

# Classifying Thyroid Cancer Tumors

# Motivation

- By predicting tumor classification of Thyroid disease based off of clinical and health factors, we can predict risk and target groups at risk for earlier detection and treatment.

# Data

- Our target variable, T, represents the classification of the tumor
  - 6 categories: T1a, T1b, T2, T3a, T3b, T4a, T4b
- All input variables were categorical, so we did one-hot encoding (got dummies)
- Variables N, M, Stage, Response, and Recurred were based off of T, the target variable, so we dropped them to focus on prediction of tumor classification without knowledge of the treatment afterwards

# Models used

- Methods that we chose:
  - Ruwan: K-nn, Voting
  - Taran: Logistic regression, random forest
  - Alice: SVM, Gradient Boosting
- Hyperparameter tuning:
  - We performed GridSearch to get optimal hyperparameters

# Performance Metrics Specific to Problem

- Precision: Of the tumors classified in this category, how many were classified correctly
- Recall: How many of the tumors that were actually this classification were classified in this category
- Accuracy: Out of all tumors, how many were classified correctly

Mapping of numerical labels to original 'T' labels:

0 is for T1a

1 is for T1b

2 is for T2

3 is for T3a

4 is for T3b

5 is for T4a

6 is for T4b

## Results

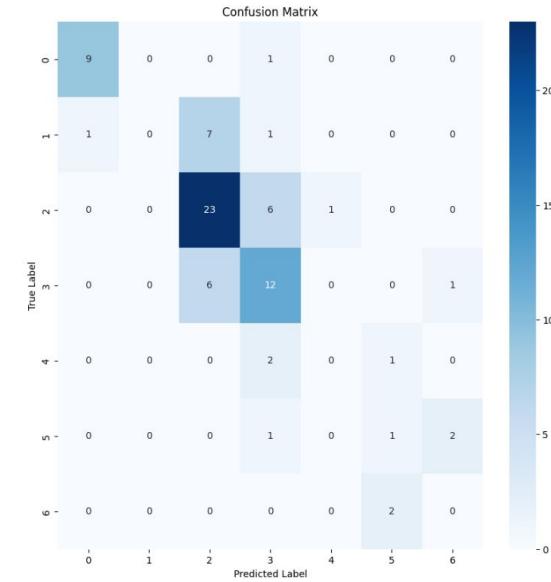
# Support Vector Machines (SVM)

Best Hyperparameters: {'C': 1, 'gamma': 1, 'kernel': 'linear'}

Accuracy: 0.5844155844155844

AUROC: 0.8244606088642603

	precision	recall	f1-score	support
0	0.90	0.90	0.90	10
1	0.00	0.00	0.00	9
2	0.64	0.77	0.70	30
3	0.52	0.63	0.57	19
4	0.00	0.00	0.00	3
5	0.25	0.25	0.25	4
6	0.00	0.00	0.00	2
accuracy			0.58	77
macro avg	0.33	0.36	0.35	77
weighted avg	0.51	0.58	0.54	77



- Linear kernel: so classes are roughly linearly separable in feature space

## Results

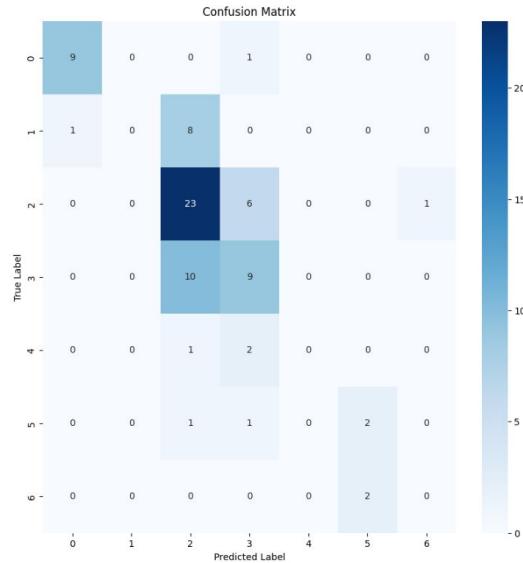
# Gradient Boosting

Best Hyperparameters: {'learning\_rate': 0.01, 'max\_depth': 3, 'n\_estimators': 100}

