In [1]:	INTRODUCTION TO MACHINE LEARNING  Import Library  import numpy as np import pandas as pd
In [2]: Out[2]:	Supervised Learning  Import Data  dataset = pd.read_csv(r"D:\SEMESTER 4\IS411 Data Modelling\LAB\Bahan Modul 7\winequality-red.csv", delimiter=';') dataset.head(5)  fixed acidity_volatile acidity_citric acid_residual sugar_chlorides_free sulfur dioxide_total sulfur dioxide_density_pH_sulphates_alcohol_quality
Out[2]: In [3]:	0       7.4       0.70       0.00       1.9       0.076       11.0       34.0       0.9978       3.51       0.56       9.4       5         1       7.8       0.88       0.00       2.6       0.098       25.0       67.0       0.9968       3.20       0.68       9.8       5         2       7.8       0.76       0.04       2.3       0.092       15.0       54.0       0.9970       3.26       0.65       9.8       5         3       11.2       0.28       0.56       1.9       0.075       17.0       60.0       0.9980       3.16       0.58       9.8       6         4       7.4       0.70       0.00       1.9       0.076       11.0       34.0       0.9978       3.51       0.56       9.4       5
In [3]:	<pre>print(dataset.groupby('quality').size()) dataset.info()  quality 3     10 4     53 5     681 6     638 7     199 8     18 dtype: int64</pre>
	<pre><class 'pandas.core.frame.dataframe'=""> RangeIndex: 1599 entries, 0 to 1598 Data columns (total 12 columns): # Column</class></pre>
In [4]:	<pre>dtypes: float64(11), int64(1) memory usage: 150.0 KB  #memisahkan kolum independent variable dengan dependent variable  redwine = dataset.copy() Y = redwine['quality'] X = redwine.drop(columns = 'quality')</pre>
In [5]:	<pre>#membagi data menjadi 2 bagian, untuk training dan testing dengan bantuan library 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, random_state=0)  Logistic Regression  from sklearn.linear_model import LogisticRegression logreg = LogisticRegression()</pre>
	logreg.fit(X_train, y_train) print('Accuracy of Logistic regression classifier on training set: {:.2f}'.format(logreg.score(X_train, y_train))) print('Accuracy of Logistic regression classifier on test set: {:.2f}'.format(logreg.score(X_test, y_test)))  Accuracy of Logistic regression classifier on training set: 0.58 Accuracy of Logistic regression classifier on test set: 0.62  C:\Users\Darren\anaconda3\lib\site-packages\sklearn\linear_model\_logistic.py:814: ConvergenceWarning: lbfgs failed to converge (status=1): STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.  Increase the number of iterations (max_iter) or scale the data as shown in:
In [7]:	https://scikit-learn.org/stable/modules/preprocessing.html  Please also refer to the documentation for alternative solver options:     https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression     n_iter_i = _check_optimize_result(  Decision Tree  from sklearn.tree import DecisionTreeClassifier  clf = DecisionTreeClassifier()
In [8]:	clf.fit(X_train, y_train) print('Accuracy of Decision Tree Classifier on training set: {:.2f}'.format(clf.score(X_train, y_train))) print('Accuracy of Decision Tree Classifier on test set: {:.2f}'.format(clf.score(X_test, y_test)))  Accuracy of Decision Tree Classifier on training set: 1.00 Accuracy of Decision Tree Classifier on test set: 0.69  KNN  from sklearn.neighbors import KNeighborsClassifier
