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Self Learning

1. Import semua library yang diperlukan

```
#Library
import pandas as pd
import numpy as np
from scipy import stats
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import StandardScaler
from sklearn.feature_selection import SelectKBest, f_classif
from sklearn.model_selection import train_test_split
import time
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
from sklearn.decomposition import PCA
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
import matplotlib.pyplot as plt
%matplotlib inline
```

2. Membuka dataset skor_toefl.csv

```
toefl = pd.read_csv('skor_toefl.csv')
    toefl
\Box
         Sebelum Sesudah 🎢
     0
             429
                      569
     1
             443
                      599
     2
             408
                      612
     3
             459
                      536
             401
                      572
             496
                      552
     94
     95
             413
                      544
             400
                      531
     96
     97
             471
                      561
     98
             443
                      588
    99 rows × 2 columns
```

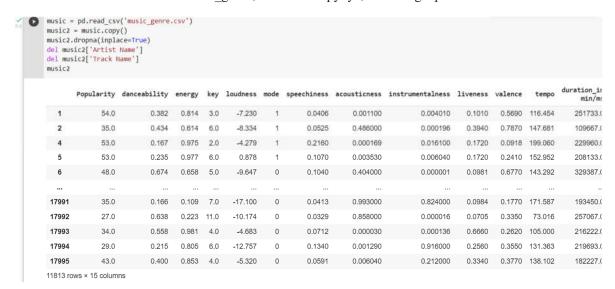
3. Menghitung **uji t**

```
#Melakukan uji t
uji_t = stats.ttest_rel(toefl['Sesudah'], toefl['Sebelum']) #hasilnya: (statistik hitungnya (t-hitung), pvalue)
print('nilai t-hitung = ', uji_t[0])
print('nilai p-value = ', uji_t[1])

nilai t-hitung = 30.255038012916643
nilai p-value = 1.638101871655264e-51
```



4. Membaca dataset music genre, membuat copynya, dan menghapus artist name dan track name



5. Memisahkan feature dan label

```
#Memisahkan Feature dan Label
#Feature
X = music2.iloc[:,:14].values
Y = music2.iloc[:, 14].values
```

6. Melakukan transform data

```
scaler = StandardScaler()
 # transform data
 X_scaled = scaler.fit_transform(X)
 print(X_scaled)
 [[ 0.6578634 -0.86108
                           0.59301271 ... -0.22000411 0.33481231
   0.22582345]
 [-0.48893955 -0.54783734 -0.2323635 ... 0.8410691 -0.89146753
   0.225823451
 [ 0.59750535 -2.15621793 1.25744056 ... 2.5868942
                                                      0.14687296
   0.225823451
 [-0.5492976
              0.19912594 1.28220184 ... -0.6092036
                                                      0.02828982
   0.22582345]
 [-0.85108785 -1.86707086 0.55587078 ... 0.28659404 0.05825067
   0.22582345]
 [-0.00607515 -0.75264985 0.75396107 ... 0.51558089 -0.26514691
   0.22582345]]
```

7. Memilih **k feature** yang akan digunakan, k = 8

```
#Ketik kodingannya disini, silahkan memilih berapa k feature yang mau digunakan selector = SelectKBest(f_classif, k=8) selector.fit(X_scaled, Y)

X_new_a = selector.fit_transform(X_scaled, Y)

print("nilai statistik hitung tiap feature : ", selector.scores_)

print('\n', "nilai p value : ", selector.pvalues_)

C. nilai statistik hitung tiap feature : [139.88066056 327.57607032 607.05627169 3.44940318 533.23015937 17.36126483 240.43782283 702.69800751 307.34168538 25.06319532 208.2249501 28.89657946 931.21961423 25.14183356]

nilai p value : [5.67588919e-278 0.00000000e+000 0.00000000e+000 1.54594261e-004 0.00000000e+000 8.80419429e-032 0.00000000e+000 0.00000000e+000 1.54594261e-004 0.00000000e+000 1.36713016e-047 0.00000000e+000 1.72827268e-055 0.000000000e+000 9.41651367e-048]
```



