

# Zara F. Amer

## Worked

1. Filling NA's with 0's -> GINI: 28.7%
2. XGBoost Classifier -> Gini: 37%
3. XGB with SMOTE -> Gini: 41%
4. 10 Generation Genetic Algorithm -> Gini: 82%
5. Balancing hyperparameters using Random Forest Classifier -> Gini: 73%
  - Bootstrap: false
  - Max\_depth: 84
  - Max\_features: sqrt
  - Min\_samples\_leaf: 1
  - Min\_samples\_split: 2
  - N\_estimators: 255

## Did not work

1. Logistic Regression
2. Applying Standard Scaling

## Wish List

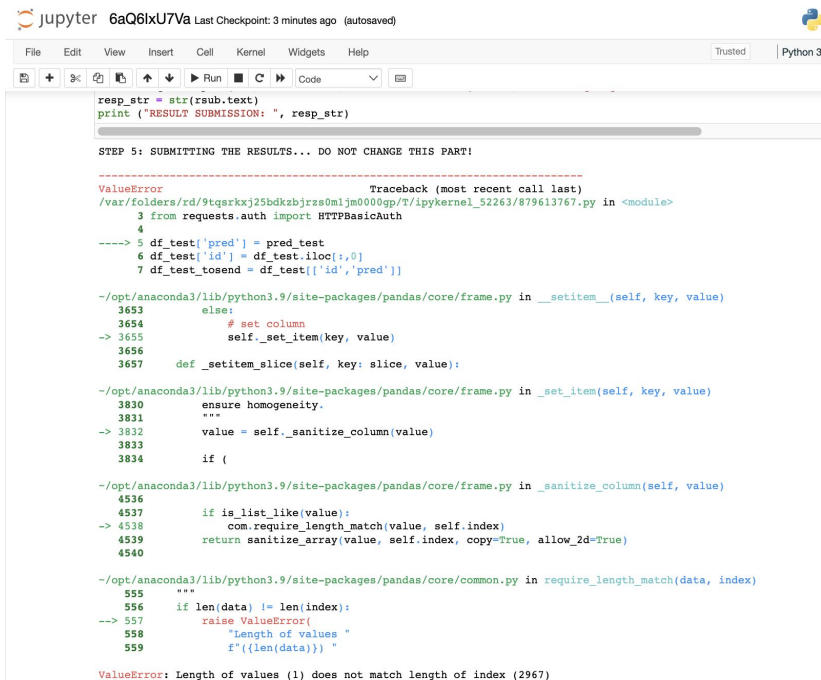
(things you wish to have tested but didn't have time)

1. Bagging
2. Filling NA's with Mean, Median
3. Chi-Squared
4. Support Vector Machines (SVM)
5. Neural Networks
6. Isolation Forests
7. K-nearest neighbor (KNN)

# GENETIC ALGORITHM

```
Generation 1/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 2/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 3/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 4/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 5/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 6/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 7/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 8/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 9/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Generation 10/10
Best Individual: [0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1]
Fitness Score: 1.0
Gini Score on Test Data: 0.8233293317326931
```

I tried a Genetic Algorithm but could not figure out how to upload/submit the result to the challenge. It gave me a Value Error in Part 5 (see below). However, the code can be found in the notebook and the picture on the left shows the result of the algorithm in Part 3.



The screenshot shows a Jupyter Notebook interface with the following details:

- Header: jupyter 6aQ6ixU7Va Last Checkpoint: 3 minutes ago (autosaved)
- Menu: File, Edit, View, Insert, Cell, Kernel, Widgets, Help
- Buttons: Run, Code, Trusted, Python 3
- Code Cell Content:

```
resp_str = str(rsub.text)
print('RESULT SUBMISSION: ', resp_str)

STEP 5: SUBMITTING THE RESULTS... DO NOT CHANGE THIS PART!

-----
ValueError                                Traceback (most recent call last)
/var/folders/rd/9tqsrkxj25bdkzbjrz0mljm0000gp/T/ipykernel_52263/879613767.py in <module>
      3 from requests.auth import HTTPBasicAuth
      4
----> 5 df_test['pred'] = pred_test
      6 df_test['id'] = df_test.iloc[:,0]
      7 df_test.to_send = df_test[['id','pred']]

~/opt/anaconda3/lib/python3.9/site-packages/pandas/core/frame.py in __setitem__(self, key, value)
   3653     else:
   3654         # set column
-> 3655         self._setitem(key, value)
   3656
   3657     def _setitem_slice(self, key: slice, value):

~/opt/anaconda3/lib/python3.9/site-packages/pandas/core/frame.py in _setitem(self, key, value)
   3800     ensure_homogeneity.
   3801     """
-> 3802     value = self._sanitize_column(value)
   3803
   3804     if (

~/opt/anaconda3/lib/python3.9/site-packages/pandas/core/frame.py in _sanitize_column(self, value)
   4536
   4537     if is_list_like(value):
-> 4538         com.require_length_match(value, self.index)
   4539     return sanitize_array(value, self.index, copy=True, allow_2d=True)
   4540

~/opt/anaconda3/lib/python3.9/site-packages/pandas/core/common.py in require_length_match(data, index)
   555     """
   556     if len(data) != len(index):
-> 557         raise ValueError(
   558             "Length of values "
   559             f"({len(data)}) "
ValueError: Length of values (1) does not match length of index (2967)
```

```
# Print the best hyperparameters
print("Best Hyperparameters:")
for param, value in best_params.items():
    print(f'{param}: {value}')

# Fit the best model on the entire dataset
best_model.fit(X, y)

# Now you can use the best_model for making predictions on new data
```

