Activity-5: Machine Learning

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Dataset Name:Cryotherapy Dataset Data Set, NBA Players Stats - 2014-2015, FIFA Season 17-18 Data Visualisation | Clustering | Machine Learning

Dataset url: https://archive.ics.uci.edu/ml/datasets/Cryotherapy+Dataset+ (https://www.kaggle.com/drgilermo/nba-players-stats-20142015 (https://www.kaggle.com/oliversinn/fifa-18-visualisation-clustering-ml/data (https://www.kaggle.com/oliversinn/fifa-18-visualisation-clustering-ml/data

Part-1: Regression

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In [12]:
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```
# -*- coding: utf-8 -*-
Created on Sat May 16 14:32:34 2020
@author: Onur
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
veriler = pd.read_csv('players_stats.csv')
pos = veriler.iloc[:,30:31]
fill = veriler[['Age', 'Height', 'Weight', 'BMI']]
pts = veriler.iloc[:,3:4]
data1 = veriler.iloc[:,1:3]
data2 = veriler.iloc[:,4:24]
from sklearn.impute import SimpleImputer
imputer = SimpleImputer(missing_values=np.nan, strategy='most_frequent')
imputer = imputer.fit(pos)
pos = imputer.transform(pos)
numeric = SimpleImputer(missing_values=np.nan, strategy='mean')
numeric = numeric.fit(fill)
fill = numeric.transform(fill)
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
pos = le.fit_transform(pos)
fill = pd.DataFrame(fill)
pos = pd.DataFrame(pos)
fill.columns = ["Age", "Height", "Weight", "BMI"]
pos.columns = ["Pos"]
lastdataCorr = pd.concat([data1,data2,fill,pos,pts],axis=1)
corr_matrix = lastdataCorr.corr()
print(corr_matrix)
last_data = pd.concat([data1,data2,fill,pos],axis=1)
last_data = last_data.values
pts = pts.values
from sklearn.model_selection import train_test_split
x_train, x_test,y_train,y_test = train_test_split(last_data,pts,test_size=0.33, random_stat
from sklearn.linear_model import LinearRegression
linear = LinearRegression()
linear.fit(x_train,y_train)
tahmin = linear.predict(x_test)
from sklearn.metrics import r2 score
```

```
print('R2 Lineer : ',r2_score(y_test,tahmin))

from sklearn.preprocessing import PolynomialFeatures
Polynomial = PolynomialFeatures(degree=2)
x_poly = Polynomial.fit_transform(x_train)
lineerPoly = LinearRegression()
lineerPoly.fit(x_poly,y_train)
tahminPoly = lineerPoly.predict(Polynomial.fit_transform(x_test))
print('R2 Polynomial : ',r2_score(y_test,tahminPoly))

from sklearn.ensemble import RandomForestRegressor
RandomForest = RandomForestRegressor(n_estimators=10,random_state=0)
RandomForest.fit(x_train,y_train)
tahminRandomF = RandomForest.predict(x_test)
print('R2 RandomForest:',r2_score(y_test,tahminRandomF))
```

| | Games Played | MIN | FGM | FGA | FG% | 3 |
|--------------|--------------|----------|----------|----------|-----------|---------|
| PM \ | | | | | | |
| Games Played | 1.000000 | 0.872959 | 0.739667 | 0.741407 | 0.369492 | 0.4920 |
| 36 | | | | | | |
| MIN | 0.872959 | 1.000000 | 0.924123 | 0.926747 | 0.304910 | 0.6017 |
| 25 | | | | | | |
| FGM | 0.739667 | 0.924123 | 1.000000 | 0.986789 | 0.310255 | 0.5487 |
| 51 | | | | | | |
| FGA | 0.741407 | 0.926747 | 0.986789 | 1.000000 | 0.231871 | 0.6243 |
| 75 | | | | | | |
| FG% | 0.369492 | 0.304910 | 0.310255 | 0.231871 | 1.000000 | -0.0453 |
| 85 | | | | | | |
| 3PM | 0.492036 | 0.601725 | 0.548751 | 0.624375 | -0.045385 | 1.0000 |
| 00 | | | | | | |
| 3PA | 0.510728 | 0.618675 | 0.560697 | 0.646638 | -0.070072 | 0.9885 |
| 63 | | | | | | |
| 3P% | 0.218361 | 0.278590 | 0.254783 | 0.309729 | -0.097218 | 0.5370 |
| 28 | | | | | | |
| FTM | 0.575342 | 0.775022 | 0.868013 | 0.865389 | 0.235433 | 0.4437 |
| CO | | | | | | |

Part-2: Classification

