

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder, StandardScaler
from sklearn.pipeline import Pipeline
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import (
    accuracy_score, precision_score, recall_score, f1_score, roc_auc_score,
    confusion_matrix, classification_report, RocCurveDisplay, ConfusionMatrixDisplay
)
```

```
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
path = '/content/drive/MyDrive/PRAKTIKUM/PRAKTIKUM7/DATA/'
```

```
df = pd.read_csv(path + "apples_and_oranges.csv")
df
```

	Weight	Size	Class
0	69	4.39	orange
1	69	4.21	orange
2	65	4.09	orange
3	72	5.85	apple
4	67	4.70	orange
5	73	5.68	apple
6	70	5.56	apple
7	75	5.11	apple
8	74	5.36	apple
9	65	4.27	orange
10	73	5.79	apple
11	70	5.47	apple
12	74	5.53	apple
13	68	4.47	orange
14	74	5.22	apple
15	65	4.48	orange
16	69	4.66	orange
17	75	5.25	apple
18	67	4.18	orange
19	74	5.50	apple
20	66	4.13	orange
21	70	4.83	orange
22	69	4.61	orange
23	68	4.08	orange
24	67	4.25	orange
25	71	5.35	apple
26	67	4.01	orange
27	70	4.22	orange
28	74	5.25	apple
29	71	5.26	apple
30	73	5.78	apple
31	66	4.68	orange
32	72	5.72	apple
33	73	5.17	apple
34	68	4.83	orange
35	69	4.11	orange
36	69	4.76	orange
37	74	5.48	apple
38	70	5.59	apple
39	73	5.03	apple

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 40 entries, 0 to 39
Data columns (total 3 columns):
 #   Column  Non-Null Count  Dtype  
--- 
 0   Weight   40 non-null    int64  
 1   Size     40 non-null    float64 
 2   Class    40 non-null    object  
dtypes: float64(1), int64(1), object(1)
memory usage: 1.1+ KB
```

```
df.isnull().sum()
```

	0
Weight	0
Size	0
Class	0

dtype: int64

```
df["Class"].value_counts()
```

	count
Class	
orange	20
apple	20

dtype: int64

```
df.duplicated().sum()
```

```
np.int64(0)
```

```
df = df.drop_duplicates()
```

```
df.duplicated().sum()  
np.int64(0)
```

```
df['Class'].unique()  
array(['orange', 'apple'], dtype=object)
```

```
# 1. Pemetaan kolom Class -> biner (apple=1, orange=0)  
map_class = {'apple': 1, 'orange': 0}  
df['Class_bin'] = df['Class'].map(map_class).astype('Int64')  
  
print("Distribusi Class_bin:\n", df['Class_bin'].value_counts())
```

Class_bin	count
0	20
1	20

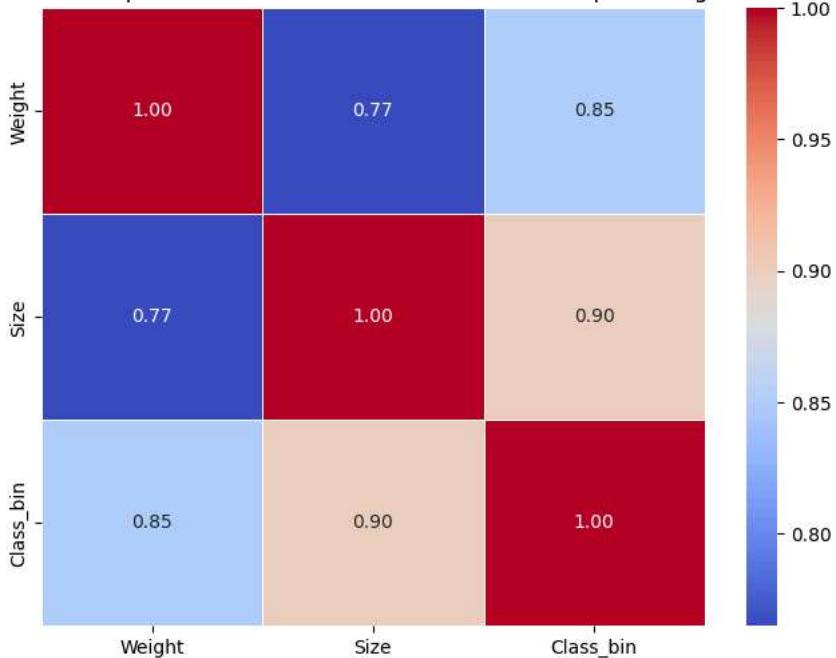
Name: count, dtype: Int64

```
corr_matrix = df.corr(numeric_only=True)  
corr_matrix
```

