

# **Contract Designation Prediction based on Monitoring Student Progress Report (MSPR)**



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# Monitoring Student Progress Report (MSPR)



- Administered by NCF since Spring 2018
- Instructors fill out a report, whether they have any concerns in regards to Student's Academic Performance
- 6 unique binary concerns (1 or 0): *Attendance, Low Participation, Late/Missing Assignments, Other Assignments Concerns, Low Test Scores, and Danger of Unsat*

# Objective

Predict student's **Contract Designation** based solely on **MSPR Submissions** to facilitate effective prioritization of students in need of academic help

# Data

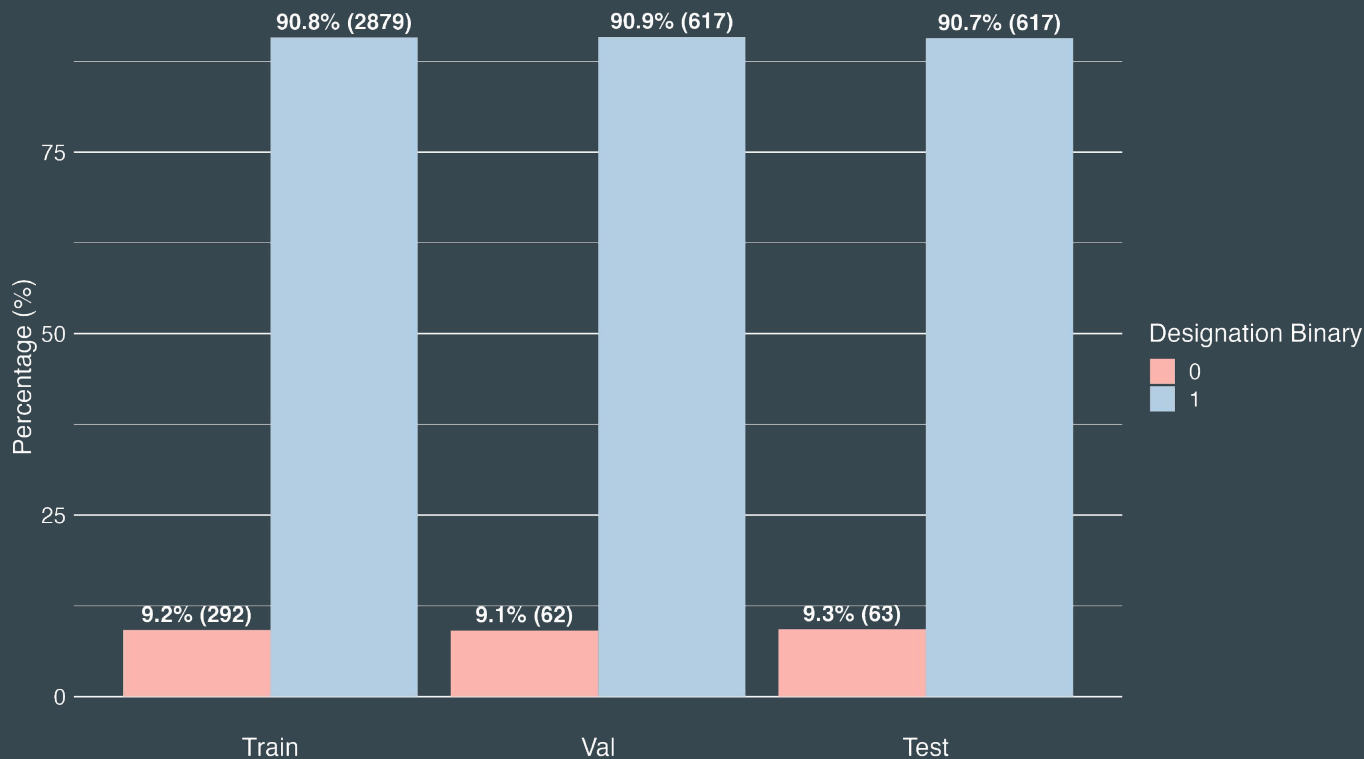
Contract Designation

Unique MSPR forms

Aggregate Concerns by Student, by Term

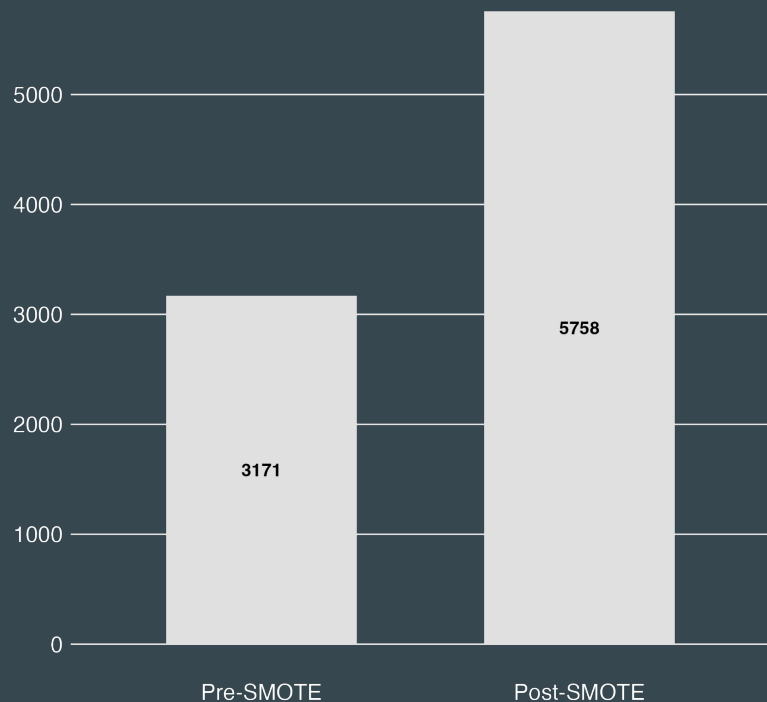
<row>	ATTENDANCE	LOW PARTICIPATION	LATE/MISSING ASSIGNMENTS	OTHER ASSIGNMENTS CONCERNS	LOW TEST SCORES	DANGER of UNSATING	COURSE COUNT	DESIGNATION BINARY
1	0	0	2	0	0	1	3	1
2	0	0	0	0	0	0	1	1
3	1	1	1	2	0	0	4	1
...	...	...	...	...	...	...	...	...
4530	0	0	0	0	0	0	1	1

