Naive Bayes Classifier in Machine Learning (Sklearn)

Pre-Processing Data

Memanggil data

```
import pandas as pd
  csv_path='http://bit.ly/PHN-IRIS'
  data=pd.read_csv(csv_path, index_col=0)
  data.head(10)
```

Out[1]:		SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
	Id					
	1	5.1	3.5	1.4	0.2	Iris-setosa
	2	4.9	3.0	1.4	0.2	Iris-setosa
	3	4.7	3.2	1.3	0.2	Iris-setosa
	4	4.6	3.1	1.5	0.2	Iris-setosa
	5	5.0	3.6	1.4	0.2	Iris-setosa
	6	5.4	3.9	1.7	0.4	Iris-setosa
	7	4.6	3.4	1.4	0.3	Iris-setosa
	8	5.0	3.4	1.5	0.2	Iris-setosa
	9	4.4	2.9	1.4	0.2	Iris-setosa
	10	4.9	3.1	1.5	0.1	Iris-setosa

Cek missing values

```
In [2]:
         data.isna().sum()
        SepalLengthCm
Out[2]:
        SepalWidthCm
                          0
        PetalLengthCm
                          0
        PetalWidthCm
                          0
        Species
        dtype: int64
        Cek kelas yang ada
In [3]:
         data.loc[:,'Species'].unique()
        array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
Out[3]:
        Encoder kolom variety karena masih dalam bentuk kategorial
In [4]:
         from sklearn.preprocessing import LabelEncoder
```

label_encoder=LabelEncoder().fit(data['Species'])

```
data['Species']= label_encoder.transform(data['Species'])
data.head()
```

Out[4]:		SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
	ld					
	1	5.1	3.5	1.4	0.2	0
	2	4.9	3.0	1.4	0.2	0
	3	4.7	3.2	1.3	0.2	0
	4	4.6	3.1	1.5	0.2	0
	5	5.0	3.6	1.4	0.2	0

Splitting Data

Menentukan data X (response) dan y (class)

Dalam Contoh Kali ini fitur X hanya menggunakan kolom 'outlook' dan 'temp'. Sedangkan untuk kelas tetap menggunakan fitur 'play'

```
In [5]:
    X = data.drop(['Species'], axis=1)
    Y = data.Species
    X.head()
```

Out[5]: SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm ld 0.2 1 3.5 5.1 1.4 2 4.9 3.0 1.4 0.2 3 3.2 0.2 4.7 1.3 4.6 3.1 0.2 1.5 5 5.0 3.6 1.4 0.2

Teknik splitting data dengan Scikit-Learn

Training 80%, testing 20%

```
In [6]:
        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)
In [7]:
        print(X_test)
            SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
        Ιd
        147
                      6.3
                                   2.5
                                                  5.0
                                                               1.9
        99
                      5.1
                                   2.5
                                                  3.0
                                                               1.1
                      4.6
        4
                                   3.1
                                                               0.2
                                                  1.5
                      5.8
                                   2.7
                                                  3.9
                                                               1.2
        83
        20
                      5.1
                                   3.8
                                                  1.5
                                                               0.3
```

3	4.7	3.2	1.3	0.2
122	5.6	2.8	4.9	2.0
52	6.4	3.2	4.5	1.5
133	6.4	2.8	5.6	2.2
49	5.3	3.7	1.5	0.2
76	6.6	3.0	4.4	1.4
10	4.9	3.1	1.5	0.1
85	5.4	3.0	4.5	1.5
90	5.5	2.5	4.0	1.3
8	5.0	3.4	1.5	0.2
112	6.4	2.7	5.3	1.9
93	5.8	2.6	4.0	1.2
116	6.4	3.2	5.3	2.3
109	6.7	2.5	5.8	1.8
140	6.9	3.1	5.4	2.1
22	5.1	3.7	1.5	0.4
73	6.3	2.5	4.9	1.5
53	6.9	3.1	4.9	1.5
100	5.7	2.8	4.1	1.3
120	6.0	2.2	5.0	1.5
89	5.6	3.0	4.1	1.3
131	7.4	2.8	6.1	1.9
32	5.4	3.4	1.5	0.4
67	5.6	3.0	4.5	1.5
135	6.1	2.6	5.6	1.4