Accuracy: 0.5584415584415584

AUROC: 0.8219494133966787

	precision	recall	f1-score	support
0	0.90	0.90	0.90	10
1	0.00	0.00	0.00	9
2	0.53	0.77	0.63	30
3	0.47	0.47	0.47	19
4	0.00	0.00	0.00	3
5	0.50	0.50	0.50	4
6	0.00	0.00	0.00	2
accuracy			0.56	77
macro avg	0.34	0.38	0.36	77
weighted avg	0.47	0.56	0.51	77



## Results

# K-Nearest Neighbors

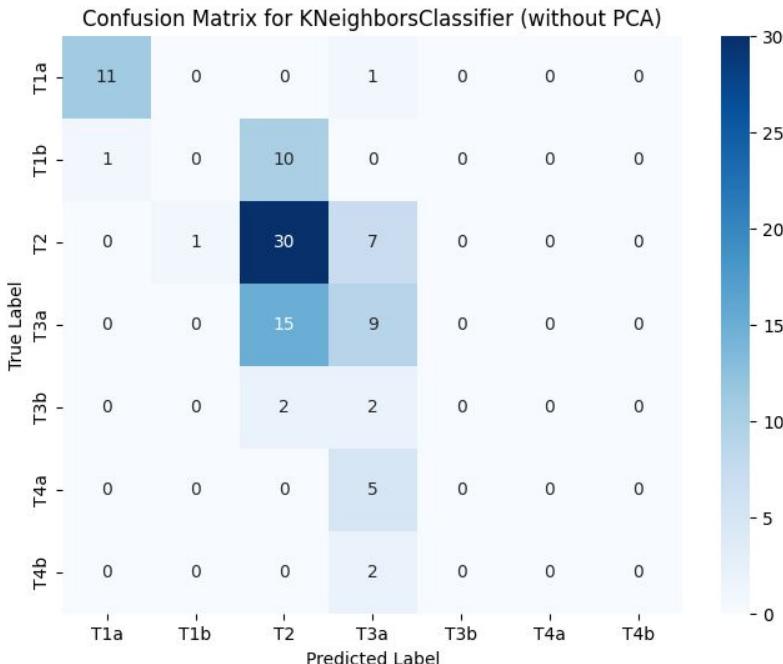
Best Hyperparameters: {'n\_neighbors': 19, 'p': 1, 'weights': 'uniform'}