# Stratified Splitting - same label distributions between splits

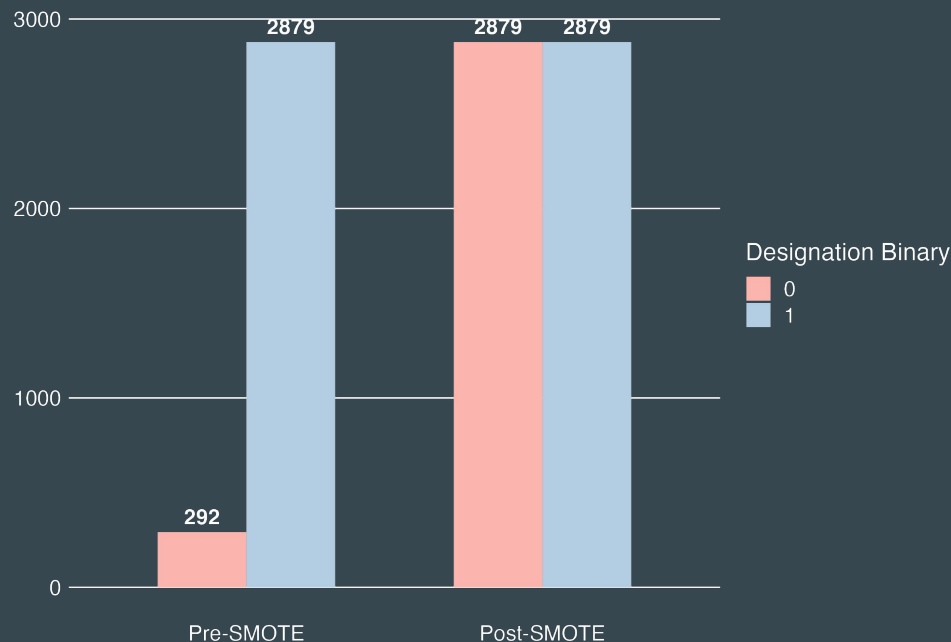


# SMOTE - balance the label classes

X\_train & y\_train Row Count



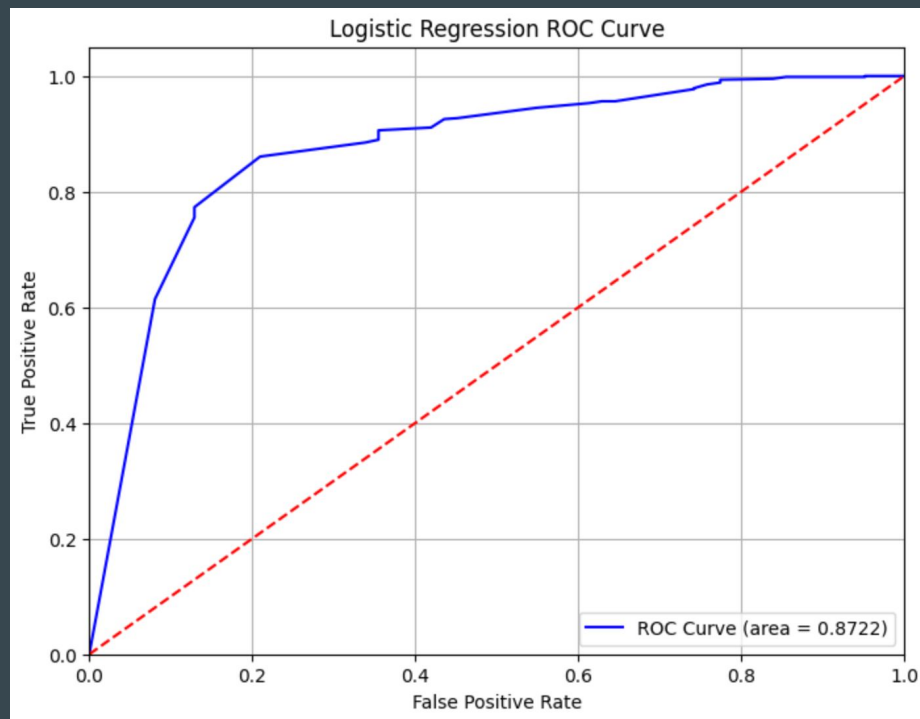
Designation Binary Count



# Baseline Model (Logistic)

Performance Metric	Validation	Test
Accuracy	0.8645	0.8618
Precision	0.9630	0.9612
Recall	0.8849	0.8833
F1-score	0.9223	0.9206
ROC AUC	0.8722	0.8390

