Logistic Regression Study Guide

Introduction to Logistic Regression

Logistic regression is a statistical model used for binary classification problems.

It predicts the probability of an event occurring, such as whether a person has a job (1) or not (0), based on input features.

Step-by-Step Explanation

Step 1: Importing Required Libraries

```python

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

from sklearn.model\_selection import train\_test\_split

from sklearn.linear\_model import LogisticRegression

from sklearn.metrics import accuracy\_score, confusion\_matrix, classification\_report, roc\_curve, roc\_auc\_score, f1\_score, precision\_score, recall\_score

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## Why?

- `numpy` for numerical operations
- `matplotlib` and `seaborn` for visualization
- `sklearn.model\_selection` for data splitting
- `sklearn.linear\_model` for logistic regression
- `sklearn.metrics` for model evaluation

```
Step 2: Creating a Sample Dataset
```python
age = np.array([16, 24, 31, 29, 40, 50, 21, 33, 45, 39])
job = np.array([0, 1, 1, 1, 1, 0, 1, 1, 1])
```

Step 3: Understanding shape and reshape

- `.shape` returns the dimensions of an array.

- `reshape(-1,1)` is used to convert a 1D array into a column vector for machine learning models. Example: ```python age = age.reshape(-1, 1)print(age.shape) # Output: (10, 1) ### Step 4: Splitting Data into Training and Testing Sets ```python X_train, X_test, y_train, y_test = train_test_split(age, job, test_size=0.2, random_state=42) ### Step 5: Training the Logistic Regression Model ```python model = LogisticRegression() model.fit(X_train, y_train) ### Step 6: Making Predictions ```python y_pred = model.predict(X_test) ### Step 7: Evaluating the Model ```python print(confusion_matrix(y_test, y_pred)) print(classification_report(y_test, y_pred)) print("Accuracy:", accuracy_score(y_test, y_pred)) ### Step 8: Plotting the ROC Curve & AUC Score ```python fpr, tpr, _ = roc_curve(y_test, model.predict_proba(X_test)[:,1]) plt.plot(fpr, tpr, label="ROC Curve") plt.xlabel("False Positive Rate")

plt.ylabel("True Positive Rate")

plt.title("ROC Curve")

```
plt.legend()
plt.show()

print("AUC Score:", roc_auc_score(y_test, model.predict_proba(X_test)[:,1]))
...
```