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## Bagi data menjadi target dan data dari fitur

```
In [34]: split = np.split(Final, [15], axis=0)
x = split[0].values
y = split[1].values.flatten()
print(x)
```

## Proses machine learning

```
In [28]: k = 9
loo = LeaveOneOut()
loo.get_n_splits(x)
scores_list = []
y_test_list = []
y_pred_list = []
for i, j in loo.split(x):
    print("TRAIN: ", i, "TEST: ", j)
    X_train, X_test = x[i], x[j]
    y_train, y_test = y[i], y[j]
    print(X_train, X_test, y_train, y_test)
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(X_train, y_train)
    y_pred = knn.predict(X_test)
    scores = metrics.accuracy_score(y_test, y_pred)
    scores_list.append(scores)
    y_test_list.append(y_test)
    y_pred_list.append(y_pred)

df = pd.DataFrame({'Akurasi': scores_list, 'Actual': y_test_list, 'Prediksi': y_pred_list})
df.to_excel('nama_file', index=False)
```

## proses menampilkan confusion matrix, akurasi, precision, recall

```
In [44]: df = pd.read_excel('nama_file', delimiter=',')
Neh = []
Fairh = []
Eh = []
Heh = []
Owh = []
akurasi = []
Hasil = {}
Neh.append(df[(df['Prediksi'] == 'Neh') & (df['Actual'] == 'Neh')].count().mean())
Fairh.append(df[(df['Prediksi'] == 'Fairh') & (df['Actual'] == 'Neh')].count().mean())
Eh.append(df[(df['Prediksi'] == 'Eh') & (df['Actual'] == 'Neh')].count().mean())
Heh.append(df[(df['Prediksi'] == 'Heh') & (df['Actual'] == 'Neh')].count().mean())
Neh.append(df[(df['Prediksi'] == 'Owh') & (df['Actual'] == 'Neh')].count().mean())
Hasil.update({'Neh' : Neh})
Fairh.append(df[(df['Prediksi'] == 'Neh') & (df['Actual'] == 'Fairh')].count().mean())
Fairh.append(df[(df['Prediksi'] == 'Fairh') & (df['Actual'] == 'Fairh')].count().mean())
Eh.append(df[(df['Prediksi'] == 'Eh') & (df['Actual'] == 'Fairh')].count().mean())
Fairh.append(df[(df['Prediksi'] == 'Heh') & (df['Actual'] == 'Fairh')].count().mean())
Fairh.append(df[(df['Prediksi'] == 'Owh') & (df['Actual'] == 'Fairh')].count().mean())
Hasil.update({'Fairh' : Fairh})
Eh.append(df[(df['Prediksi'] == 'Neh') & (df['Actual'] == 'Eh')].count().mean())
Eh.append(df[(df['Prediksi'] == 'Fairh') & (df['Actual'] == 'Eh')].count().mean())
Eh.append(df[(df['Prediksi'] == 'Eh') & (df['Actual'] == 'Eh')].count().mean())
Eh.append(df[(df['Prediksi'] == 'Heh') & (df['Actual'] == 'Eh')].count().mean())
Eh.append(df[(df['Prediksi'] == 'Owh') & (df['Actual'] == 'Eh')].count().mean())
Hasil.update({'Eh' : Eh})
Heh.append(df[(df['Prediksi'] == 'Neh') & (df['Actual'] == 'Heh')].count().mean())
Heh.append(df[(df['Prediksi'] == 'Fairh') & (df['Actual'] == 'Heh')].count().mean())
Heh.append(df[(df['Prediksi'] == 'Eh') & (df['Actual'] == 'Heh')].count().mean())
Heh.append(df[(df['Prediksi'] == 'Heh') & (df['Actual'] == 'Heh')].count().mean())
Heh.append(df[(df['Prediksi'] == 'Owh') & (df['Actual'] == 'Heh')].count().mean())
Hasil.update({'Heh' : Heh})
Owh.append(df[(df['Prediksi'] == 'Neh') & (df['Actual'] == 'Owh')].count().mean())
Owh.append(df[(df['Prediksi'] == 'Fairh') & (df['Actual'] == 'Owh')].count().mean())
Owh.append(df[(df['Prediksi'] == 'Eh') & (df['Actual'] == 'Owh')].count().mean())
Owh.append(df[(df['Prediksi'] == 'Heh') & (df['Actual'] == 'Owh')].count().mean())
Owh.append(df[(df['Prediksi'] == 'Owh') & (df['Actual'] == 'Owh')].count().mean())
Hasil.update({'Owh' : Owh})
df2 = pd.DataFrame(Hasil, index=['Neh', 'Fairh', 'Eh', 'Heh', 'Owh'])
print("Hasil Akurasi Suara Tangisan Bayi")
print(df2)
```



```

Neh = 0 if (df[(df['Actual'] == '['Neh'])].count().mean()) == 0 else(df[(df['Prediksi'] == '['Neh']) & (df['Actual'] == '['Neh')
akurasi.append(Neh)
Eairh = 0 if (df[(df['Actual'] == '['Eairh'])].count().mean()) == 0 else(df[(df['Prediksi'] == '['Eairh']) & (df['Actual'] == '['Eairh')
akurasi.append(Eairh)
Eh = 0 if (df[(df['Actual'] == '['Eh'])].count().mean()) == 0 else(df[(df['Prediksi'] == '['Eh']) & (df['Actual'] == '['Eh')
akurasi.append(Eh)
Heh = 0 if (df[(df['Actual'] == '['Heh'])].count().mean()) == 0 else (df[(df['Prediksi'] == '['Heh']) & (df['Actual'] == '['Heh')
akurasi.append(Heh)
Owh = 0 if (df[(df['Actual'] == '['Owh'])].count().mean()) == 0 else (df[(df['Prediksi'] == '['Owh']) & (df['Actual'] == '['Owh')
akurasi.append(Owh)
print('\n')
ulang = 5
label = ['Neh', "Eairh", "Eh", "Heh", "Owh"]
label_fix = []
for i in range(ulang):
    #print(akurasi[i])
    if akurasi[i] != 0:
        print("Akurasi "+str(label[i])+" = "+str(akurasi[i]))
        label_fix.append(label[i])

print('\n')
#print(label_fix)
total = (akurasi[0]+akurasi[1]+akurasi[2]+akurasi[3]+akurasi[4])/len(label_fix)
print("Hasil Akurasinya = "+str(total)+"%")
a = df['Actual'].values
b = df['Prediksi'].values
print(classification_report(a, b))