	<pre>knn = KNeighborsClassifier() knn.fit(X_train, y_train) print('Accuracy of K-NN classifier on training set: {:.2f}'.format(knn.score(X_train, y_train))) print('Accuracy of K-NN classifier on test set: {:.2f}'.format(knn.score(X_test, y_test)))  Accuracy of K-NN classifier on training set: 0.66 Accuracy of K-NN classifier on test set: 0.48  Naive Bayes</pre>
In [9]:	<pre>gnb = GaussianNB() gnb.fit(X_train, y_train) print('Accuracy of GNB classifier on training set: {:.2f}'.format(gnb.score(X_train, y_train))) print('Accuracy of GNB classifier on test set: {:.2f}'.format(gnb.score(X_train, y_train))) Accuracy of GNB classifier on training set: 0.55 Accuracy of GNB classifier on test set: 0.55</pre>
In [10]: In [11]:	#Cobalah hilangkan data dengan kualitas 3,4,7, dan 8 sehingga data hanya memiliki 2 output kategori, dan #bagaimana hasil prediksinya? Manakah prediksi data yang lebih baik? (Boleh pilih gunakan salah satu algoritma).  df_filtered = dataset[-dataset['quality'].isin([3, 4, 7, 8])]  from sklearn.ensemble import RandomForestClassifier from sklearn.metrics import accuracy_score from sklearn.model_selection import train_test_split
In [12]: In [13]:	<pre>y = df_filtered['quality'] X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)  rf = RandomForestClassifier(n_estimators=100, random_state=42)  rf.fit(X_train, y_train)</pre>
Tp [14].	<pre>y_pred = rf.predict(X_test) accuracy = accuracy_score(y_test, y_pred) print("Accuracy: {:.2f}%".format(accuracy * 100)) Accuracy: 76.89%  Unsupervised learning winecluster = dataset.copy()</pre>
In [14]: Out[14]:	<pre>winecluster = winecluster.drop(columns='quality') winecluster</pre>
	4         7.4         0.700         0.00         1.9         0.076         11.0         34.0         0.99780         3.51         0.56         9.4 <t< td=""></t<>
In [15]:	1598 6.0 0.310 0.47 3.6 0.067 18.0 42.0 0.99549 3.39 0.66 11.0  1599 rows × 11 columns  #menggunakan grafik import matplotlib.pyplot as plt  plt.scatter(winecluster['volatile acidity'], winecluster['fixed acidity'])  plt.xlim(0,2)
	plt.show()  16
In [16]:	x = dataset.iloc[:,0:2] x.info()
	<pre>from sklearn.cluster import KMeans  <class 'pandas.core.frame.dataframe'=""> RangeIndex: 1599 entries, 0 to 1598 Data columns (total 2 columns):     # Column</class></pre>
In [17]:	<pre>wcss = [] for i in range(1,11):     kmeans = KMeans(n_clusters = i, init='k-means++', random_state = 42)     kmeans.fit(x)     wcss.append(kmeans.inertia_)  plt.plot(range(1,11), wcss) plt.xlabel('Number of clusters') plt.ylabel('wcss') plt.show()</pre>
	C:\Users\Darren\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1036: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=7.  warnings.warn(  5000  4000  4000
In [18]:	2000 - 1000 - 2 4 6 8 10  kmeans = KMeans(4)
In [18]: Out[18]: In [19]:	<pre>kmeans = kmeans(4) kmeans.fit(x)  identified_clusters = kmeans.fit_predict(x) identified_clusters  array([2, 0, 0,, 2, 2, 2])  wine_clusters = winecluster.copy() wine_clusters['Clusters'] = identified_clusters plt.scatter(winecluster['fixed acidity'], winecluster['volatile acidity'], c=wine_clusters['Clusters'], cmap='Spectral')</pre>
Out[19]:	<pre> </pre> <pre> <pre> <pre></pre></pre></pre>
	0.6
