

Model Development Phase Template

Date	July 2024
Team ID	739663
Project Title	Estimating the Stock keeping units using Machine Learning
Maximum Marks	10 Marks

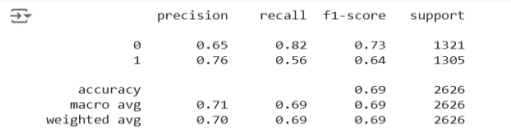
Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include a summary and training and validation performance metrics for multiple models, presented through respective screenshots.

Initial Model Training Code (5 marks):

Paste the screenshot of the model training code

Model Validation and Evaluation Report (5 marks):

Model	Summary	Training and Validation Performance Metrics
Model 1	Linear regression model typically include accuracy, precision, recall, R2 score to evaluate its predictive performance and generalization capability.	<p>LOGISTIC REGRESSION</p> <pre>[] #LOGISTIC REGRESSION # importing the library from sklearn.linear_model import LogisticRegression #initializing the model lr=LogisticRegression() #fit the model lr.fit(x_train,y_train) #predict the model predic=lr.predict(x_test) #finding accuracy,classification report from sklearn.metrics import classification_report print(classification_report(y_test,predic))</pre>  <pre> precision recall f1-score support 0 0.65 0.82 0.73 1321 1 0.76 0.56 0.64 1305 accuracy 0.69 2626 macro avg 0.71 0.69 0.69 2626 weighted avg 0.70 0.69 0.69 2626 </pre>

Model 2	Random forest regression model often encompass accuracy, precision, recall, R2 score to measure its prediction quality and robustness.	<div>RANDOM FOREST CLASSIFIER</div> <div><pre>[45] #random forest classifier from sklearn.ensemble import RandomForestClassifier rfc=RandomForestClassifier() rfc.fit(x_train,y_train) pred=rfc.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,pred))</pre></div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>0</td><td>0.69</td><td>0.87</td><td>0.77</td><td>1321</td></tr><tr><td>1</td><td>0.82</td><td>0.61</td><td>0.70</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.74</td><td>2626</td></tr><tr><td>macro avg</td><td>0.76</td><td>0.74</td><td>0.73</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.76</td><td>0.74</td><td>0.73</td><td>2626</td></tr></table></div>		precision	recall	f1-score	support	0	0.69	0.87	0.77	1321	1	0.82	0.61	0.70	1305	accuracy			0.74	2626	macro avg	0.76	0.74	0.73	2626	weighted avg	0.76	0.74	0.73	2626
	precision	recall	f1-score	support																												
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Model 3	Decision tree regression model commonly include accuracy, precision, recall, R2 score which help assess the model's prediction accuracy and generalizability.	<div>#decision tree classifier</div> <div><pre>from sklearn.tree import DecisionTreeClassifier dec=DecisionTreeClassifier() dec.fit(x_train,y_train) predi=dec.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,predi))</pre></div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>0</td><td>0.71</td><td>0.69</td><td>0.70</td><td>1321</td></tr><tr><td>1</td><td>0.69</td><td>0.71</td><td>0.70</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.70</td><td>2626</td></tr><tr><td>macro avg</td><td>0.70</td><td>0.70</td><td>0.70</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.70</td><td>0.70</td><td>0.70</td><td>2626</td></tr></table></div>		precision	recall	f1-score	support	0	0.71	0.69	0.70	1321	1	0.69	0.71	0.70	1305	accuracy			0.70	2626	macro avg	0.70	0.70	0.70	2626	weighted avg	0.70	0.70	0.70	2626
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macro avg	0.70	0.70	0.70	2626																												
weighted avg	0.70	0.70	0.70	2626																												
Model 4	K-nearest neighbors classifier model typically include accuracy, precision, recall, R2 score to evaluate its prediction performance and generalization ability	<div>K-NEAREST NEIGHBORS</div> <div><pre>[48] #knn from sklearn.neighbors import KNeighborsClassifier knn=KNeighborsClassifier() knn.fit(x_train,y_train) p=knn.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,p))</pre></div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>0</td><td>0.69</td><td>0.79</td><td>0.74</td><td>1321</td></tr><tr><td>1</td><td>0.75</td><td>0.65</td><td>0.69</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.72</td><td>2626</td></tr><tr><td>macro avg</td><td>0.72</td><td>0.72</td><td>0.72</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.72</td><td>0.72</td><td>0.72</td><td>2626</td></tr></table></div>		precision	recall	f1-score	support	0	0.69	0.79	0.74	1321	1	0.75	0.65	0.69	1305	accuracy			0.72	2626	macro avg	0.72	0.72	0.72	2626	weighted avg	0.72	0.72	0.72	2626
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macro avg	0.72	0.72	0.72	2626																												
weighted avg	0.72	0.72	0.72	2626																												
Model 5	XGB classifier model typically include accuracy, precision, recall, R2 score used to evaluate the model's predictive performance and ability to generalize	<div>XGBOOST CLASSIFIER</div> <div><pre>[49] #XGBOOST CLASSIFICATION from xgboost import XGBClassifier xg=XGBClassifier() xg.fit(x_train,y_train) p=xg.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,p))</pre></div> <div><table><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr><tr><td>0</td><td>0.71</td><td>0.80</td><td>0.75</td><td>1321</td></tr><tr><td>1</td><td>0.77</td><td>0.66</td><td>0.71</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.73</td><td>2626</td></tr><tr><td>macro avg</td><td>0.74</td><td>0.73</td><td>0.73</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.74</td><td>0.73</td><td>0.73</td><td>2626</td></tr></table></div>		precision	recall	f1-score	support	0	0.71	0.80	0.75	1321	1	0.77	0.66	0.71	1305	accuracy			0.73	2626	macro avg	0.74	0.73	0.73	2626	weighted avg	0.74	0.73	0.73	2626
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Model 6

Ridge classifier model typically include accuracy, precision, recall, R2 score, and mean squared error to evaluate its prediction performance and generalization.

RIDGE CLASSIFIER

```
[50] #RIDGE CLASSIFIER
from sklearn.linear_model import RidgeClassifier
rg=RidgeClassifier()
rg.fit(x_train,y_train)
p=rg.predict(x_test)
from sklearn.metrics import classification_report
print(classification_report(y_test,p))
```

	precision	recall	f1-score	support
0	0.65	0.80	0.72	1321
1	0.74	0.56	0.64	1305
accuracy			0.68	2626
macro avg	0.69	0.68	0.68	2626
weighted avg	0.69	0.68	0.68	2626