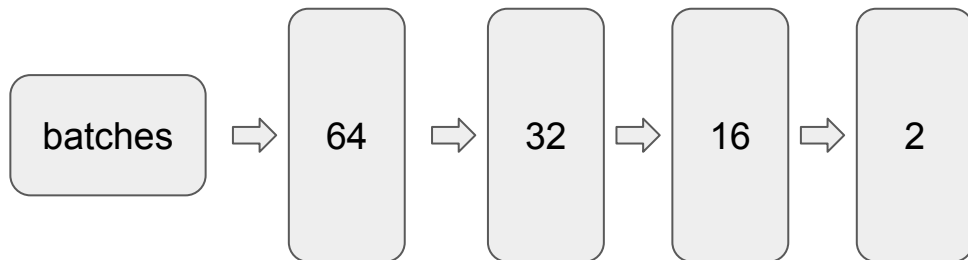
	Predict Benign	Predict Malignance
Actual Benign	97.0	5.0
Actual Malignance	11.0	46.0

=====

Accuracy: 0.89937106918239  
Precision: 0.899514942245242  
Recall: 0.89937106918239  
F1-score: 0.8980133772586603

Select all features

# Neural Network



Drop

'fractal\_dimension\_mean',  
'texture\_se', 'smoothness\_se',  
'symmetry\_se',  
'fractal\_dimension\_se'

Confusion Matrix:

	Predict Benign	Predict Malignance
Actual Benign	100.0	2.0
Actual Malignance	10.0	47.0

=====  
Accuracy: 0.9245283018867925  
Precision: 0.9270486925473448  
Recall: 0.9245283018867925  
F1-score: 0.9231043075827697

# Neural Network

	accuracy	precision	recall	f1-score
All mean features	0.9308	0.9314	0.9308	0.9301
All standard error features	0.9057	0.9082	0.9057	0.9036
<b>All worst features</b>	<b>0.9371</b>	<b>0.9371</b>	<b>0.9371</b>	<b>0.9371</b>
All features	0.8944	0.8947	0.8944	0.8945
All features(large)	0.8994	0.8995	0.8994	0.8980
All features+Drop(large)	0.9245	0.9270	0.9245	0.9231



# Additional Models

Algorithm	Data Used	AUC	Runtime (seconds)
Logistic Regression	“Worst” Cells	95.5%	2.1
Random Forest	All	97.9%	6.2
Logistic Regression (Increased iterations)	Mean/Std Dev	98.3%	4.8
GBT	All	98.9%	11.8
Logistic Regression	Mean/Std Dev	99.1%	2.14
LinearSVC	All	99.6%	28.3
Logistic Regression	All	99.7%	4.2

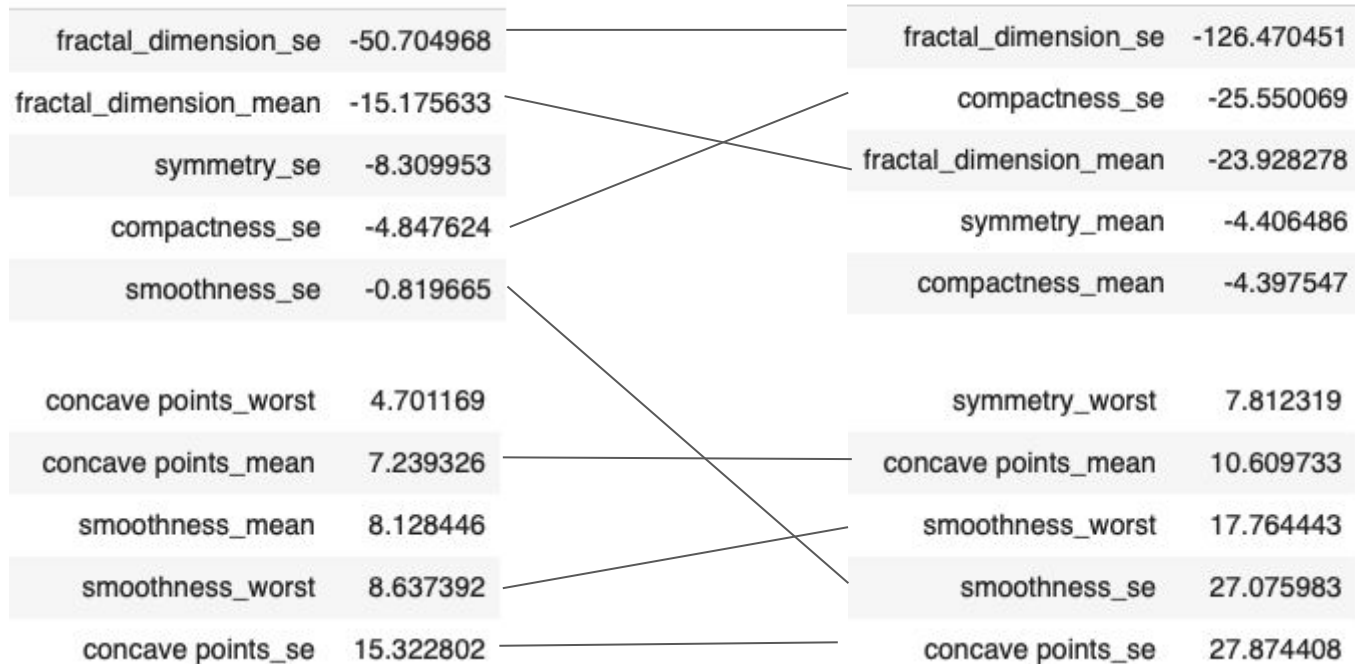
# Best Models Bottom/Top Coefficients

## Logistic Regression

column	weight
fractal_dimension_se	-50.704968
fractal_dimension_mean	-15.175633
symmetry_se	-8.309953
compactness_se	-4.847624
smoothness_se	-0.819665
concave points_worst	4.701169
concave points_mean	7.239326
smoothness_mean	8.128446
smoothness_worst	8.637392
concave points_se	15.322802

## LinearSVC

column	weight
fractal_dimension_se	-126.470451
compactness_se	-25.550069
fractal_dimension_mean	-23.928278
symmetry_mean	-4.406486
compactness_mean	-4.397547
symmetry_worst	7.812319
concave points_mean	10.609733
smoothness_worst	17.764443
smoothness_se	27.075983
concave points_se	27.874408



# Best Models Quality

	Linear Regression	LinearSVC
Runtime	4.2 seconds	28.3 seconds
AUC	99.7%	99.6%
Accuracy	98%	98%
Precision	96%	98%
Recall	100%	99%
F-measure	98%	99%
False Positives	4	2
False Negatives	0	1

Thanks!