In [20]:	#Coba lakukan clustering berdasarkan pada kolom free sulfur dioxide dan #total sulfur dioxide, dan tampilkan hasilnya. Apakah nilai K-nya sama dengan clustering sebelumnya?  import matplotlib.pyplot as plt  plt.scatter(winecluster['free sulfur dioxide'], winecluster['total sulfur dioxide'])  plt.xlim(0,80)  plt.title('free sulfur dioxide vs total sulfur dioxide')  plt.show()
	free sulfur dioxide vs total sulfur dioxide  250 -  200 -  150 -
In [21]:	x2 = dataset.iloc[:,5:7] x2.info()
	<pre>from sklearn.cluster import KMeans  <class 'pandas.core.frame.dataframe'=""> RangeIndex: 1599 entries, 0 to 1598 Data columns (total 2 columns): # Column Non-Null Count Dtype</class></pre>
In [22]: In [23]:	<pre>#!pip install yellowbrick from yellowbrick.cluster import KElbowVisualizer  wcss2 = [] for i in range(1,11):     kmeans2 = KMeans(n_clusters = i, init='k-means++', random_state = 42)     kmeans2.fit(x2)     wcss2.append(kmeans2.inertia_)</pre>
	<pre>visualizer = KElbowVisualizer(kmeans2, k=(2,10))  plt.plot(range(1,11), wcss2) plt.xlabel('Number of clusters') plt.ylabel('wcss') plt.show() visualizer.fit(x2) visualizer.show()  C:\Users\Darren\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1036: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than</pre>
	available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=7.  warnings.warn(  1.75  1.50  1.25
	95 1.00 0.75 0.50 0.25
	Distortion Score Elbow for KMeans Clustering  elbow at k = 4, score = 273399.671  022  020
	200000 0.18 g g g g g g g g g g g g g g g g g g g
Out[23]: In [24]:	<pre>// Indeptified_clusters2 = kmeans2.fit_predict(x) identified_clusters2 = kmeans2.fit_predict(x)</pre> <pre>// Resolution Score Elbow for Kmeans Clustering'   xlabel = 'k', ylabel = 'distortion score'   xlabel = 'k', ylabel = 'k', ylabel = 'distortion score'   xlabel = 'k', ylabel = 'distortion score'   xlabel = 'k', ylabel = 'k', ylabel = 'distortion score'   xlabel = 'k', ylabel = 'k', ylabel = 'distortion score'   xlabel = 'k', ylabel = 'k', yla</pre>
Out[24]: In [25]: Out[25]:	<pre>identified_clusters2 array([1, 1, 1,, 0, 0, 0])  wine_clusters2 = winecluster.copy() wine_clusters2['Clusters'] = identified_clusters2 plt.scatter(winecluster['free sulfur dioxide'], winecluster['total sulfur dioxide'], c=wine_clusters2['Clusters'],cmap='cividis')  <matplotlib.collections.pathcollection 0x17fa0549070="" at=""></matplotlib.collections.pathcollection></pre>
	250 200 150
	White Data wine
In [26]: Out[26]:	#Cobalah lakukan clustering pada data whitewine dan bandingkan hasilnya. datasetwhiteclus = pd.read_csv(r"D:\SEMESTER 4\IS411 Data Modelling\LAB\Bahan Modul 7\winequality-white.csv") datasetwhiteclus.head(5)  fixed acidity volatile acidity citric acid residual sugar chlorides free sulfur dioxide total sulfur dioxide density pH sulphates alcohol quality  0 7.0 0.27 0.36 20.7 0.045 45.0 170.0 1.0010 3.00 0.45 8.8 6  1 6.3 0.30 0.34 1.6 0.049 14.0 132.0 0.9940 3.30 0.49 9.5 6  2 8.1 0.28 0.40 6.9 0.050 30.0 97.0 0.9951 3.26 0.44 10.1 6
In [27]: Out[27]:	3 7.2 0.23 0.32 8.5 0.058 47.0 186.0 0.9956 3.19 0.40 9.9 6 4 7.2 0.23 0.32 8.5 0.058 47.0 186.0 0.9956 3.19 0.40 9.9 6  datasetwhiteclus = datasetwhiteclus drop(columns='quality')  fixed acidity   volatile acidity   volatile acidity   volatile acidity   citric acid   residual sugar   chlorides   free sulfur dioxide   total sulfur dioxide   density   pH   sulphates   alcohol