Training Model

```
In [8]: from sklearn.naive_bayes import MultinomialNB from sklearn.naive_bayes import GaussianNB model_NB = GaussianNB() model_NB  

Out[8]: GaussianNB()  

Naive bayes itu adalah P(A|B) = \frac{P(B|A)P(A)}{P(B)}  

In [9]: model_NB.fit(X_train,Y_train)  
Y_pred_NB=model_NB.predict(X_test)  
print(Y_pred_NB)
```

Evaluation

```
from sklearn.metrics import plot_confusion_matrix
plot_confusion_matrix(model_NB, X_test, Y_test)
```

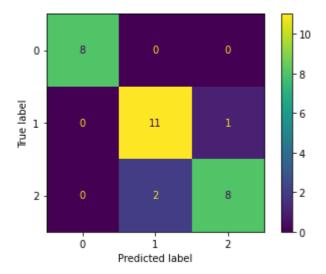
 $[2 \ 1 \ 0 \ 1 \ 0 \ 0 \ 2 \ 1 \ 2 \ 0 \ 1 \ 0 \ 1 \ 1 \ 0 \ 2 \ 1 \ 2 \ 2 \ 2 \ 0 \ 1 \ 2 \ 1 \ 1 \ 1 \ 2 \ 0 \ 1 \ 1]$

C:\Users\USER\anaconda3\envs\tensorflow\lib\site-packages\sklearn\utils\deprecation. py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_c onfusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_e stimator.

warnings.warn(msg, category=FutureWarning)

<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x259fab5b370>

Out[10]:



```
from sklearn.metrics import classification_report, confusion_matrix
print(confusion_matrix(Y_test, Y_pred_NB))
```

[[8 0 0] [0 11 1] [0 2 8]]

In [12]: print(classification_report(Y_test, Y_pred_NB))

	precision	recall	f1-score	support
0	1.00	1.00	1.00	8
1	0.85	0.92	0.88	12
2	0.89	0.80	0.84	10
accuracy			0.90	30
macro avg	0.91	0.91	0.91	30
weighted avg	0.90	0.90	0.90	30

Multinomial

```
from sklearn.naive_bayes import MultinomialNB

model_NB = MultinomialNB(alpha=1.0)
model_NB
```

Out[13]: MultinomialNB()

```
import numpy as np
model_NB.fit(X_train,Y_train)

Y_pred_NB=model_NB.predict(X_test)

print(Y_pred_NB)
print(np.array(Y_test))
```

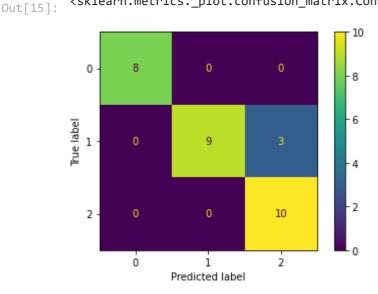
 $\begin{bmatrix} 2 & 1 & 0 & 1 & 0 & 0 & 2 & 1 & 2 & 0 & 1 & 0 & 2 & 1 & 0 & 2 & 1 & 2 & 2 & 2 & 0 & 2 & 1 & 1 & 2 & 1 & 2 & 0 & 2 & 2 \\ [2 & 1 & 0 & 1 & 0 & 0 & 2 & 1 & 2 & 0 & 1 & 0 & 1 & 1 & 0 & 2 & 1 & 2 & 2 & 2 & 0 & 1 & 1 & 1 & 2 & 1 & 2 & 0 & 1 & 2 \end{bmatrix}$

```
In [15]: plot_confusion_matrix(model_NB, X_test, Y_test)
```

C:\Users\USER\anaconda3\envs\tensorflow\lib\site-packages\sklearn\utils\deprecation. py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_c onfusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the cla ss methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_e stimator.

warnings.warn(msg, category=FutureWarning)

