Tugas Pemodelan dan Simulasi

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▼ Logistic Regression

Dataset berisi 8 kolom atribut dan 1 kolom label yang berisi 2 kelas yaitu 1 dan 0. Angka 1 menandakan bahwa orang tersebut positif diabetes dan 0 menandakan sebaliknya. Terdapat 768 sampel yang merupakan 768 pasien perempuan keturunan suku Indian Pima.

Model machine learning yang akan kita buat bertujuan untuk mengklasifikasikan apakah seorang pasien positif diabetes atau tidak

Tahapan latihan kali ini adalah:

- 1. Import data dan ubah data ke dalam Dataframe.
- 2. Bagi dataset.
- 3. Melakukan standarisasi.
- 4. Membuat dan melatih model.
- 5. Evaluasi model.

```
import pandas as pd
import numpy as np
```

df = pd.read_csv('/content/sample_data/diabetes.csv')

df.head()

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigre
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
A	0	127	40	25	160	12 1	>

#mengecek nilai yang missing value
df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
```

#	Column	Non-Null Count	Dtype
0	Pregnancies	768 non-null	int64
1	Glucose	768 non-null	int64
2	BloodPressure	768 non-null	int64

```
3
     SkinThickness
                                768 non-null
                                                 int64
                                768 non-null
                                                 int64
4
     Insulin
                                                 float64
5
                                768 non-null
     DiabetesPedigreeFunction
                                768 non-null
                                                 float64
6
7
                                768 non-null
                                                 int64
     Outcome
                                768 non-null
                                                 int64
dtypes: float64(2), int64(7)
```

memory usage: 54.1 KB

```
#Replace nilai Nol
import numpy as np
zero_not_allowed = ['Glucose', 'BloodPressure','SkinThickness','Insulin', 'BMI']
for column in zero_not_allowed:
    df[column] = df[column].replace(0, np.NaN)
    mean = int(df[column].mean(skipna=True))
    df[column] = df[column].replace(np.NaN, mean)
```

df.head()

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigre
0	6	148.0	72.0	35.0	155.0	33.6	
1	1	85.0	66.0	29.0	155.0	26.6	
2	8	183.0	64.0	29.0	155.0	23.3	
3	1	89.0	66.0	23.0	94.0	28.1	
^	Λ.	127 N	40 O	35 N	16 <u>2</u> 0	12 1	•

```
# memisahkan atribut pada dataset dan menyimpannya pada sebuah variabel
X = df[df.columns[:8]]
# memisahkan label pada dataset dan menyimpannya pada sebuah variabel
y = df['Outcome']
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import confusion_matrix, classification_report
# standarisasi nilai-nilai dari dataset
scaler = StandardScaler()
scaler.fit(X)
X = scaler.transform(X)
from sklearn.model selection import train test split
# memisahkan 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)
from sklearn import linear_model
# latih model dengan fungsi fit
model = linear model.LogisticRegression() #membuat sebuah objek logistic regression.
model.fit(X train, y train)
     ▼ LogisticRegression
```

LogisticRegression()

```
model.score(X_test, y_test)
     0.7532467532467533
cm = confusion_matrix(y_test, y_pred)
\mathsf{cm}
     array([[70, 29],
            [15, 40]])
print(classification report(y test, y pred))
                                  recall f1-score
                    precision
                                                      support
                 0
                         0.82
                                    0.71
                                               0.76
                                                            99
                 1
                         0.58
                                    0.73
                                               0.65
                                                            55
         accuracy
                                               0.71
                                                          154
                         0.70
                                    0.72
                                               0.70
                                                          154
        macro avg
                                               0.72
                                                          154
     weighted avg
                         0.74
                                    0.71
```

→ k-NN

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import confusion matrix
from sklearn.metrics import f1 score
from sklearn.metrics import accuracy_score
# Define the model --> K-NN, value K got from square root leng y test, K=11
classifire = KNeighborsClassifier(n neighbors=11, metric='euclidean')
#fit model
classifire.fit(X_train, y_train)
                        KNeighborsClassifier
     KNeighborsClassifier(metric='euclidean', n_neighbors=11)
# Tes hasil Predikksi
y pred = classifire.predict(X test)
y_pred
     array([1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0,
            1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0,
            0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0,
            0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0,
            0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1,
            0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1,
            0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0])
# Evaluate the model
cm = confusion_matrix(y_test, y_pred)
cm
     array([[81, 18],
            [20, 35]])
```

print(classification_report(y_test, y_pred))

	precision	recall	f1-score	support
0	0.82	0.71	0.76	99
	0.58	0.73	0.65	55
accuracy			0.71	154
macro avg	0.70	0.72	0.70	154
weighted avg	0.74	0.71	0.72	154