	0         7.0         0.27         0.36         20.7         0.045         45.0         170.0         1.00100         3.00         0.45         8.8           1         6.3         0.30         0.34         1.6         0.049         14.0         132.0         0.99400         3.30         0.49         9.5           2         8.1         0.28         0.40         6.9         0.050         30.0         97.0         0.99510         3.26         0.44         10.1           3         7.2         0.23         0.32         8.5         0.058         47.0         186.0         0.99560         3.19         0.40         9.9           4         7.2         0.23         0.32         8.5         0.058         47.0         186.0         0.99560         3.19         0.40         9.9
In [28]:	4894 6.6 0.32 0.36 8.0 0.047 57.0 168.0 0.99490 3.15 0.46 9.6  4895 6.5 0.24 0.19 1.2 0.041 30.0 111.0 0.99254 2.99 0.46 9.4  4896 5.5 0.29 0.30 1.1 0.022 20.0 110.0 0.9869 3.34 0.38 12.8  4897 6.0 0.21 0.38 0.8 0.020 22.0 98.0 0.98941 3.26 0.32 11.8  4898 rows × 11 columns  #menggunakan grafik
	<pre>import matplotlib.pyplot as plt  plt.scatter(datasetwhiteclus_['free sulfur dioxide'], datasetwhiteclus['total sulfur dioxide']) plt.xlim(0,190) plt.show()</pre>
	200
In [29]:	x3 = dataset.iloc[:,5:7] x3.info() from sklearn.cluster import KMeans
Tn ~	<pre><class 'pandas.core.frame.dataframe'=""> RangeIndex: 1599 entries, 0 to 1598 Data columns (total 2 columns): # Column</class></pre>
In [30]:	<pre>wcss3 = [] for i in range(1,11):     kmeans3 = KMeans(n_clusters = i, init='k-means++', random_state = 42)     kmeans3.fit(x3)     wcss3.append(kmeans2.inertia_)  visualizer = KElbowVisualizer(kmeans3, k=(2,10)) plt.plot(range(1,11), wcss3)</pre>
	plt.xlabel('Number of clusters') plt.ylabel('WCSS') plt.show() visualizer.fit(x3) visualizer.show()  C:\Users\Darren\anaconda3\lib\site-packages\sklearn\cluster\_kmeans.py:1036: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=7. warnings.warn(
	620 610 800 590
	570  2 4 6 8 10  Number of clusters  Distortion Score Elbow for KMeans Clustering  700000  Distortion Score = 273399 671  020
	elbow at k = 4, score = 273399.671  500000  500000  0.16 90000  0.14 00000  0.14 00000
Out[30]:	200000 200000 2 3 4 5 6 7 8 9  Avec Subplict this less (Jeonters J. Distortion, Seere Fibrus for Manne Chartering), violed Fibrus violed Fibru
In [31]: Out[31]:	<pre>kmeans3 = KMeans(4) kmeans3.fit(x3)  identified_clusters3 = kmeans3.fit_predict(x3) identified_clusters3 array([1, 3, 0,, 0, 0, 0])</pre>
In [32]:	<pre>wine_clusters3 = datasetwhitecluscopy() wine_clusters3['Clusters'] = identified_clusters3 plt.scatter(datasetwhiteclus_['free sulfur dioxide'], datasetwhiteclus_['total sulfur dioxide'], c=wine_clusters3['Clusters'],cmap='plasma')  valueError</pre>
	File ~\anaconda3\lib\site-packages\pandas\core\frame.py:3655, in DataFramesetitem(self, key, value)  3652
	<pre>3825 () 3830    ensure homogeneity. 3831    """ -&gt; 3832    value = selfsanitize_column(value) 3834    if ( 3835</pre>
	if not self.columns.is_unique or isinstance(self.columns, MultiIndex):  File ~\anaconda3\lib\site-packages\pandas\core\frame.py:4535, in DataFramesanitize_column(self, value)  4532
In [ ]:	<pre>!jupyter nbconvertto html "./00000054804_Christopher Darren_Week7.ipynb"output-dir="./"</pre>