<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x2598065ef40>



In [16]: print(classification_report(Y_test, Y_pred_NB))

	precision	recall	f1-score	support
0	1.00	1.00	1.00	8
1	1.00	0.75	0.86	12
2	0.77	1.00	0.87	10
accuracy			0.90	30
macro avg	0.92	0.92	0.91	30
weighted avg	0.92	0.90	0.90	30

EXERCISE/ HOMEWORK

- Buatlah program klasifikasi diabetes menggunakan model Naive Bayes. Data dapat diunduh pada tautan berikut ini https://github.com/dhirajk100/Naive-Bayes/blob/master/Naive-Bayes-Classification-Data.csv
- 1. Buatlah program klasifikasi Social Media Analysis menggunakan model Naive Bayes. Data dapat diunduh pada tautan berikut ini https://www.kaggle.com/rakeshrau/social-network-ads
- 1. Gunakan fungsi Naive Bayes lainnya selain Gaussian dan Multinomial NB. Lihat https://scikit-learn.org/stable/modules/naive_bayes.html
- 1. Buatlah program klasifikasi diabetes menggunakan model Naive Bayes. Data dapat diunduh pada tautan berikut ini

Pre-Processing Data

Memanggil data

```
import pandas as pd
  data=pd.read_csv('Naive-Bayes-Classification-Data.csv')
  data
```

```
Out[17]:
                 glucose bloodpressure diabetes
             0
                                                0
                      40
                                     85
              1
                      40
                                     92
                                                0
             2
                      45
                                     63
                                                1
             3
                      45
                                     80
                                                0
              4
                      40
                                     73
                                                1
           990
                      45
                                     87
                                                0
                      40
                                                0
           991
                                     83
                                                0
           992
                      40
                                     83
           993
                      40
                                     60
                                                1
                      45
                                                0
           994
                                     82
```

995 rows × 3 columns

array([0, 1], dtype=int64)

Cek missing values

```
In [18]:
         data.isna().sum()
         glucose
                        0
Out[18]:
         bloodpressure
                        0
         diabetes
                        0
         dtype: int64
In [19]:
         data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 995 entries, 0 to 994
         Data columns (total 3 columns):
            Column Non-Null Count Dtype
         #
         ---
                           -----
             glucose 995 non-null int64
         1
             bloodpressure 995 non-null int64
         2
             diabetes
                           995 non-null
                                          int64
        dtypes: int64(3)
        memory usage: 23.4 KB
        Cek kelas yang ada
In [20]:
         data.loc[:,'diabetes'].unique()
```

```
Out[20]:
```

Splitting Data

Menentukan data X (response) dan y (class)

```
In [21]:     X = data.drop(['diabetes'], axis=1)
     Y = data.diabetes
     X.head()
```

```
Out[21]:
              glucose bloodpressure
                                  85
           0
                   40
           1
                   40
                                  92
           2
                                  63
                   45
           3
                   45
                                  80
           4
                   40
                                  73
```

Teknik splitting data dengan Scikit-Learn

Training 80%, testing 20%

```
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)
```

```
In [23]: print(X_test)
```

	glucose	bloodpressure
438	55	70
836	40	85
608	45	73
759	60	75
453	45	90
• •		• • •
146	50	77
446	50	90
589	50	83
756	55	67
762	40	93

[199 rows x 2 columns]

Training Model

```
from sklearn.naive_bayes import MultinomialNB
from sklearn.naive_bayes import GaussianNB

model_NB = GaussianNB()
model_NB
```

Out[24]: GaussianNB()

```
Naive bayes itu adalah P(A|B) = rac{P(B|A)P(A)}{P(B)}
```

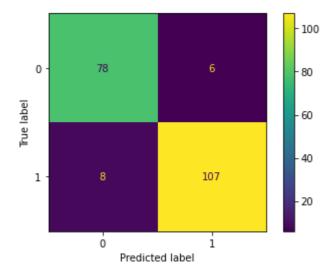
Evaluation

In [26]: from sklearn.metrics import plot_confusion_matrix
plot_confusion_matrix(model_NB, X_test, Y_test)