```
In [10]:
```

```
# -*- coding: utf-8 -*-
Created on Sat May 16 16:42:09 2020
@author: Onur
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
import warnings
warnings.filterwarnings("ignore")
veriler = pd.read_csv('Cryotherapy.csv')
x = veriler.iloc[:,0:6]
y = veriler.iloc[:,6:7]
from sklearn.model_selection import train_test_split
x_train, x_test,y_train,y_test = train_test_split(x,y,test_size=0.33, random_state=0)
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(x_train)
X_test = sc.transform(x_test)
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=5,metric='minkowski')
knn.fit(X_train,y_train)
tahminKNN = knn.predict(X test)
print('KNN confusion_matrix \n : ', confusion_matrix(y_test,tahminKNN))
print('KNN classification_report \n :',classification_report(y_test,tahminKNN))
print('KNN accuracy_score \n :', accuracy_score(y_test,tahminKNN))
print("----- \n")
from sklearn.naive bayes import GaussianNB
gnb = GaussianNB()
gnb.fit(X_train,y_train)
tahminGNB = gnb.predict(X_test)
print('GNB confusion_matrix \n : ', confusion_matrix(y_test,tahminGNB))
print('GNB classification_report \n :',classification_report(y_test,tahminGNB))
print('GNB accuracy_score \n :', accuracy_score(y_test,tahminGNB))
from sklearn.svm import SVC
svc = SVC(kernel='rbf')
svc.fit(X train,y train)
tahminSVC = svc.predict(X test)
print('SVC confusion_matrix \n : ', confusion_matrix(y_test,tahminSVC))
print('SVC classification_report \n :',classification_report(y_test,tahminSVC))
print('SVC accuracy_score \n :', accuracy_score(y_test,tahminSVC))
```

| KNN confusi : [[8 3 [0 19]] KNN classif : | _] ication | _report | recall | f1-score | support | |
|---|-------------------|--------------|--------------|----------------------|----------------|--|
| | 0 1 | 1.00 0.86 | | 0.84 0.93 | 11 19 | |
| accurac macro av weighted av | g | | | 0.90 0.88 0.90 | | |
| KNN accurac : 0.9 | y_score | | | | | |
| GNB confusi : [[9 2 [4 15]] GNB classif : | _] ication | _report | recall | f1-score | support | |
| | | | | 0.75 | | |
| | 1 | | | 0.83 | | |
| accurac macro av weighted av | g | 0.79 0.81 | 0.80 | 0.80 0.79 0.80 | 30 30 30 | |
| GNB accuracy_score : 0.8 | | | | | | |
| SVC confusi : [[11 0 [2 17]] SVC classif |] ication | | recall | f1-score | support | |
| | · | | | | | |
| | 0 1 | 0.85 1.00 | 1.00 0.89 | 0.92 0.94 | 11 19 | |
| accurac macro av weighted av | g | 0.92 0.94 | 0.95 0.93 | 0.93 0.93 0.93 | 30 30 30 | |
| SVC accurac : 0.933333 | | | | | | |

Part-3: Clustering