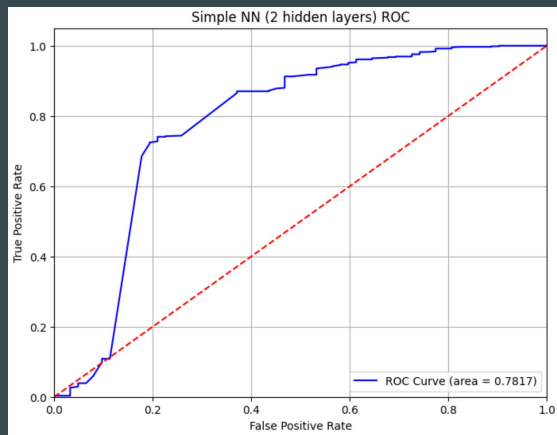
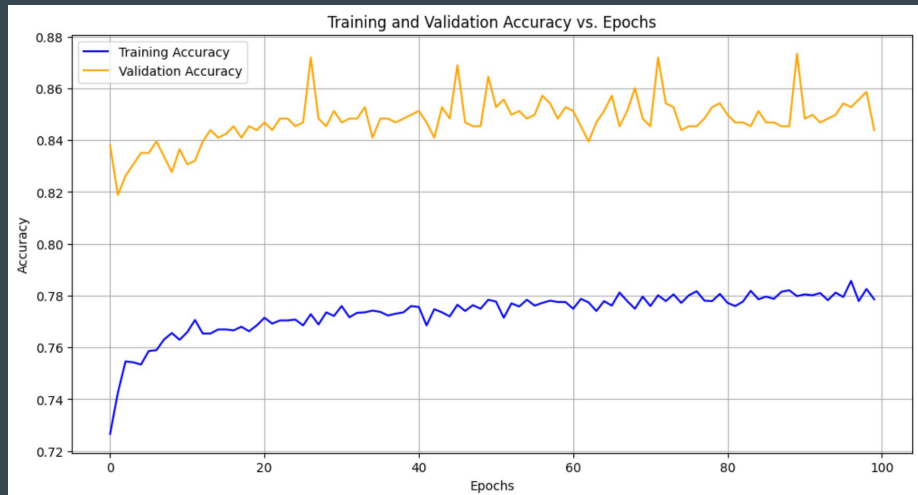
Best hyperparameters:  
{C: 0.001, penalty: l1, solver: saga}



# Simple NN(2 hidden layers)

Performance Metric	Validation	Test
Accuracy	0.8439	0.8412
Precision	0.9522	0.9504
Recall	0.8720	0.8703
F1-score	0.9103	0.9086
ROC AUC	0.7817	0.7528

Best hyperparameters:  
{learning rate: 0.01, dropout: 0.2,  
layer1: 64, layer2: 8}