C:\Users\USER\anaconda3\envs\tensorflow\lib\site-packages\sklearn\utils\deprecation. py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_c onfusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_e stimator.

warnings.warn(msg, category=FutureWarning)

Out[26]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x259807d9490>



from sklearn.metrics import classification_report, confusion_matrix
print(confusion_matrix(Y_test, Y_pred_NB))

[[78 6] [8 107]]

In [28]: print(classification_report(Y_test, Y_pred_NB))

	precision	recall	f1-score	support
0	0.91	0.93	0.92	84
1	0.95	0.93	0.94	115
accuracy			0.93	199
macro avg	0.93	0.93	0.93	199
weighted avg	0.93	0.93	0.93	199

ComplementNB

```
In [29]:
      from sklearn.naive bayes import ComplementNB
      model NB = ComplementNB()
      model NB
     ComplementNB()
Out[29]:
In [30]:
      import numpy as np
      model_NB.fit(X_train,Y_train)
      Y_pred_NB=model_NB.predict(X_test)
      print(Y_pred_NB)
      print(np.array(Y_test))
     00011110110110]
     [1\ 0\ 1\ 1\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 0\ 1\ 1\ 0
      0 0 0 1 1 1 1 1 1 1 0 0 1 0]
In [31]:
      plot_confusion_matrix(model_NB, X_test, Y_test)
     C:\Users\USER\anaconda3\envs\tensorflow\lib\site-packages\sklearn\utils\deprecation.
     py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_c
     onfusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the cla
     ss methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_e
     stimator.
       warnings.warn(msg, category=FutureWarning)
     <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x259fcf10190>
Out[31]:
                              80
                              70
       0
            79
                              60
                              50
     Frue labe
                              40
```

```
In [32]: print(classification_report(Y_test, Y_pred_NB))
```

- 30

20

10

precision recall f1-score support

82

1

1

0

Predicted label

0	0.71	0.94	0.81	84
1	0.94	0.71	0.81	115
accuracy			0.81	199
macro avg	0.82	0.83	0.81	199
weighted avg	0.84	0.81	0.81	199

2. Buatlah program klasifikasi Social Media Analysis menggunakan model Naive Bayes. Data dapat diunduh pada tautan berikut ini

https://www.kaggle.com/rakeshrau/social-network-ads

Pre-Processing Data

Memanggil data

```
import pandas as pd
  data=pd.read_csv('Social_Network_Ads.csv', index_col=0)
  data.head(10)
```

Out[33]: Gender Age EstimatedSalary Purchased

User ID				
15624510	Male	19	19000	0
15810944	Male	35	20000	0
15668575	Female	26	43000	0
15603246	Female	27	57000	0
15804002	Male	19	76000	0
15728773	Male	27	58000	0
15598044	Female	27	84000	0
15694829	Female	32	150000	1
15600575	Male	25	33000	0
15727311	Female	35	65000	0

Cek missing values

<class 'pandas.core.frame.DataFrame'>
Int64Index: 400 entries, 15624510 to 15594041

```
Data columns (total 4 columns):

# Column Non-Null Count Dtype
--- 0 Gender 400 non-null object
1 Age 400 non-null int64
2 EstimatedSalary 400 non-null int64
3 Purchased 400 non-null int64
dtypes: int64(3), object(1)
memory usage: 15.6+ KB
```

Cek kelas yang ada

```
In [36]:
    data.loc[:,'Purchased'].unique()
```

Out[36]: array([0, 1], dtype=int64)

Encoder kolom gender karena masih dalam bentuk object

```
In [37]:
    from sklearn.preprocessing import LabelEncoder
    label_encoder=LabelEncoder().fit(data['Gender'])
    data['Gender']= label_encoder.transform(data['Gender'])
    data.head()
```

Out[37]: Gender Age EstimatedSalary Purchased

User ID				
15624510	1	19	19000	0
15810944	1	35	20000	0
15668575	0	26	43000	0
15603246	0	27	57000	0
15804002	1	19	76000	0