Accuracy: 0.5208333333333334

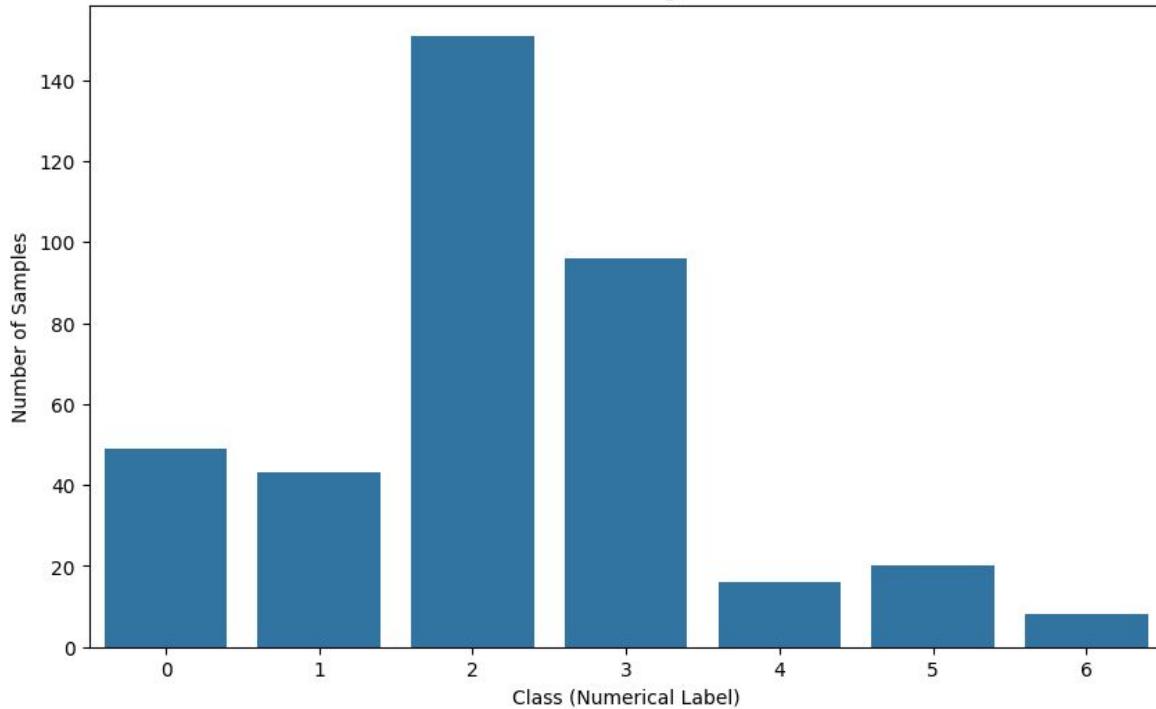
AUROC (OvR): 0.8374711773097282

	precision	recall	f1-score	support
0	0.92	0.92	0.92	12
1	0.00	0.00	0.00	11
2	0.53	0.79	0.63	38
3	0.35	0.38	0.36	24
4	0.00	0.00	0.00	4
5	0.00	0.00	0.00	5
6	0.00	0.00	0.00	2
accuracy			0.52	96
macro avg	0.26	0.30	0.27	96
weighted avg	0.41	0.52	0.45	96

- Note: Poor precision and recall for labels 1, 4, 5, and 6



Distribution of Target Classes



- Class imbalance
- Possible explanation for poor precision and recall for labels 4, 5, and 6
- Fewer instances of 4, 5, 6 because later stages less common and maybe more complex (?)

## Results

# Voting

Best Hyperparameters for Voting Classifier: {'dt\_criterion': 'dt\_criterion': 'entropy', 'dt\_max\_depth': 3, 'dt\_min\_samples\_split': 2, 'knn\_n\_neighbors': 3, 'knn\_p': 1, 'knn\_weights': 'uniform', 'voting': 'soft', 'weights': [1, 2]}

Accuracy: 0.6354166666666666

Voting Classifier Report:

	precision	recall	f1-score	support
0	0.92	0.92	0.92	12
1	0.00	0.00	0.00	11
2	0.67	0.76	0.72	38
3	0.50	0.71	0.59	24
4	0.00	0.00	0.00	4
5	0.50	0.60	0.55	5
6	1.00	0.50	0.67	2
accuracy			0.64	96
macro avg	0.51	0.50	0.49	96
weighted avg	0.55	0.64	0.59	96

Best Hyperparameters:

'dt\_criterion': 'entropy',  
'dt\_max\_depth': 3,  
'dt\_min\_samples\_split': 2,  
'knn\_n\_neighbors': 3,  
'knn\_p': 1,  
'knn\_weights': 'uniform',  
'voting': 'soft',  
'weights': [1, 2]

- Used Knn and decision trees as base models
- greater weight given to decision trees -> confident that decision tree reason for improvement
- Poor precision and recall still; Class imbalance problem persists