```

Lampiran 3 : Source Code K-Nearest Neighbours dengan data sampling Percentage Rate

### Import library

```

In [3]: %matplotlib inline
import warnings
warnings.filterwarnings('ignore')
import matplotlib.pyplot as plt
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix, classification_report
import numpy as np
import pandas as pd

```

### proses load data

```

In [105]: final = pd.read_csv('Sumber/Final-VI.csv', delimiter=',')
nama = 'Hasil/Final-VI-ts0.3-k1.xls'

In [106]: split = np.split(final, [15], axis=1)
x = split[0]
y = split[1]

```

### proses machine learning

```

In [107]: k = 1
tsize=0.3
scores_list = []
y_test_list = []
y_pred_list = []
for i in range(0,150):
    X_train,X_test,y_train,y_test = train_test_split(x,y,test_size=tsize,random_state=i)
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(X_train,y_train)
    y_pred=knn.predict(X_test)
    score=metrics.accuracy_score(y_test,y_pred)
    scores_list.append(score)
    y_test_list.append(y_test)
    y_pred_list.append(y_pred)
    print("Tabel Confussion Matrix nilai random state = "+str(i)+" dan akurasi = "+str(score))
    print(confusion_matrix(y_test, y_pred, labels=['Neh', "Eairh", "Eh", "Heh", "Owh"]))
    print(classification_report(y_test, y_pred))

df = pd.DataFrame({'akurasi': scores_list, 'Actual': y_test_list, 'Prediksi': y_pred_list})
df.to_excel(nama,index=True)

```

## Lampiran 4: Source Code K-Means

### import Library

```
In [1]: %matplotlib inline
import pandas as pd
import numpy as np
import sklearn as sm
from sklearn.cluster import KMeans
from mpl_toolkits.mplot3d import Axes3D
from sklearn.preprocessing import scale
from sklearn import datasets
from sklearn.metrics import confusion_matrix, classification_report
import matplotlib.pyplot as plt
from sklearn import metrics
from sklearn.metrics import accuracy_score
```

### Load dataset

```
In [4]: FinalV1 = pd.read_csv('Sumber/Final-V1.csv', delimiter=',')
nama_file = 'Final_Kmeans.xls'
```

```
In [5]: split = np.split(FinalV1, [15], axis=1)
xV1 = split[0].values
yV1 = split[1].values.flatten()
#print(xV1)
#print(yV1)
```

### Proses Kmeans ¶

```
In [1]: cluster = KMeans(n_clusters=5, random_state=0)
cluster.fit(xV1)
```

### simpan menjadi file excel ¶

```
In [7]: list = pd.Series(cluster.labels_)
FinalV1['cluster'] = list
FinalV1.to_excel(nama_file)
```