

Model Optimization and Tuning Phase Template

Date	15 March 2024
Team ID	SWTID1720437019
Project Title	Thyroid Classification
Maximum Marks	10 Marks

Model Optimization and Tuning Phase

The Model Optimization and Tuning Phase involves refining machine learning 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 (6 Marks):

Model	Tuned Hyperparameters	Optimal Values
Random Forest Model	<pre>[39]: from sklearn.ensemble import RandomForestClassifier RFclassifier = RandomForestClassifier(max_leaf_nodes=30) RFclassifier.fit(x_train, y_train)</pre> <pre>[42]: from sklearn.model_selection import GridSearchCV rf = { 'n_estimators': [100, 200, 300], 'max_depth': [3, 5, 7, 10, None] } RFclassifier = RandomForestClassifier(random_state=42) grid_rf = GridSearchCV(RFclassifier, rf, cv=5) grid_rf.fit(x_train, y_train) print("Best parameters for Random Forest:", grid_rf.best_params_) y_pred = grid_rf.predict(x_test) print(classification_report(y_test, y_pred))</pre>	<pre>[44]: RFAcc = accuracy_score(y_pred, y_test) print("Random Forest accuracy is: {:.2f}%".format(RFAcc*100)) Random Forest accuracy is: 94.20%</pre>

SVC Model	<pre>[45]: from sklearn.svm import SVC SVCclassifier = SVC(kernel='linear', max_iter=251) SVCclassifier.fit(x_train, y_train)</pre>	<pre>[50]: SVCacc = accuracy_score(y_pred, y_test) print('SVC accuracy is: {:.2f}%'.format(SVCacc*100)) SVC accuracy is: 86.61%</pre>
XGB Classifier Model	<pre>[46]: svc_params = { 'kernel': ['linear', 'poly', 'rbf', 'sigmoid'], 'C': [1, 10, 100], 'gamma': ['scale', 'auto'] } grid_svc = GridSearchCV(SVC(), svc_params, cv=5) grid_svc.fit(x_train, y_train) print("Best parameters for SVC:", grid_svc.best_params_)</pre> <pre>[52]: from xgboost import XGBClassifier from sklearn.preprocessing import LabelEncoder le = LabelEncoder() y_train_encoded = le.fit_transform(y_train) xgb = XGBClassifier() xgb.fit(x_train, y_train_encoded)</pre> <pre>[55]: from sklearn.model_selection import GridSearchCV param_grid = { 'learning_rate': [0.01, 0.1, 0.2], 'max_depth': [3, 5, 7], 'n_estimators': [100, 200, 300] } grid_search = GridSearchCV(xgb, param_grid, cv=5) grid_search.fit(x_train, y_train_encoded) print("Best parameters for XGBClassifier:", grid_search.best_params_) y_pred = grid_search.best_estimator_.predict(x_test) print(classification_report(y_test_encoded, y_pred))</pre>	<pre>[57]: XGBAcc = accuracy_score(y_test_encoded, y_pred) print('XGB accuracy is: {:.2f}%'.format(XGBAcc*100)) XGB accuracy is: 95.54%</pre>

Performance Metrics Comparison Report (2 Marks):

Model	Optimized Metric
-------	------------------

Random Forest

Model

```
[40]: y_pred = RFclassifier.predict(x_test)

print(classification_report(y_test, y_pred))
print(confusion_matrix(y_test, y_pred))
```

	precision	recall	f1-score	support
0	0.00	0.00	0.00	7
1	0.86	0.86	0.86	74
2	0.90	1.00	0.95	85
3	0.86	0.82	0.84	38
4	0.95	1.00	0.97	122
5	0.92	0.86	0.89	51
6	0.99	0.93	0.96	71
accuracy			0.92	448
macro avg	0.78	0.78	0.78	448
weighted avg	0.91	0.92	0.91	448

```
[[ 0  0  0  0  6  1  0]
 [ 0 64  5  3  1  1  0]
 [ 0  0 85  0  0  0  0]
 [ 0  4  1 31  0  1  1]
 [ 0  0  0  0 122  0  0]
 [ 0  4  1  2  0 44  0]
 [ 0  2  2  0  0  1 66]]
```

SVC Model

```
[47]: y_pred = SVCclassifier.predict(x_test)
      print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
0	0.67	0.86	0.75	7
1	0.79	0.80	0.79	74
2	0.83	0.74	0.78	85
3	0.73	0.58	0.65	38
4	0.89	0.95	0.92	122
5	0.76	0.75	0.75	51
6	0.87	0.96	0.91	71
accuracy			0.83	448
macro avg	0.79	0.80	0.79	448
weighted avg	0.83	0.83	0.83	448

```
[48]: print(confusion_matrix(y_test, y_pred))
```

```
[[ 6  0  0  0  1  0  0]
 [ 1 59  7  3  2  2  0]
 [ 1  4 63  0  9  3  5]
 [ 0  6  2 22  0  7  1]
 [ 1  1  0  0 116  0  4]
 [ 0  4  3  4  2 38  0]
 [ 0  1  1  1  0  0 68]]
```

XGB Classifier

```
Best parameters for XGBClassifier: {'learning_rate': 0.1, 'max_depth': 3, 'n_estimators': 300}
precision    recall  f1-score   support

   0         1.00    0.71    0.83         7
   1         0.93    0.95    0.94        74
   2         0.95    0.98    0.97        85
   3         0.88    0.92    0.90        38
   4         0.98    1.00    0.99       122
   5         0.92    0.90    0.91        51
   6         1.00    0.94    0.97        71

 accuracy          0.96         448
 macro avg          0.95         448
 weighted avg       0.96         448

[56]: print(confusion_matrix(y_test_encoded, y_pred))

[[ 5  0  0  0  2  0  0]
 [ 0 70  1  3  0  0  0]
 [ 0  0 83  0  0  2  0]
 [ 0  2  0 35  0  1  0]
 [ 0  0  0  0 122  0  0]
 [ 0  2  1  2  0 46  0]
 [ 0  1  2  0  0  1 67]]
```

Final Model Selection Justification (2 Marks):

Final Model	Reasoning
XGB classifier Model	<p>The XGB classifier model was selected for its superior performance, exhibiting high accuracy during hyperparameter tuning. Its ability to handle complex relationships, minimize overfitting, and optimize predictive accuracy aligns with project objectives, justifying its selection as the final model.</p>