## Results

# Logistic Regression

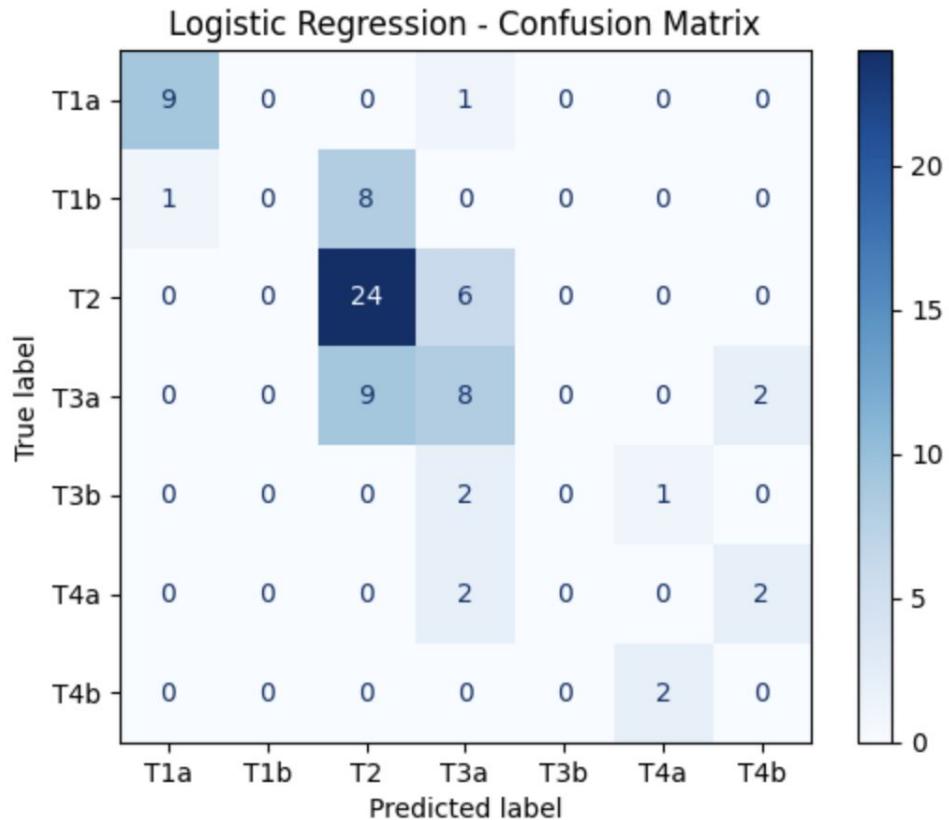
Accuracy 0.5324675324675324

Classification Report

	precision	recall	f1-score	support
0	0.90	0.90	0.90	10
1	0.00	0.00	0.00	9
2	0.59	0.80	0.68	30
3	0.42	0.42	0.42	19
4	0.00	0.00	0.00	3
5	0.00	0.00	0.00	4
6	0.00	0.00	0.00	2
accuracy			0.53	77
macro avg	0.27	0.30	0.29	77
weighted avg	0.45	0.53	0.48	77

## Results

# Logistic Regression



## Results

# Random Forest

```
Grid Search Best Params {'max_depth': None, 'max_features': 'sqrt', 'min_samples_leaf': 2, 'min_samples_split': 5, 'n_estimators': 100}
Grid Search Best Score 0.6565309360126917
```