Splitting Data

Menentukan data X (response) dan y (class)

```
In [38]: X = data.drop(['Purchased'], axis=1)
    Y = data.Purchased
    X.head()
```

Out[38]: Gender Age EstimatedSalary

User ID			
15624510	1	19	19000
15810944	1	35	20000
15668575	0	26	43000
15603246	0	27	57000
15804002	1	19	76000

Teknik splitting data dengan Scikit-Learn

```
In [39]:
         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)
In [40]:
         print(X_test)
                 Gender Age EstimatedSalary
        User ID
                    1 40
        15794493
                                      57000
                    1 34
        15570932
                                     115000
                    0 33
        15571059
                                     41000
        15697997
                    0 38
                                     80000
                    1 37
        15741049
                                     72000
                                       . . .
        . . .
                  ...
        15591433 1 36
                                     52000
                    1 39
                                     71000
        15627220
        15768072
                    0 47
                                     50000
        15594577
                    1 25
                                     22000
                  1 31
                                    18000
        15669656
        [80 rows x 3 columns]
        Training Model
In [41]:
         from sklearn.naive_bayes import MultinomialNB
         from sklearn.naive_bayes import GaussianNB
         model_NB = GaussianNB()
         model NB
        GaussianNB()
Out[41]:
       Naive bayes itu adalah P(A|B) = \frac{P(B|A)P(A)}{P(B)}
In [42]:
        model_NB.fit(X_train,Y_train)
         Y_pred_NB=model_NB.predict(X_test)
         print(Y_pred_NB)
        [0\ 1\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 0\ 0\ 1\ 0\ 0\ 1\ 1\ 1\ 0\ 0\ 0\ 1\ 1\ 1\ 0\ 0\ 0\ 0
```

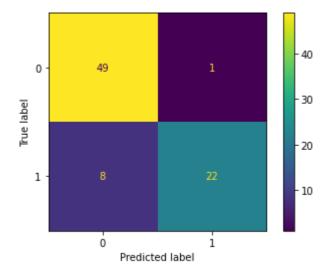
Evaluation

1 0 0 0 0 0]

```
from sklearn.metrics import plot_confusion_matrix
plot_confusion_matrix(model_NB, X_test, Y_test)
```

C:\Users\USER\anaconda3\envs\tensorflow\lib\site-packages\sklearn\utils\deprecation. py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_c onfusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_e stimator.

warnings.warn(msg, category=FutureWarning)



```
from sklearn.metrics import classification_report, confusion_matrix
print(confusion_matrix(Y_test, Y_pred_NB))
```

[[49 1] [8 22]]

In [45]: print(classification_report(Y_test, Y_pred_NB))

	precision	recall	f1-score	support
0	0.86	0.98	0.92	50
1	0.96	0.73	0.83	30
accuracy			0.89	80
macro avg	0.91	0.86	0.87	80
weighted avg	0.90	0.89	0.88	80

ComplementNB

```
from sklearn.naive_bayes import ComplementNB
model_NB = ComplementNB()
model_NB
```

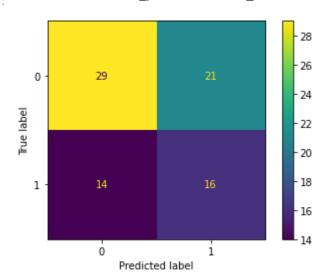
Out[46]: ComplementNB()

In [48]: plot_confusion_matrix(model_NB, X_test, Y_test)

C:\Users\USER\anaconda3\envs\tensorflow\lib\site-packages\sklearn\utils\deprecation. py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_c onfusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the cla ss methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_e stimator.

warnings.warn(msg, category=FutureWarning)

Out[48]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x259807d98b0>



In [49]: print(classification_report(Y_test, Y_pred_NB))

	precision	recall	f1-score	support
0	0.67	0.58	0.62	50
1	0.43	0.53	0.48	30
accuracy			0.56	80
macro avg	0.55	0.56	0.55	80
weighted avg	0.58	0.56	0.57	80