# Simple NN (2 hidden layers) architecture

```
model = keras.Sequential([
    keras.layers.Dense(64, activation='relu', input_shape=(X_train_resampled.shape[1],)),
    keras.layers.Dropout(0.2),
    keras.layers.Dense(8, activation='relu'),
    keras.layers.Dropout(0.2),
    keras.layers.Dense(1, activation='sigmoid')
])

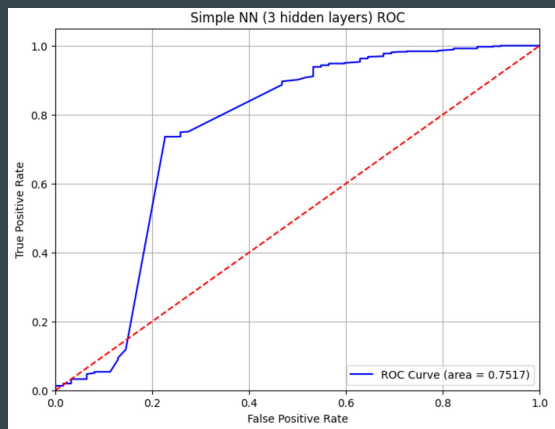
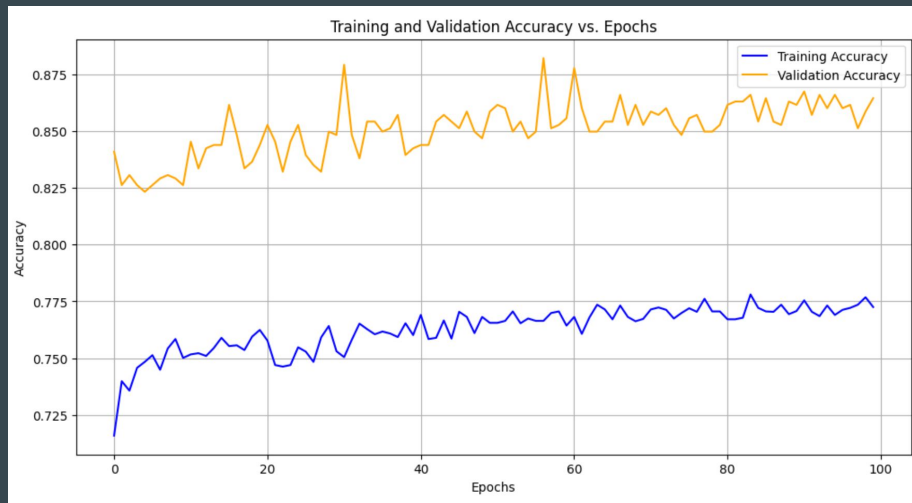
model.compile(optimizer=keras.optimizers.Adam(learning_rate=0.01),
              loss='binary_crossentropy', metrics=['accuracy'])

model.fit(X_train_resampled, y_train_resampled, epochs=100, batch_size=64, validation_data=(X_val_scaled, y_val))
```

# Simple NN(3 hidden layers)

Performance Metric	Validation	Test
Accuracy	0.8645	0.8559
Precision	0.9472	0.9482
Recall	0.9011	0.8898
F1-score	0.9236	0.9181
ROC AUC	0.7517	0.7559

Best hyperparameters:  
{learning rate: 0.01, dropout: 0.2,  
layer1: 128, layer2: 64, layer3 = 4}



# Simple NN (3 hidden layers) architecture

```
model = keras.Sequential([
    keras.layers.Dense(128, activation='relu', input_shape=(X_train_resampled.shape[1],)),
    keras.layers.Dropout(0.2),
    keras.layers.Dense(64, activation='relu'),
    keras.layers.Dropout(0.2),
    keras.layers.Dense(4, activation='relu'),
    keras.layers.Dropout(0.2),
    keras.layers.Dense(1, activation='sigmoid')
])

model.compile(optimizer=keras.optimizers.Adam(learning_rate=0.01),
              loss='binary_crossentropy', metrics=['accuracy'])

model.fit(X_train_resampled, y_train_resampled, epochs=100, batch_size=64, validation_data=(X_val_scaled, y_val))
```

# Performance Comparison

	Test Data Set		
Performance Metric	Logistic Regression	Simple NN (2 layers)	Simple NN (3 layers)
Accuracy	0.8618	0.8412	0.8559
Precision	0.9612	0.9504	0.9482
Recall	0.8833	0.8703	0.8898
F1	0.9206	0.9086	0.9181
ROC AUC	0.8390	0.7528	0.7559

# Potential Improvements

- More features for the dataset
- More epochs for NN models + early stopping
- Use random seeds to lock down the models
- Don't split into train-val-test, only train-val?



Thank You