STEP 3: DEVELOPING THE MODEL...

```
Best Hyperparameters:
bootstrap: False
max_depth: 84
max_features: sqrt
min_samples_leaf: 1
min_samples_split: 2
n_estimators: 255
```

```
Out[6]: RandomForestClassifier(bootstrap=False, max_depth=84, max_features='sqrt',
                               n_estimators=255, random_state=42)
```

WHAT IS GINI?

- watch this video for reference: <https://youtu.be/MIBUVUC8kE>

In [14]: `import numpy as np`

```
def gini_coefficient(y_true, y_pred):
    # Sort the true values and predicted values in descending order
    sorted_indices = np.argsort(y_pred)[::-1]
    sorted_true = y_true[sorted_indices]
    sorted_pred = y_pred[sorted_indices]

    # Calculate the cumulative sum of true values
    cum_true = np.cumsum(sorted_true)

    # Calculate the cumulative sum of predicted values
    cum_pred = np.cumsum(sorted_pred)

    # Calculate the Lorenz curve values
    lorenz_curve_true = cum_true / np.sum(sorted_true)
    lorenz_curve_pred = cum_pred / np.sum(sorted_pred)

    # Calculate the Gini coefficient
    gini_coeff = np.sum((lorenz_curve_pred[1:-1] + lorenz_curve_pred[1:]) * (lorenz_curve_true[1:] - lorenz_curve_true[0]))

    return gini_coeff
```

```
# Example usage:
y_true = np.array([0, 1, 0, 1, 0, 1]) # True values
y_pred = np.array([0.2, 0.8, 0.4, 0.6, 0.1, 0.9]) # Predicted values

gini_score = gini_coefficient(y_true, y_pred)
print("Gini Coefficient:", gini_score)
```

Gini Coefficient: 0.7333333333333334

In [15]: `from sklearn.metrics import roc_auc_score`

```
# Example usage:
y_true = [0, 1, 0, 1, 0, 1] # True values
y_pred = [0.2, 0.8, 0.4, 0.6, 0.1, 0.9] # Predicted values

roc_auc = roc_auc_score(y_true, y_pred)
print("ROC AUC Score:", roc_auc)
```

ROC AUC Score: 1.0

Random Forest Classifier with Hyperparameters as follows:

- Bootstrap: false
- Max\_depth: 84
- Max\_features: sqrt
- Min\_samples\_leaf: 1
- Min\_samples\_split: 2
- N\_estimators: 255

However, I was unable to submit the result in Part 5 and got the same Value Error as in the previous slide on Genetic Algorithms.

# Main steps in Machine Learning (role 3)

## (A) Sampling and Performance definition

- Fraud data is often skewed
- Cross Validation is important to ensure all scenarios are covered by the machine learning model
- Overfitting is not a major issue in Fraud Detection
- In Fraud, 0s are independent and 1s are dependent
- Recall is more important than precision

## (B) Feature Engineering

- Filling missing values by imputing or creating separate indicator variables
- Using one-hot encoding for categorical variables
- Scaling numerical features to ensure comparable ranges and prevent bias
- Transforming existing features to create ratios
- Creating time-based feature to understand temporal patterns

## (C) Preprocessing

- Train-Test split
- Handling Skewed Data (balancing using SMOTE)
- Feature scaling for numerical and categorical variables
- Filling missing values

## (D) Model Fit & Feature Selection

- Hyperparameter tuning is essential
- Using various algorithms to figure out feature importance
- Methods: Random Forest / XGB
- Univariate Analysis to understand relationship with target variable

## (E) Model Evaluation

- Accuracy Score
- ROC curve and Gini Score
- Confusion Matrix
- Hit score
- F1 score

## (F) Cross Validation

- Hyperparameter tuning
- Using Train & Test set
- Using model evaluation metrics to evaluate the model
- K-fold-cross validation

### (D.2) Hyperparameter (normally wrongly skipped)

- Identifying best hyperparameters using param grid and using it with algorithms like Random Forest Classifiers, XG Boost, etc is one of the best ways to find the best model