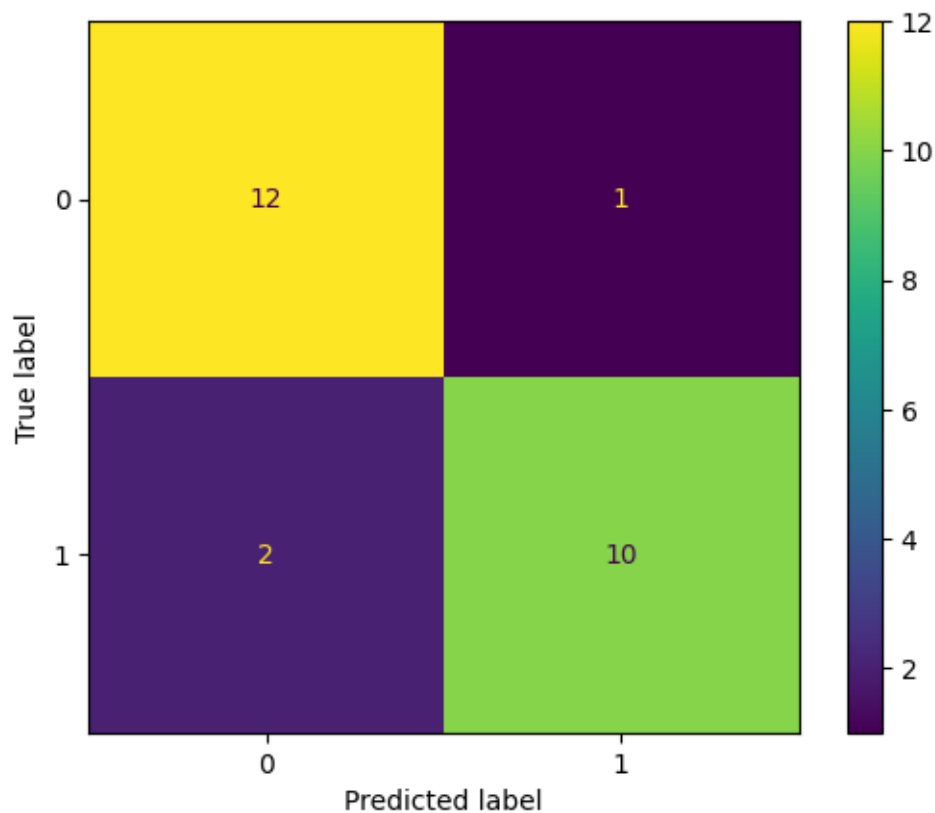
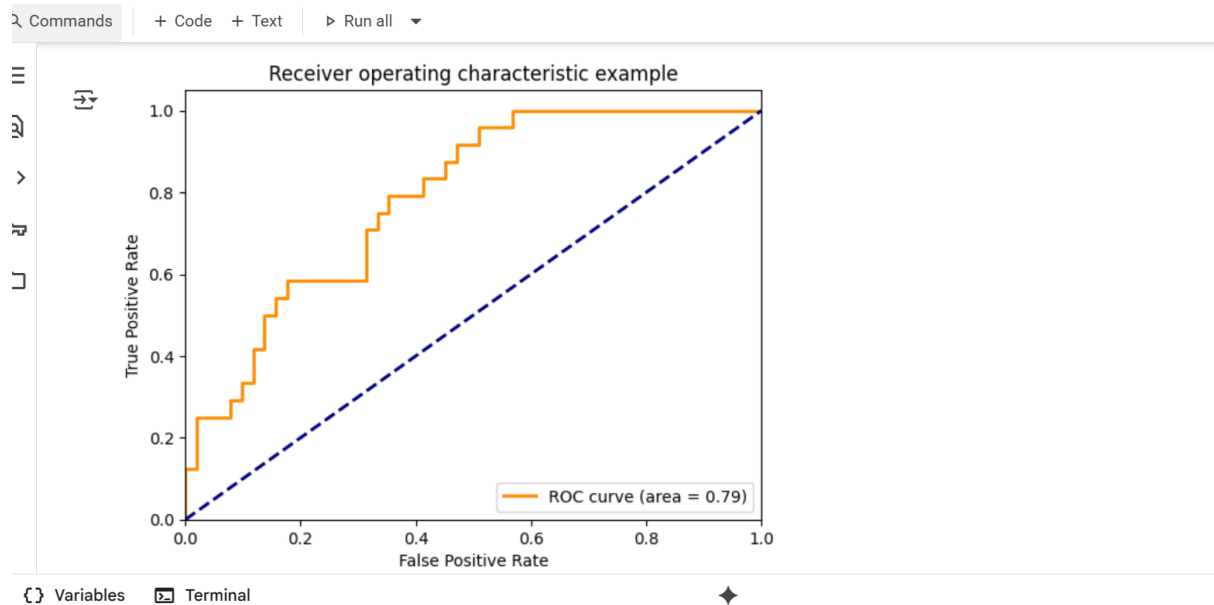


Activity 6

In this activity, we explored different performance evaluation techniques to assess model effectiveness. Metrics such as the F1-score and R-squared were used to measure how well the model balanced precision and recall or explained variance, depending on the task. Additionally, we constructed a confusion matrix to visualise true positives, false positives, true negatives, and false negatives. This helped us understand the practical impact of model errors, especially in sensitive applications where misclassification can have real-world consequences.





Commands + Code + Text Run all

RMSE

```
[14] from sklearn.metrics import mean_squared_error
y_true = [3, -0.5, 2, 7]
y_pred = [2.5, 0.0, 2, 8]
mean_squared_error(y_true, y_pred)
```

0.375

MAE

```
[15] from sklearn.metrics import mean_absolute_error
y_true = [3, -0.5, 2, 7]
y_pred = [2.5, 0.0, 2, 8]
mean_absolute_error(y_true, y_pred)
```

0.5

r squared

```
[16] from sklearn.metrics import r2_score
r2_score(y_true, y_pred)
```

0.9486081370449679

Variables Terminal

Os

[8] from sklearn.metrics import classification_report
y_true = [0, 1, 2, 2, 2]
y_pred = [0, 0, 2, 2, 1]
target_names = ['class 0', 'class 1', 'class 2']
print(classification_report(y_true, y_pred, target_names=target_names))

	precision	recall	f1-score	support
class 0	0.50	1.00	0.67	1
class 1	0.00	0.00	0.00	1
class 2	1.00	0.67	0.80	3
accuracy			0.60	5
macro avg	0.50	0.56	0.49	5
weighted avg	0.70	0.60	0.61	5

Os

[9] from sklearn.datasets import load_breast_cancer
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import roc_auc_score
X, y = load_breast_cancer(return_X_y=True)
clf = LogisticRegression(solver="liblinear", random_state=0).fit(X, y)
roc_auc_score(y, clf.predict_proba(X)[:, 1])

{}

Variables

Terminal

6