20250423 01

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1 Logistic Regression & Classification Evaluation

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[8]: # Import dataset
      from sklearn.datasets import load_breast_cancer
      import pandas as pd
      data = load_breast_cancer()
      X = pd.DataFrame(data.data, columns = data.feature_names)
      y = pd.Series(data.target, name = 'cancer') # 0 = malignant, 1 = benign
[10]: # Creating training and test sets
      from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, __
       \negrandom_state = 42)
[12]: # Creating Logistic Regression model
      from sklearn.linear_model import LogisticRegression
      model = LogisticRegression(solver = 'liblinear')
      model.fit(X_train, y_train)
      # Predicting categories
      y_pred = model.predict(X_test)
      # Predicting probability (for ROC curve)
      y_prob = model.predict_proba(X_test)[:,1]
[44]: # Accuracy and basic evaluation
      from sklearn.metrics import accuracy_score, classification_report,_
       ⇔confusion matrix
      # Accuracy
      print("Accuracy:", accuracy_score(y_test, y_pred), '\n')
      # Confusion Matrix
      print("Confusion Matrix: \n", confusion_matrix(y_test, y_pred), '\n')
```

```
# Classification report
print(classification_report(y_test, y_pred))
```

Accuracy: 0.956140350877193

Confusion Matrix:

[[39 4] [1 70]]

	precision	recall	f1-score	support
0	0.97	0.91	0.94	43
1	0.95	0.99	0.97	71
accuracy			0.96	114
macro avg	0.96	0.95	0.95	114
weighted avg	0.96	0.96	0.96	114

```
[56]: # Drawing ROC curve
from sklearn.metrics import roc_curve, auc
import matplotlib.pyplot as plt

# ROC
fpr, tpr, thresholds = roc_curve(y_test, y_prob)
roc_auc = auc(fpr, tpr)

#
plt.figure(figsize = (8, 6))
plt.plot(fpr, tpr, color = 'darkorange', lw = 2, label = f'ROC curve (AUC = Groc_auc: 2f})')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('ROC Curve - Logistic Regression')
plt.legend(loc = "lower right")
plt.show()
```

