

Model Development Phase Template

Date	11 July 2024
Team ID	SWTID1720455879
Project Title	Human Resource Management: Predicting Employee Promotions Using Machine Learning
Maximum Marks	4 Marks

Initial Model Training Code, Model Validation and Evaluation Report

The 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.

Initial Model Training Code:

```
#Importing the models from sklearn
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.neighbors import KNeighborsClassifier
from xgboost import XGBClassifier
from sklearn.model_selection import cross_val_score

# Initialize models
dt_model = DecisionTreeClassifier(random_state=42)
rf_model = RandomForestClassifier(random_state=42)
knn_model = KNeighborsClassifier()
xgb_model = XGBClassifier(random_state=42)

# Train the models
dt_model.fit(X_train, y_train)
rf_model.fit(X_train, y_train)
knn_model.fit(X_train, y_train)
xgb_model.fit(X_train, y_train)
```

```
# Evaluate each model
models = {
    'Decision Tree': dt_model,
    'Random Forest': rf_model,
    'KNN': knn_model,
    'XGBoost': xgb_model
}

for model_name, model in models.items():
    y_pred = model.predict(X_test)
    print(f"Evaluation for {model_name}:")
    print(confusion_matrix(y_test, y_pred))
    print(classification_report(y_test, y_pred))
    print("="*80)
```

Model Validation and Evaluation Report:

Model	Classification Report	Accuracy	Confusion Matrix																														
Decision Tree	<pre>print(classification_report(y_test, y_pred))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.93</td><td>0.93</td><td>0.93</td><td>15180</td></tr><tr><td>1</td><td>0.93</td><td>0.93</td><td>0.93</td><td>14904</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.93</td><td>30084</td></tr><tr><td>macro avg</td><td>0.93</td><td>0.93</td><td>0.93</td><td>30084</td></tr><tr><td>weighted avg</td><td>0.93</td><td>0.93</td><td>0.93</td><td>30084</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.93	0.93	0.93	15180	1	0.93	0.93	0.93	14904	accuracy			0.93	30084	macro avg	0.93	0.93	0.93	30084	weighted avg	0.93	0.93	0.93	30084	<pre># Decision Tree Classifier dt_pred = dt_model.predict(X_test) dt_accuracy = accuracy_score(y_test, dt_pred) print(f"Decision Tree Classifier Accuracy: {dt_accuracy:.4f}") Decision Tree Classifier Accuracy: 0.9333</pre>	<pre>print(confusion_matrix(y_test, y_pred))</pre> <pre>[[14187 993] [1064 13840]]</pre>
	precision	recall	f1-score	support																													
0	0.93	0.93	0.93	15180																													
1	0.93	0.93	0.93	14904																													
accuracy			0.93	30084																													
macro avg	0.93	0.93	0.93	30084																													
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Random Forest	<pre>print(classification_report(y_test, y_pred))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.94</td><td>0.96</td><td>0.95</td><td>15180</td></tr><tr><td>1</td><td>0.96</td><td>0.94</td><td>0.95</td><td>14904</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.95</td><td>30084</td></tr><tr><td>macro avg</td><td>0.95</td><td>0.95</td><td>0.95</td><td>30084</td></tr><tr><td>weighted avg</td><td>0.95</td><td>0.95</td><td>0.95</td><td>30084</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.94	0.96	0.95	15180	1	0.96	0.94	0.95	14904	accuracy			0.95	30084	macro avg	0.95	0.95	0.95	30084	weighted avg	0.95	0.95	0.95	30084	<pre># Random Forest Classifier rf_pred = rf_model.predict(X_test) rf_accuracy = accuracy_score(y_test, rf_pred) print(f"Random Forest Classifier Accuracy: {rf_accuracy:.4f}") Random Forest Classifier Accuracy: 0.9506</pre>	<pre>print(confusion_matrix(y_test, y_pred))</pre> <pre>[[14552 628] [877 14027]]</pre>
	precision	recall	f1-score	support																													
0	0.94	0.96	0.95	15180																													
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weighted avg	0.95	0.95	0.95	30084																													
KNN	<pre>print(classification_report(y_test, y_pred))</pre>	<pre># K-Nearest Neighbors Classifier knn_pred = knn_model.predict(X_test) knn_accuracy = accuracy_score(y_test, knn_pred) print(f"K-Nearest Neighbors Classifier Accuracy: {knn_accuracy:.4f}") K-Nearest Neighbors Classifier Accuracy: 0.8924</pre>	<pre>print(confusion_matrix(y_test, y_pred))</pre> <pre>[[12406 2774] [439 14465]]</pre>																														

	<pre> precision recall f1-score support 0 0.97 0.82 0.89 15180 1 0.84 0.97 0.90 14904 accuracy 0.89 30084 macro avg 0.90 0.89 0.89 30084 weighted avg 0.90 0.89 0.89 30084 </pre>		
XGBoost	<pre> print(classification_report(y_test, y_pred)) precision recall f1-score support 0 0.91 0.98 0.94 15180 1 0.98 0.90 0.94 14904 accuracy 0.94 30084 macro avg 0.94 0.94 0.94 30084 weighted avg 0.94 0.94 0.94 30084 </pre>	<pre> # XGBoost Classifier xgb_pred = xgb_model.predict(X_test) xgb_accuracy = accuracy_score(y_test, xgb_pred) print(f"XGBoost Classifier Accuracy: {xgb_accuracy:.4f}") XGBoost Classifier Accuracy: 0.9416 </pre>	<pre> print(confusion_matrix(y_test, y_pred)) [[14863 317] [1421 13483]] </pre>