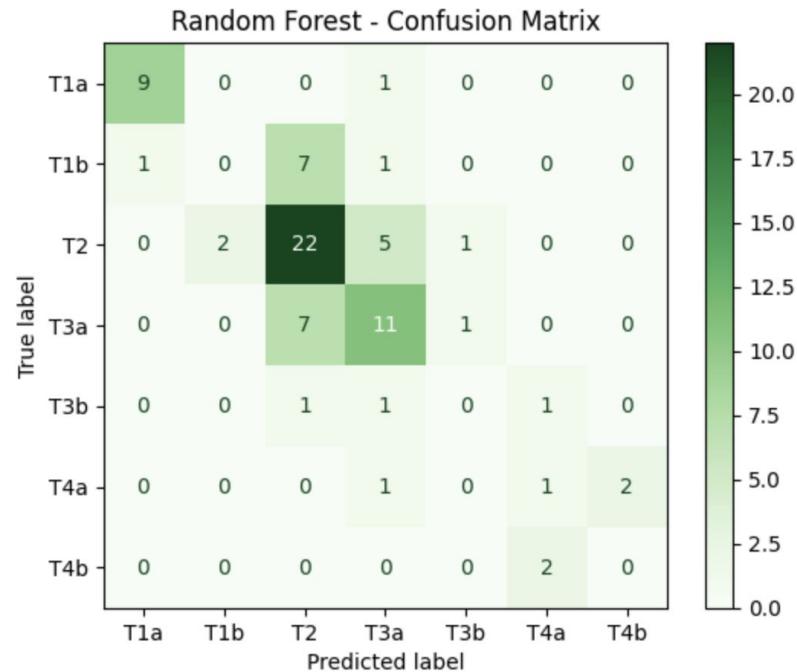
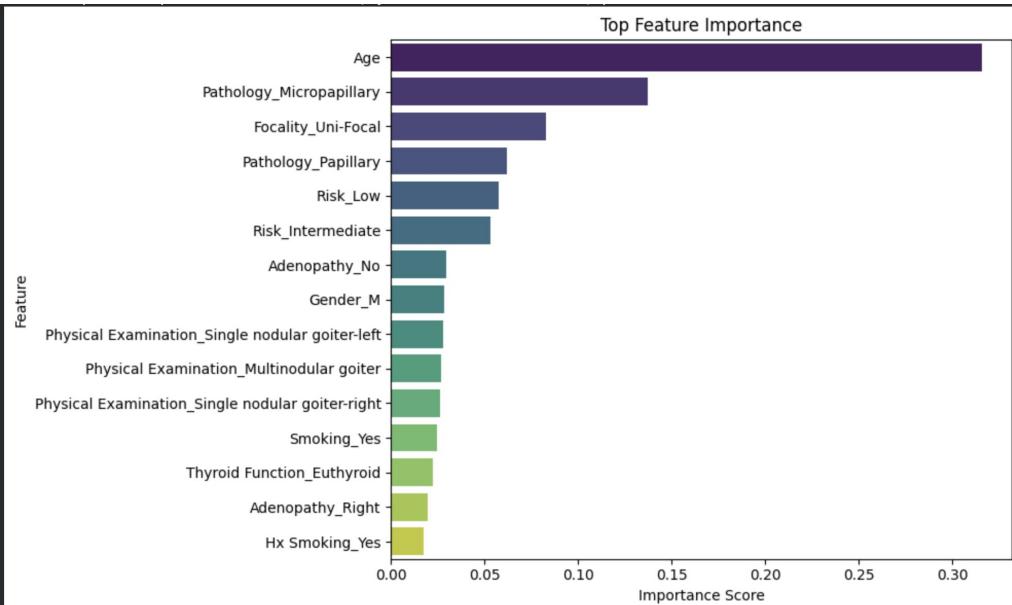
Accuracy: 0.6233766233766234

Classification Report

	precision	recall	f1-score	support
0	0.90	0.90	0.90	10
1	0.00	0.00	0.00	9
2	0.62	0.83	0.71	30
3	0.55	0.63	0.59	19
4	0.00	0.00	0.00	3
5	0.50	0.50	0.50	4
6	0.00	0.00	0.00	2
accuracy			0.62	77
macro avg	0.37	0.41	0.39	77
weighted avg	0.52	0.62	0.57	77

## Results

# Random Forest



# Next Steps / Limitations

- Address class imbalance
  - Would try to get more data samples, e.g., using the Synthetic Minority Over-Sampling Technique (SMOTE)
- Experiment with other base models for ensemble models