

## Model Development Phase Template

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| Date          | 10 July 2024  |
| Team ID       | 865503  |
| Project Title | Frappe Activity:Mobile Phone Activity<br>Classification |
| Maximum Marks | 4 Marks   |

### Initial Model Training Code, Model Validation and Evaluation Report

This initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include classification reports, accuracy, and confusion matrices for multiple models, presented through respective screenshots.

### Model Training Code :

```
# Feature Selection
features = ['item', 'daytime', 'weekday', 'isweekend', 'cost', 'weather', 'name']
X = df[features]
y = df[["homework"]]
```

```
print(df['name'])
```

```
0      260
1     3217
2       698
3     1489
4     2523
...
96198    260
96199   1037
96200   1994
96201   1307
96202   2429
Name: name, Length: 96203, dtype: int32
```

```
from imblearn.over_sampling import SMOTE
```

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```

```
from imblearn.over_sampling import SMOTE
```

```
df['homework'].value_counts()
```

```
homework
1    75670
0    15771
2     4762
Name: count, dtype: int64
```

```
smote_sampler=SMOTE(random_state=42)
X_smote,y_smote = smote_sampler.fit_resample(X,y)
smote_data = pd.concat([X_smote,y_smote],axis=1)
smote_data.shape
```

```
(227010, 8)
```

```
smote_data['homework'].value_counts()
```

```
homework
1    75670
0    75670
2    75670
Name: count, dtype: int64
```

```
X_train, X_test, y_train, y_test = train_test_split(X_smote, y_smote, test_size=0.8, random_state=42)
```

```
# Scaling Data so that all the features has similar weight.
```

```
scaler = StandardScaler()  
scaler.fit(X_train)  
X_train = scaler.transform(X_train)  
X_test = scaler.transform(X_test)
```

```
from sklearn.neighbors import KNeighborsClassifier  
from sklearn.ensemble import RandomForestClassifier  
from sklearn.tree import DecisionTreeClassifier  
from xgboost import XGBClassifier  
from sklearn.ensemble import BaggingClassifier  
from sklearn.ensemble import AdaBoostClassifier
```

```
models=[]  
models = [('AdaBoost', AdaBoostClassifier(algorithm='SAMME'))]  
models.append(('KNeighborsClassifier',KNeighborsClassifier()))  
models.append(('DecisionTreeClassifier',DecisionTreeClassifier()))  
models.append(('RandomForestClassifier',RandomForestClassifier()))  
models.append(('XGBClassifier',XGBClassifier()))  
models.append(('BaggingClassifier',BaggingClassifier()))
```

```
from sklearn.metrics import precision_score, recall_score, accuracy_score, f1_score
```

```
model_perform = {}  
|  
for name, model in models:  
    # Fit the model  
    model.fit(X_train, y_train.values.ravel())  
  
    # Make predictions  
    y_pred = model.predict(X_test)  
  
    # Calculate performance metrics  
    p = precision_score(y_test, y_pred, average='micro')  
    r = recall_score(y_test, y_pred, average='micro')  
    a = accuracy_score(y_test, y_pred)  
    f = f1_score(y_test, y_pred, average='micro')  
  
    # Store the performance metrics  
    s = {'Precision': p, 'Recall': r, 'Accuracy': a, 'F1 Score': f}  
    model_perform[name] = s
```

```
for model in model_perform:  
    print(model)  
    print("Precision: ",model_perform[model]['Precision'])  
    print("Recall: ",model_perform[model]['Recall'])  
    print("Accuracy: ",model_perform[model]['Accuracy'])  
    print("F1 Score: ",model_perform[model]['F1 Score'])  
    print()
```

## Model Valuation And Evalution Report

| Model                 | Classification Report  | F1 Score |
|-----------------------|--|----------|
| Bagging Classifier    | <pre> BaggingClassifier Precision: 0.635522664199815 Recall: 0.635522664199815 Accuracy: 0.635522664199815 F1 Score: 0.635522664199815 </pre>          | 63%      |
| Random Forest         | <pre> RandomForestClassifier Precision: 0.6382538214175587 Recall: 0.6382538214175587 Accuracy: 0.6382538214175587 F1 Score: 0.6382538214175587 </pre> | 63%      |
| Decision Tree         | <pre> DecisionTreeClassifier Precision: 0.6193945200651954 Recall: 0.6193945200651954 Accuracy: 0.6193945200651954 F1 Score: 0.6193945200651954 </pre> | 61%      |
| KNeighbors Classifier | <pre> KNeighborsClassifier Precision: 0.5377571472622351 Recall: 0.5377571472622351 Accuracy: 0.5377571472622351 F1 Score: 0.5377571472622351 </pre>   | 53%      |
| XGB Classifier        | <pre> XGBClassifier Precision: 0.64430531694639 Recall: 0.64430531694639 Accuracy: 0.64430531694639 F1 Score: 0.64430531694639 </pre>                  | 64%      |

|          |   |     |
|----------|---|-----|
| AdaBoost | AdaBoost<br>Precision: 0.4572430729923792<br>Recall: 0.4572430729923792<br>Accuracy: 0.4572430729923792<br>F1 Score: 0.4572430729923792 | 45% |
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