

## Model Optimization and Tuning Phase Template

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Project Title	Frappe Activity: mobile Phone Activity classification
Maximum Marks	10 Marks

### Model Optimization and Tuning Phase

The Model Optimization and Tuning Phase involves refining neural network models for peak performance. It includes optimized model code, fine-tuning hyperparameters, comparing performance metrics, and justifying the final model selection for enhanced predictive accuracy and efficiency.

#### Hyperparameter Tuning Documentation (8 Marks):

Model	Tuned Hyperparameters

## Bagging Classifier

The (params) define a grid for hyperparameter tuning of the Bagging Classifier (BClassifier), including min\_child\_weight, gamma, colsample\_bytree, and max\_depth. The Bagging Classifier is configured with a learning rate of 0.5, 100 estimators, using a binary logistic regression objective, and utilizing 3 threads for processing. GridSearchCV (xg\_cv) is used with 5-fold cross-validation (cv=5), refitting the best model

```
# Define the hyperparameters and their possible values for tuning

param_grid = {

    'n_estimators': [10, 50, 100],
    'max_samples': [0.5, 0.7, 1.0],
    'max_features': [0.5, 0.7, 1.0],
    'bootstrap': [True, False],
    'bootstrap_features': [True, False]
}

random_search = RandomizedSearchCV(estimator=bagging_classifier, param_distributions=param_grid,
| | | | | | | | | | scoring='accuracy', cv=2, random_state=42)

random_search.fit(X_train,y_train)
```

(refit=True), evaluating based on accuracy (scoring="accuracy")

```
print("Best Parameters:",random_search.best_params_)
print("Best Score:",random_search.best_score_)
[46]
... Best Parameters: {'n_estimators': 100, 'max_samples': 0.7, 'max_features': 1.0, 'bootstrap_features': True, 'bootstrap': False}
Best Score: 0.6545086119554204

print("Best Score:",random_search.score(X_test,y_test))
[47]
... Best Score: 0.6861702127659575
```

## Decision Tree

The parameters (params) define a grid for hyperparameter tuning of the Decision Tree Classifier (DecisionTreeClassifier), including max\_depth, min\_samples\_leaf, and criterion ('gini' or 'entropy'). GridSearchCV (dec\_cv) is used with 5-fold cross-validation (cv=5), evaluating model performance based on accuracy (scoring="accuracy")

```
# Define the hyperparameters and their possible values for tuning
param_grid = {
    'criterion': ['gini', 'entropy'],
    'splitter': ['best', 'random'],
    'max_depth': [None, 2, 4, 6, 8, 10],
    'min_samples_split': [2, 5, 10],
    'min_samples_leaf': [1, 2, 4],
    'max_features': [None, 'sqrt', 'log2'],
    'min_impurity_decrease': [0.0, 0.1, 0.2],
    'ccp_alpha': [0.0, 0.1, 0.2]
}
```

```
# Initialize RandomizedSearchCV with DecisionTreeClassifier
random_search = RandomizedSearchCV(estimator=dt_classifier,
                                   param_distributions=param_grid,
                                   scoring='accuracy',
                                   cv=3,
                                   n_iter=100,
                                   random_state=42)
```

```
random_search.fit(X_train, y_train)
RandomizedSearchCV(cv=3, estimator=DecisionTreeClassifier(), n_iter=100,
                  param_distributions={'ccp_alpha': [0.0, 0.1, 0.2],
                                      'criterion': ['gini', 'entropy'],
                                      'max_depth': [None, 2, 4, 6, 8, 10],
                                      'max_features': [None, 'sqrt', 'log2'],
                                      'min_impurity_decrease': [0.0, 0.1, 0.2],
                                      'min_samples_leaf': [1, 2, 4],
                                      'min_samples_split': [2, 5, 10],
                                      'splitter': ['best', 'random']},
                  random_state=42, scoring='accuracy')
```

```
print("Best Parameters:", random_search.best_params_)
```

```
print("Best Score:", random_search.best_score_)
```

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**Final Model Selection Justification (2 Marks):**

<b>Final Model</b>	<b>Reasoning</b>
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### Bagging Classifier

Bagging Classifier model is chosen for its robustness in handling complex datasets and its ability to mitigate overfitting while providing high predictive accuracy.

```
print(classification_report(y_test,y_pred,digits=4))
```

	precision	recall	f1-score	support
0	0.5500	0.6396	0.5914	60466
1	0.6052	0.4767	0.5333	60427
2	0.6683	0.7010	0.6843	60715
accuracy			0.6060	181608
macro avg	0.6078	0.6058	0.6030	181608
weighted avg	0.6079	0.6060	0.6031	181608

Above all the models Bagging classifier have the highest accuracy among all the models.