	Statistic for Data Science	
	Scikit-Learn	
In [1]:	Import libraries mport numpy as np	
In [2]:	import pandas as pd simport dataset iris	
In [3]:	rom sklearn.datasets import load_iris *read data	
	ris = load_iris() rint(iris.data.shape) 150, 4)	
	gumlah baris dan kolumnya: 150, dan 4 Gemengubah data frame agar lebih memudah untuk melakukan memproses data	
111 [5].	ataset_iris = load_iris(as_frame =True)	
In [6]:	ataset_iris.data.info() class 'pandas.core.frame.DataFrame'> angeIndex: 150 entries, 0 to 149	
	ata columns (total 4 columns): # Column Non-Null Count Dtype 	
	0 sepal length (cm) 150 non-null float64 1 sepal width (cm) 150 non-null float64 2 petal length (cm) 150 non-null float64 3 petal width (cm) 150 non-null float64	
In [7]:	types: float64(4) emory usage: 4.8 KB smelihat description	
Out[7]:	ataset_iris.data.describe() sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)	
	ount 150.000000 150.000000 150.000000 nean 5.843333 3.057333 3.758000 1.199333 std 0.828066 0.435866 1.765298 0.762238	
	min 4.300000 2.000000 1.000000 0.100000 25% 5.100000 2.800000 1.600000 0.300000	
	50%5.8000003.0000004.3500001.30000075%6.4000003.3000005.1000001.800000	
In [8]:	max 7.90000 4.40000 6.90000 2.50000 Frandom sampling	
Out[8]:	sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)	
	4 5.8 4.0 1.2 0.2 8 5.1 2.5 3.0 1.1 5 6.6 3.0 4.4 1.4	
In [9]:	5 6.6 3.0 4.4 1.4 trepresentative sampling	
	<pre>ataset_iris.data.sample(n=10, random_state = random.randint(0,5))</pre>	
Out[9]:	sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) 28 6.4 2.8 5.6 2.1	
	18 5.7 3.8 1.7 0.3 30 7.4 2.8 6.1 1.9	
	05 7.6 3.0 6.6 2.1 07 7.3 2.9 6.3 1.8 70 6.0 3.0 4.5 1.5	
	78 6.0 2.9 4.5 1.5 83 6.0 2.7 5.1 1.6 14 5.8 4.0 1.2 0.2	
	5 5.4 3.9 1.7 0.4 33 6.3 2.8 5.1 1.5	
In [10]:	representative sampling	
	esample yang setiap barisnya adalah unik, artinya jika ada sample dengan nilai yang sama, esample tersebut akan ditimpa dengan ke sample yang lama esebingga jumlah sample akan menjadi lebih sedikit	
Out[10]:	sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)	
	37 4.9 3.6 1.4 0.1 40 6.7 3.1 5.6 2.4 72 6.3 2.5 4.9 1.5	
	37 6.4 3.1 5.5 1.8 33 6.3 2.8 5.1 1.5	
	79 5.7 2.6 3.5 1.0 44 6.7 3.3 5.7 2.5	
	29 7.2 3.0 5.8 1.6 71 6.1 2.8 4.0 1.3 34 6.1 2.6 5.6 1.4	
	Melihat Penyebaran Data	
In [11]:	Import matlotlib	
In [12]:	mport matplotlib.pyplot as plt visualisasi	
Out[12]:	lt.scatter(dataset_iris.data['sepal length (cm)'], dataset_iris.data['sepal width (cm)'], c=dataset_iris.target) matplotlib.collections.PathCollection at 0x2381e9e6ac0>	
	1.0	
	35 -	
	2.5 -	
	4.5 5.0 5.5 6.0 6.5 7.0 7.5 8.0	
In [13]:	menambahkan funcformatter formatter = plt FuncFormatter(lambde i *arget injectorget names[int(i)])	
	<pre>formatter = plt.FuncFormatter(lambda i, *args: iris.target_names[int(i)]) ult.scatter(dataset_iris.data['sepal length (cm)'], dataset_iris.data['sepal width (cm)'], c=dataset_iris.target) ult.colorbar(ticks=[0, 1, 2], format=formatter)</pre>	
Out[13]:	lt.xlabel('sepal length (cm)') lt.ylabel('sepal width (cm)') ext(0, 0.5, 'sepal width (cm)')	
	4.5 virginica	
	3.5	
	3.5 - 3.0 -	
	2.5 - 2.0 - setosa	
In [14]:	4.5 5.0 5.5 6.0 6.5 7.0 7.5 8.0 sepal length (cm) Elakukan visualisasi terhadap Petal !!!	
Out[14]:	lt.scatter(dataset_iris.data['petal length (cm)'], dataset_iris.data['petal width (cm)'], c=dataset_iris.target) matplotlib.collections.PathCollection at 0x2381ebacc10>	
out[14].	2.5 -	
	2.0 -	
In [15]:	menambahkan funcformatter	
	formatter = plt.FuncFormatter(lambda i, *args: iris.target_names[int(i)]) lt.scatter(dataset_iris.data['petal length (cm)'], dataset_iris.data['petal width (cm)'], c=dataset_iris.target)	
	lt.colorbar(ticks=[0, 1, 2], format=formatter) lt.xlabel('petal length (cm)') lt.ylabel('petal width (cm)') lt.title('Petal Visualization')	
Out[15]:	ext(0.5, 1.0, 'Petal Visualization') Petal Visualization virginica	
	2.5 -	
	15 - 10 -	
	0.5	
	0.0 1 2 3 4 5 6 7 setosa setosa petal length (cm)	
In [16]:	testing data with Scikit-Learn	
	rom sklearn.model_selection import train_test_split _train, X_test, y_train, y_test = train_test_split(dataset_iris.data, dataset_iris.target, test_size=0.33, random_state=42)	
In [17]:	Data Model metode Logistic Regression dan Evaluasi rom sklearn.linear_model import LogisticRegression	
	rom sklearn.metrics import classification_report rom sklearn.metrics import accuracy_score dodel=LogisticRegression(solver='liblinear')	
In [18]: Out[18]:	odel=LogisticRegression(solver='liblinear') odel.fit(X_train, y_train) ogisticRegression(solver='liblinear')	
	odel.score(X_train, y_train)	
In [20]:	odel.score(X_test, y_test)	
Out[20]:	.0 Evaluation	
In [21]:	Test the model predict(X_test)	
	rint(predictions)# printing predictions rint()# Printing new Line	
	Check precision, recall, f1-score print(classification_report(y_test, predictions)) print(accuracy_score(y_test, predictions))	
	1 0 2 1 1 0 1 2 1 1 2 0 0 0 0 1 2 1 1 2 0 2 0	
	precision recall f1-score support 0 1.00 1.00 19 1 1.00 1.00 15	
	2 1.00 1.00 1.00 16 accuracy 1.00 50	
	macro avg 1.00 1.00 1.00 50 eighted avg 1.00 1.00 1.00 50 .0	
In [22]:	jupyter nbconvertto html "./00000054804_Christopher Darren_Week5_LAB_TAMBAHAN_DIKIT.ipynb"output-dir="./" NbConvertApp] Converting notebook ./00000054804_Christopher Darren_Week5.ipynb to html	
	NbConvertApp] Writing 725478 bytes to 00000054804_Christopher Darren_Week5.html	