	Weight	Size	Class_bin
Weight	1.000000	0.765027	0.849528
Size	0.765027	1.000000	0.900641
Class_bin	0.849528	0.900641	1.000000

```
import seaborn as sns  
import matplotlib.pyplot as plt  
  
plt.figure(figsize=(8,6))  
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', fmt=".2f", linewidths=0.5)  
plt.title("Heatmap Korelasi Antar Variabel Numerik terhadap Stunting", fontsize=12)  
plt.show()
```

Heatmap Korelasi Antar Variabel Numerik terhadap Stunting



```
# Fitur numerik
feature_num = ['Weight', 'Size']

# Gabungkan & drop missing
use_cols = feature_num + ['Class_bin']
df_model = df[use_cols].dropna().copy()

X = df_model[feature_num]
y = df_model['Class_bin']

print("X shape:", X.shape)
print("y shape:", y.shape)

# Output:
# X shape: (100000, 4)
# y shape: (100000,)

X shape: (40, 2)
y shape: (40,)
```

```
X_train, X_test, y_train, y_test = train_test_split(
    X,
    y,
    test_size=0.2,
    random_state=42,
    stratify=y
)

print("Data latih:", X_train.shape)
print("Data uji:", X_test.shape)

Data latih: (32, 2)
Data uji: (8, 2)
```

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
import pandas as pd

# 1. Muat dan Persiapkan Data
dfdf = pd.read_csv(path + "apples_and_oranges.csv")
map_class = {'apple': 1, 'orange': 0}
df['Class_bin'] = df['Class'].map(map_class).astype('Int64')

# 2. Definisikan Features dan Target
feature_num = ['Weight', 'Size'] # Fitur numerik
X = df[feature_num]
y = df['Class_bin']

# 3. Bagi Data Menjadi Data Latih dan Uji
```

```

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.3, random_state=42, stratify=y
)

# 4. Inisialisasi Preprocessing
# Scale hanya fitur numerik
preprocess = ColumnTransformer(
    transformers=[
        ('num', StandardScaler(), feature_num),
    ],
    remainder='drop'
)

# 5. Inisialisasi Model
model = LogisticRegression(
    max_iter=1000,
    solver='lbfgs',
    class_weight='balanced',
    random_state=42
)

# 6. Buat Pipeline
clf = Pipeline([
    ('preprocess', preprocess),
    ('model', model)
])

# 7. Latih Model
clf.fit(X_train, y_train)
print("✅ Model Logistic Regression trained successfully on the apples_and_oranges dataset.")


```

✅ Model Logistic Regression trained successfully on the apples_and_oranges dataset.

```

# Prediction & probability
y_pred = clf.predict(X_test)
y_prob = clf.predict_proba(X_test)[:, 1]

# Calculate metrics
print(f"Akurasi : {accuracy_score(y_test, y_pred):.4f}")
print(f"Precision : {precision_score(y_test, y_pred, zero_division=0):.4f}")
print(f"Recall : {recall_score(y_test, y_pred, zero_division=0):.4f}")
print(f"F1-Score : {f1_score(y_test, y_pred, zero_division=0):.4f}")
print(f"ROC-AUC : {roc_auc_score(y_test, y_prob):.4f}")

Akurasi : 1.0000
Precision : 1.0000
Recall : 1.0000
F1-Score : 1.0000
ROC-AUC : 1.0000

```

```

# --- Visualisasi (Langkah Terakhir Anda) ---
# 1. Prediksi
y_pred = clf.predict(X_test)

# 2. Confusion Matrix
# display_labels disesuaikan menjadi ['Orange', 'Apple']
cm = confusion_matrix(y_test, y_pred)
disp = ConfusionMatrixDisplay(confusion_matrix=cm,
                               display_labels=['Orange', 'Apple'])

fig_cm, ax_cm = plt.subplots()
disp.plot(values_format='d', ax=ax_cm)
ax_cm.set_title("Confusion Matrix")
plt.show()

# 3. ROC Curve
fig_roc, ax_roc = plt.subplots()
RocCurveDisplay.from_estimator(clf, X_test, y_test, ax=ax_roc)
ax_roc.set_title("ROC Curve Logistic Regression")
plt.show()

```