8. Membuat visualisasi cumulative explained ratio

```
#Ketik kodingan disini
    pca = PCA()
    pca_data = pca.fit_transform(X_scaled)
    plt.plot(np.cumsum(pca.explained_variance_ratio_))
    plt.xlabel("number of components")
    plt.ylabel("cumulative explained variance")
Text(0, 0.5, 'cumulative explained variance')
       1.0
       09
       0.8
     D 0.7
       0.6
     0.5 0.4 0.3 0.3
       0.2
                                                   12
                          number of components
```

9. Untuk Cumulative explained variance minimal 75% maka disini dimasukkan n = 9.

```
# Ketik kodingannya disini

n = 9

pca = PCA(n_components = n)

X_new_pca = pca.fit_transform(X_scaled)

total_var = pca.explained_variance_ratio_.sum() * 100

print(total_var, '% of total variance is explained by',n, 'principal components')

P. 82.70526081543656 % of total variance is explained by 9 principal components
```

10. Berikut adalah visualisasi dari model LDA



11. Untuk mendapatkan **cumulative explained variance** minimal 75%, maka dimasukkan 1 = 3

12. Pada code dibawah ini, adalah code untuk train model dan predict menggunakan data asli

```
#Train Model and Predict menggunakan data asli

Xmusix - X_scaled

Ymusix = Y

Xmusix_train, Xmusix_test, Ymusix_train, Ymusix_test = train_test_split(Xmusix, Ymusix, test_size = 0.2, random_state = 22, stratify = Ymusix)

start = time.time()

knn = KNelghborsClassifier(n_neighbors = 9).fit(Xmusix_train, Ymusix_train)

Ypred = knn.predict(Xmusix_test)

end = time.time()

print("kakurasi : ", metrics.accuracy_score(Ymusix_test, Ypred))

print("waktu : ", end-start)

Akurasi : 0.4265763859500635

waktu : 0.5650367736816406
```

13. Selanjutnya, pada code ini adalah code untuk train model dan predict menggunakan data dengan feature selection



14. Pada code ini juga, adalah kode untuk train model dan predict menggunakan data dengan PCA

```
Instant Model and Predict menggunakan data dengan PCA
    Xmusix = X_scaled
    Ymusix = Y
    Xmusix = X_scaled
    Ymusix = Y
    Xmusix_train, Xmusix_test, Ymusix_train, Ymusix_test = train_test_split(Xmusix, Ymusix, test_size = 0.2, random_state = 22, stratify = Ymusix)

    start = time.time()
    n = 8
    pca = PCA(n_components = n)
    X_new_pca = pca.fit_transform(X_scaled)
    total_var = pca.explained_variance_ratio_.sum() * 100
    print(total_var, '% of total variance is explained by',n, 'principal components')
    end = time.time()
    print("Akurasi : ", metrics.accuracy_score(Ymusix_test, Ypred))
    print("waktu : ", end-start)

76.68715479081389 % of total variance is explained by 8 principal components
    Akurasi : 0.4265763859500635
    waktu : 0.83221845626831055
```

15. Kemudian untuk code dibawah ini adalah code untuk train model dan predict menggunakan data dengan LDA

```
#Train Model and Predict menggunakan data dengan LDA

Xmusix = X_scaled

Ymusix = Y

Xmusix_train, Xmusix_test, Ymusix_train, Ymusix_test = train_test_split(Xmusix, Ymusix, test_size = 0.2, random_state = 22, stratify = Ymusix)

start = time.time()

1 = 4

1da = LinearDiscriminantAnalysis(n_components=1)

X_naw_lda = lda.fit_transform(X_scaled, Y)

total_var = lda.explained_variance_ratio_.sum() * 100

print(total_var, '% of total variance is expalained by', 1, 'features')
end = time.time()

print("Akurasi : ", metrics.accuracy_score(Ymusix_test, Ypred))
print("waktu : ", end-start)

1- 91.36003363150662 % of total variance is expalained by 4 features

Akurasi : 0.42657638595096035

waktu : 0.058984994888305664
```

Kesimpulannya adalah akurasi yang dihasilkan tetaplah sama, namun pada lama waktu setiap model berbeda – beda

Untuk source code, bisa di lihat pada link berikut : https://github.com/Muhamad-naufal/Project-MSIB-3/blob/main/Exercise 19.ipynb