```
In [11]:
```

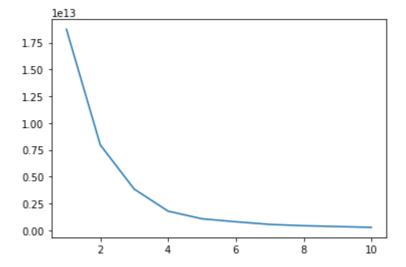
```
# -*- coding: utf-8 -*-
Created on Sat May 16 17:25:22 2020
@author: Onur
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import warnings
warnings.filterwarnings("ignore")
veriler = pd.read_csv('PlayerPlayingPositionData.csv')
last_data = veriler[veriler.columns.difference(['Preferred Positions'])]
last_data.drop('Unnamed: 0', axis=1, inplace=True)
from sklearn.impute import SimpleImputer
numeric = SimpleImputer(missing_values=np.nan, strategy='mean')
numeric = numeric.fit(last_data)
last data = numeric.transform(last data)
from sklearn.cluster import AgglomerativeClustering
ac = AgglomerativeClustering(n_clusters=5,affinity='euclidean',linkage='ward')
tahminAgg = ac.fit_predict(last_data)
print("tahminAgg : ",tahminAgg)
from sklearn.cluster import KMeans
kmeans = KMeans(n_clusters=5,init = 'k-means++')
kmeans.fit(last_data)
tahminKMeans = kmeans.predict(last_data)
print("tahminKMeans : ",tahminKMeans)
print("Kümelerin orta noktas1 :",kmeans.cluster_centers_)
sonuclar = []
for i in range(1,11):
    kmeans = KMeans(n clusters= i,init='k-means++',random state=123)
    kmeans.fit(last_data)
    sonuclar.append(kmeans.inertia )
plt.plot(range(1,11), sonuclar)
tahminAgg : [2 0 0 ... 1 1 1]
tahminKMeans : [3 1 4 ... 2 2 2]
Kümelerin orta noktası : [[6.10970467e+01 5.70405500e+01 5.85828749e+01 6.08
968718e+01
  6.03559727e+01 2.08464326e+05 6.10970467e+01 5.86477355e+01
  5.70405500e+01 6.03559727e+01 5.85828749e+01 6.08968718e+01
  6.19645235e+01 6.00104941e+01 6.12389341e+01 5.94591783e+01
  6.10970467e+01 5.86477355e+01 5.70405500e+01 6.03559727e+01
  5.85828749e+01 6.08968718e+01 6.19645235e+01 6.00104941e+01
  6.12389341e+01 5.94591783e+01 6.00104941e+01]
 [6.19616294e+01 6.03363118e+01 6.14640762e+01 6.15436568e+01
```

6.22928911e+01 1.52246644e+05 6.19616294e+01 6.03527452e+01

```
6.03363118e+01 6.22928911e+01 6.14640762e+01 6.15436568e+01
6.22345414e+01 6.10330432e+01 6.14595147e+01 6.09509362e+01
6.19616294e+01 6.03527452e+01 6.03363118e+01 6.22928911e+01
6.14640762e+01 6.15436568e+01 6.22345414e+01 6.10330432e+01
6.14595147e+01 6.09509362e+01 6.10330432e+01]
[5.60161285e+01 5.18174247e+01 5.29519393e+01 5.58710682e+01
5.47723046e+01 2.32368655e+05 5.60161285e+01 5.36781128e+01
5.18174247e+01 5.47723046e+01 5.29519393e+01 5.58710682e+01
5.69933613e+01 5.49482942e+01 5.63693505e+01 5.43267264e+01
5.60161285e+01 5.36781128e+01 5.18174247e+01 5.47723046e+01
5.29519393e+01 5.58710682e+01 5.69933613e+01 5.49482942e+01
5.63693505e+01 5.43267264e+01 5.49482942e+01]
[6.12226448e+01 5.97575191e+01 6.08471978e+01 6.06322289e+01
 6.18720154e+01 3.50052454e+04 6.12226448e+01 5.83839992e+01
5.97575191e+01 6.18720154e+01 6.08471978e+01 6.06322289e+01
6.05740845e+01 6.04750490e+01 5.99938972e+01 5.89737662e+01
6.12226448e+01 5.83839992e+01 5.97575191e+01 6.18720154e+01
6.08471978e+01 6.06322289e+01 6.05740845e+01 6.04750490e+01
5.99938972e+01 5.89737662e+01 6.04750490e+01]
[6.26944413e+01 5.96192752e+01 6.11110953e+01 6.23714113e+01
6.25409830e+01 1.84749598e+05 6.26944413e+01 6.05854840e+01
5.96192752e+01 6.25409830e+01 6.11110953e+01 6.23714113e+01
6.33409948e+01 6.16804718e+01 6.25343817e+01 6.13938697e+01
6.26944413e+01 6.05854840e+01 5.96192752e+01 6.25409830e+01
6.11110953e+01 6.23714113e+01 6.33409948e+01 6.16804718e+01
6.25343817e+01 6.13938697e+01 6.16804718e+01]]
```

Out[11]:

[<matplotlib.lines.Line2D at 0x2c68017edc8>]



In []: