### Import library

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.metrics import classification_report, accuracy_score, confusion_matrix
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
```

#### Load dataset

```
data = pd.read_csv('/content/obesitas.csv')
```

## Eksplorasi dataset

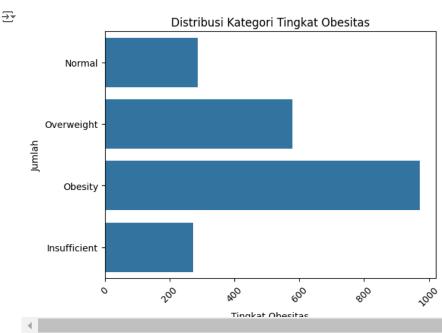
```
print("\nDataset Overview:\n", data.head())
print("\nInfo Dataset:\n")
data.info()
print("\nMissing Values:\n", data.isnull().sum())
print("\nDescriptive Statistics:\n", data.describe())
```

```
неіgnt
                                    ZIII non-nuII
                                                    †10at64
   Weight
                                    2111 non-null
                                                    float64
    family_history_with_overweight 2111 non-null
                                                    object
   FAVC
                                    2111 non-null
   FCVC
                                    2111 non-null
                                                    float64
   NCP
                                    2111 non-null
                                                    float64
   CAEC
                                    2111 non-null
                                                    object
   SMOKE
                                    2111 non-null
                                                    object
10 CH20
                                    2111 non-null
                                                    float64
11 SCC
                                    2111 non-null
                                                    object
                                    2111 non-null
                                                    float64
13 TUE
                                    2111 non-null
                                                    float64
14 CALC
                                    2111 non-null
                                                    object
15 MTRANS
                                    2111 non-null
                                                    object
16 NObeyesdad
                                    2111 non-null
                                                    object
```

	CH20	FAF	TUE
count	2111.000000	2111.000000	2111.000000
mean	2.008011	1.010298	0.657866
std	0.612953	0.850592	0.608927
min	1.000000	0.000000	0.000000
25%	1.584812	0.124505	0.000000
50%	2.000000	1.000000	0.625350
75%	2.477420	1.666678	1.000000
max	3.000000	3.000000	2.000000

#### Visualisasi data

```
sns.countplot(data['NObeyesdad'])
plt.title('Distribusi Kategori Tingkat Obesitas')
plt.xlabel('Tingkat Obesitas')
plt.ylabel('Jumlah')
plt.xticks(rotation=45)
plt.show()
```



# **Preprocessing Data**

# Encoding untuk fitur kategorikal

```
encoder = LabelEncoder()
for col in data.select_dtypes(include=['object']).columns:
    data[col] = encoder.fit_transform(data[col])
```

#### Pisahkan fitur dan label

```
X = data.drop(columns=['NObeyesdad'])
y = data['NObeyesdad']
```

## Standarisasi fitur

```
scaler = StandardScaler()
X = scaler.fit_transform(X)
```

## Split data untuk training dan testing

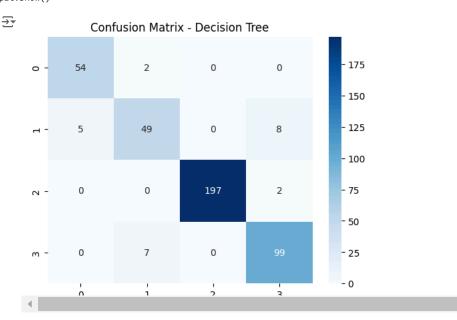
```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

#### Model 1: Decision Tree

```
print("\n=== Decision Tree Classifier ===")
dt_model = DecisionTreeClassifier(random_state=42)
dt_model.fit(X_train, y_train)
dt_predictions = dt_model.predict(X_test)
print("Accuracy:", accuracy_score(y_test, dt_predictions))
print("Classification Report:\n", classification_report(y_test, dt_predictions))
     === Decision Tree Classifier ===
     Accuracy: 0.9432624113475178
     Classification Report:
                    precision
                                  recall f1-score
                                                     support
                0
                        0.92
                                   0.96
                                             0.94
                                                         56
                        0.84
                                   0.79
                                             0.82
                1
                                                         62
                2
                        1.00
                                   0.99
                                             0.99
                                                         199
                3
                        0.91
                                   0.93
                                             0.92
                                                        106
                                                        423
                                             0.94
         accuracy
                                             0.92
        macro avg
                        0.92
                                   0.92
                                                        423
     weighted avg
                        0.94
                                   0.94
                                             0.94
                                                        423
```

#### Confusion Matrix Decision Tree

```
sns.heatmap(confusion_matrix(y_test, dt_predictions), annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix - Decision Tree')
plt.show()
```

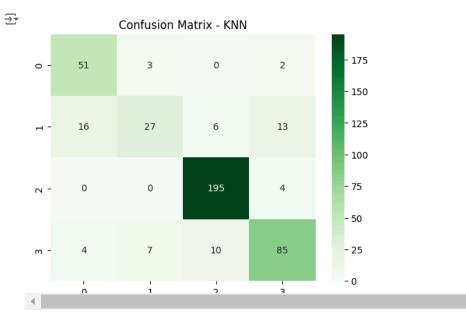


### Model 2: K-Nearest Neighbors

```
KIIII_IIIOGCI..IIC(/\_CIGII) y_CIGIII/
knn_predictions = knn_model.predict(X_test)
print("Accuracy:", accuracy_score(y_test, knn_predictions))
print("Classification Report:\n", classification_report(y_test, knn_predictions))
₹
     === K-Nearest Neighbors (KNN) ===
     Accuracy: 0.8463356973995272
     Classification Report:
                    precision
                                  recall f1-score
                                                      support
                0
                         0.72
                                   0.91
                                              0.80
                         0.73
                                   0.44
                                              0.55
                                                          62
                1
                                   0.98
                2
                         0.92
                                              0.95
                                                         199
                3
                         0.82
                                   0.80
                                             0.81
                                                         106
                                              0.85
                                                         423
         accuracy
                         0.80
                                   0.78
        macro avg
                                              0.78
                                                         423
     weighted avg
                         0.84
                                   0.85
                                              0.84
                                                         423
```

#### Confusion Matrix KNN

```
sns.heatmap(confusion_matrix(y_test, knn_predictions), annot=True, fmt='d', cmap='Greens')
plt.title('Confusion Matrix - KNN')
plt.show()
```



# Perbandingan Kinerja

```
dt_accuracy = cross_val_score(dt_model, X, y, cv=5, scoring='accuracy').mean()
knn_accuracy = cross_val_score(knn_model, X, y, cv=5, scoring='accuracy').mean()

print("\n=== Model Comparison ===")
print(f"Decision Tree Accuracy (CV): {dt_accuracy:.2f}")

print(f"KNN Accuracy (CV): {knn_accuracy:.2f}")

=== Model Comparison ===
Decision Tree Accuracy (CV): 0.93
KNN Accuracy (CV): 0.81
```

# Visualisasi Perbandingan

```
models = ['Decision Tree', 'KNN']
accuracies = [dt_accuracy, knn_accuracy]
sns.barplot(x=models, y=accuracies, palette='viridis')
nlt.title('Perhandingan Akurasi Model')
```

plt.ylabel('Accuracy')
plt.show()

<ipython-input-21-7df307d9cee7>:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend sns.barplot(x=models, y=accuracies, palette='viridis')

