# CS253 Assignment 3

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#### 1 Methodology

This document outlines the process of training a machine learning model to predict the education level of individuals using a RandomForestClassifier. The dataset includes various features such as party affiliation, criminal cases, total assets, and liabilities. For more detailed information about RandomForestClassifier, see the official documentation here.

#### 1.1 Imports

The Python script utilizes several libraries:

- pandas for data manipulation.
- scikit-learn for machine learning tools.

#### 1.2 Numeric Conversion Function

```
def convert_to_numeric(value):
    value = str(value)
    if 'Crore+' in value:
        return float(value.replace('Crore+', '')) * 10**7
    elif 'Lac+' in value:
        return float(value.replace('Lac+', '')) * 10**5
    else:
        return pd.to_numeric(value, errors='coerce')
```

This function converts string representations of numbers into floats, handling 'Crore+' and 'Lac+' suffixes appropriately.

#### 1.3 Data Loading

The datasets are loaded from CSV files using:

```
train_df = pd.read_csv('path_to_train.csv')
test_df = pd.read_csv('path_to_test.csv')
```

#### 1.4 Data Preprocessing

The script includes conversion of certain columns to numeric types, filling missing values, and encoding categorical features.

#### 1.5 Feature and Label Split

Features are isolated in X and the target variable 'Education' in y.

#### 1.6 Hyperparameter Tuning

Using GridSearchCV from scikit-learn, the script performs hyperparameter tuning to find the optimal settings for the RandomForestClassifier. The parameters tested include:

- n\_estimators: Number of trees in the forest. Tested values: 100, 200, 300. More trees can lead to better performance but increase computational cost. Best Parameter: 200
- max\_depth: Maximum number of levels in each tree. Tested values: None (unlimited), 10, 20. Limiting depth can reduce overfitting. Best Parameter: None
- min\_samples\_split: Minimum number of samples required to split an internal node. Tested values: 2, 5, 10. Higher values prevent the model from learning overly specific patterns, reducing overfitting. Best Parameter: 2
- min\_samples\_leaf: Minimum number of samples required to be at a leaf node. Tested values: 1, 2, 4. Like min\_samples\_split, higher values increase the robustness of the model. Best Parameter: 1

#### 1.7 Model Training

```
model = RandomForestClassifier(**best_params, random_state=42)
model.fit(X, y)
```

The model is trained using the best parameters found from the grid search.

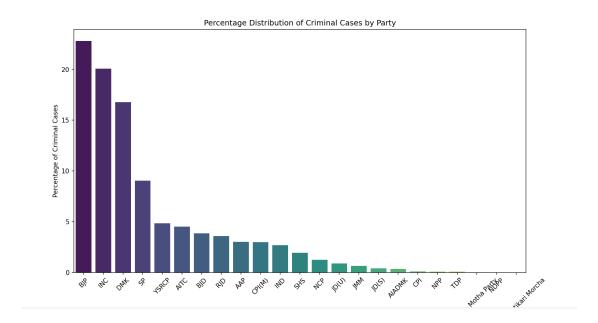
#### 1.8 Prediction and Submission

```
y_pred = model.predict(X1)
test_df['Education'] = le.inverse_transform(y_pred)
test_df[['ID', 'Education']].to_csv('submission_7.csv', index=False)
```

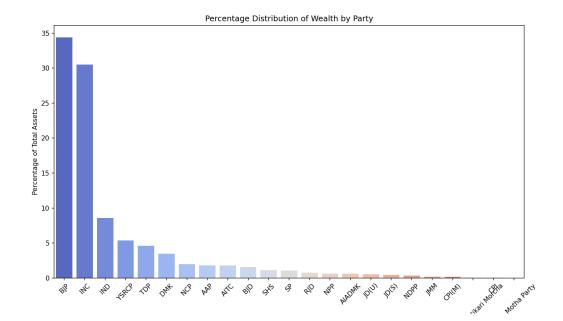
Predictions are made on the test set, transformed back to original labels, and saved for submission.

# 2 Data Insights

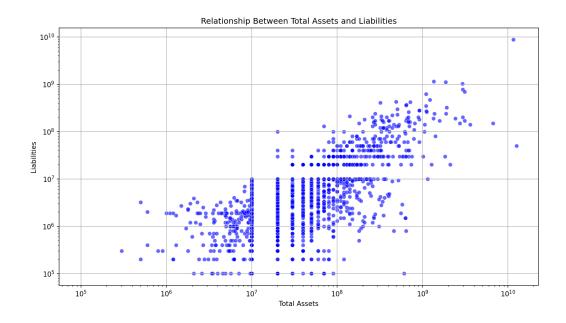
2.1 Percentage distribution of parties with candidates having the most criminal records



# 2.2 Percentage distribution of parties with the most wealthy candidates



#### 2.3 Bonus Plot: Total Assets vs Liabilities



## 3 Results

RANK: 215 (Many deleted submission, this rank in including those also)

SCORE : 0.20071

#### 4 Github Link

https://github.com/sakshamm21/CS253-Assignment-3.git

## 5 References