→ Random Forest

	precision	recall	f1-score	support
0	0.82	0.71	0.76	99
1	0.58	0.73	0.65	55
accuracy			0.71	154
macro avg	0.70	0.72	0.70	154
weighted avg	0.74	0.71	0.72	154

→ Decision Tree

```
from sklearn.tree import DecisionTreeClassifier
clf = DecisionTreeClassifier(max_depth=4)
```

clf.fit(X_train, y_train)

```
DecisionTreeClassifier
DecisionTreeClassifier(max_depth=4)
```

```
clf.score(X_train, y_train)
```

0.8045602605863192

print(classification_report(y_test, y_pred))

	precision	recall	f1-score	support
0 1	0.82 0.58	0.71 0.73	0.76 0.65	99 55
accuracy macro avg weighted avg	0.70 0.74	0.72 0.71	0.71 0.70 0.72	154 154 154

Dapat diambil kesimpulan dari Evaluasi Model:

▼ Visualize Decision Tree

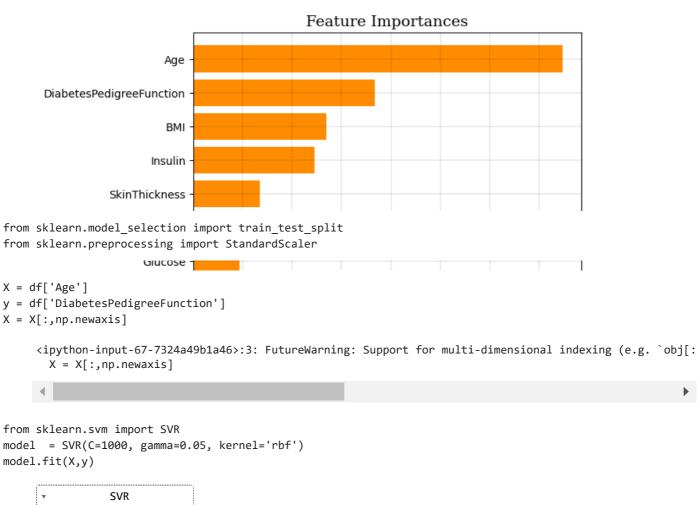
```
from sklearn.tree import DecisionTreeClassifier
from sklearn import tree
from sklearn.metrics import accuracy_score
from sklearn.metrics import confusion_matrix

# Implementasi decision tree menggunakan scikit-learn

model = DecisionTreeClassifier(criterion = "gini", max_depth = 10)
model = model.fit(X_train, y_train)

# Identifikasi feature
import matplotlib.pyplot as plt
importance = model.feature_importances_
indices = np.argsort(importance)

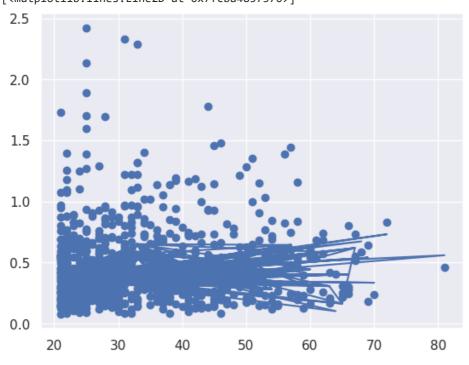
importance_plt = plt.barh(list(df.columns[: -1]), importance[indices], color = "darkorange")
importance_plt = plt.title("Feature Importances", fontsize = 14, fontfamily = "serif")
importance_plt = plt.grid(color='black', linewidth=0.1)
```



▼ SVR SVR(C=1000, gamma=0.05)

plt.scatter(X, y)
plt.plot(X, model.predict(X))

[<matplotlib.lines.Line2D at 0x7fcba4857370>]



Dapat dilihat kecendurangan umur yang menyebabkan seseorang bisa terkena diabetes adalah kisaran 20 - 50 tahun. Umur seseorang dapat mempengaruhi kecenderungan untuk pasien memiliki diabetes.

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