```
In [2]: # Import necessary libraries
import pandas as pd
from sklearn.datasets import fetch_openml
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
import seaborn as sns
# Step 1: Load the Dataset
# Fetch the Wine Quality dataset and save it as a CSV file
data = fetch_openml(name='wine-quality-red', as_frame=True)
df = pd.concat([data.data, data.target.rename('quality')], axis=1)
df.to_csv("wine_quality.csv", index=False) # Save as CSV
# Load the CSV file
df = pd.read_csv("wine_quality.csv")
# Print the dataset structure to verify column names
print(df.head())
print(df.columns)
# Step 2: Preprocess the Data
# Separate features and target variable
features = df.drop(columns=['quality'])
target = df['quality']
# Standardize the features
scaler = StandardScaler()
features_scaled = scaler.fit_transform(features)
# Step 3: Apply KMeans Clustering
# Initialize KMeans with 3 clusters
kmeans = KMeans(n_clusters=3, random_state=42, n_init=10)
kmeans.fit(features_scaled)
# Add cluster labels to the dataframe
df['cluster'] = kmeans.labels_
# Step 4: Visualize the Clusters
# Use the correct column names from the dataset
# Replace 'fixed acidity' and 'alcohol' with the actual column names
if 'fixed acidity' in features.columns and 'alcohol' in features.columns:
    x_{col} = 'alcohol'
    y_col = 'fixed acidity'
else:
    # Replace with alternative columns if those do not exist
    x_col = features.columns[0]
    y_col = features.columns[1]
plt.figure(figsize=(10, 6))
sns.scatterplot(
   x=features[x_col],
    y=features[y_col],
    hue=df['cluster'],
    palette='viridis',
    s=100,
    alpha=0.7
plt.title(f"Clustering of Wine Quality Data ({x_col} vs. {y_col})")
plt.xlabel(x_col.capitalize())
plt.ylabel(y_col.capitalize())
plt.legend(title="Cluster")
plt.grid(True)
plt.show()
C:\Users\Kritika Agrahari\anaconda3\Lib\site-packages\sklearn\datasets\_openml.py:1002: FutureWarning: The default value of `parser` will change from
`'liac-arff'` to `'auto'` in 1.4. You can set `parser='auto'` to silence this warning. Therefore, an `ImportError` will be raised from 1.4 if the datase
t is dense and pandas is not installed. Note that the pandas parser may return different data types. See the Notes Section in fetch_openml's API doc for
details.
 warn(
   fixed_acidity volatile_acidity citric_acid residual_sugar chlorides
                                                                  0.076
            7.4
                             0.70
                                           0.00
                                                           1.9
                              0.88
                                           0.00
                                                                     0.098
             7.8
                              0.76
                                                            2.3
                                                                     0.092
                                           0.04
            11.2
                              0.28
                                           0.56
                                                            1.9
                                                                     0.075
             7.4
                              0.70
                                           0.00
                                                            1.9
                                                                     0.076
   free_sulfur_dioxide total_sulfur_dioxide density
                                                         pH sulphates \
                  11.0
                                        34.0
                                               0.9978 3.51
                  25.0
                                        67.0
                                               0.9968 3.20
                  15.0
                                        54.0
                                               0.9970 3.26
                                                                  0.65
3
                  17.0
                                        60.0
                                               0.9980 3.16
                                                                  0.58
                  11.0
                                        34.0 0.9978 3.51
                                                                  0.56
4
   alcohol quality
       9.4
       9.8
2
       9.8
       9.8
       9.4
Index(['fixed_acidity', 'volatile_acidity', 'citric_acid', 'residual_sugar',
       'chlorides', 'free_sulfur_dioxide', 'total_sulfur_dioxide', 'density',
       'pH', 'sulphates', 'alcohol', 'quality'],
      dtype='object')
                      Clustering of Wine Quality Data (fixed_acidity vs. volatile_acidity)
   1.6
                                                                                                   Cluster
                                                                                                        0
                                                                                                        1
   1.4
                                                                                                        2
   1.2
```

