LAPORAN

TUGAS KECIL 2 IF2121 STRATEGI ALGORITMA

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PROGRAM STUDI TEKNIK INFORMATIKA

SEKOLAH TEKNIK ELEKTRO DAN INFORMATIKA
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BABI

ALGORITMA DIVIDE AND CONQUER

Pada tugas kecil ini, digunakan algoritma *divide and conquer* dengan deskripsi sebagai berikut:

1. Persiapan data

- a. Mulai dengan menomori seluruh titik yang hendak diproses.
- b. Lalu, urutkan seluruh titik yang diproses, menaik berdasarkan sumbu-X. Bila ada nilai pada sumbu-X yang sama, urutkan menaik berdasarkan sumbu-Y.

2. Pemrosesan

- a. Pilih 2 titik, masing-masing p1 dan pn, mewakili dua titik terujung, minimum dan maksimum, berdasarkan sumbu-X.
- b. Jika titik yang diproses kosong, kembalikan p1 dan pn sebagai pembentuk convex hull.
- c. Jika ada titik yang dapat diproses, dilakukan pengelompokan titik-titik ini berdasarkan area yang dipisahkan garis p1pn, kiri dan kanan.
- d. Memanggil fungsi rekursif untuk memproses masing-masing titik yang berada di kiri dan kanan
- e. Mengembalikan kumpulan pasangan titik yang membentuk garis-garis convex hull.

3. Rekursi

- a. Bila titik yang akan diproses sudah habis, mengembalikan p1 dan pn sebagai pembentuk *convex hull*.
- b. Mencari titik terjauh dari garis p1pn, diberi nama pmax. Bila terdapat dua titik yang sama jauhnya, dipilih yang membentuk sudut p1pmaxpn terbesar
- c. Bila pmax ditemukan, bagi area yang dipisahkan garis p1pmax dan pmaxpn menjadi kiri dan kanan masing-masing
- d. Titik yang berada pada area di dalam segitiga p1pmaxpn diabaikan sehingga untuk bagian kiri, hanya diambil kumpulan titik yang berada pada sisi luar, yakni kiri. Sebaliknya, untuk bagian kanan, diambil kumpulan titik yang berada pada sisi kanan.
- e. Ulangi rekursi untuk kumpulan titik yang terpilih untuk masing-masing sisi, hingga kumpulan titik kiri dan kanan habis.
- f. Mengembalikan kumpulan pasangan titik yang membentuk garis-garis convex hull.

4. Keluaran

a. Hasil pemrosesan pustaka berupa 2D numpy array yang merupakan simplices.

BAB II

KODE SUMBER

Kode ditulis dalam Bahasa Python. Berikut merupakan kode sumber yang terdapat di dalam file myConvexHull.py. Kode sumber juga dapat diakses melalui Github https://github.com/rannnayy/stima-convexhull atau Google Drive pengumpulan.

```
import numpy as np
# function to compute determinant
def determinant(p1, p2, p3):
return p1[1]*p2[2] + p3[1]*p1[2] + p2[1]*p3[2] - p3[1]*p2[2] - p2[1]*p1[2] - p1[1]*p3[2]
# function to determine position of a point p3 towards line p1p2
def leftOrRight(p1, p2, p3):
    # p3 is on left side of line p1p2 if determinant is positive
     det = determinant(p1, p2, p3)
     if (det > 0):
     elif (det < 0):
def divide(points, p1, pn):
    return left, right
    # classifying each points to three categories through leftOrRight function, # namely left, right, and inline
         if (not (point[0] == p1[0] or point[0] == pn[0])):
    loc = leftOrRight(p1, pn, point)
              if (loc == "left"):
                  left = np.append(left, np.array([point]), axis=0)
             elif (loc == "right"):
         right = np.append(right, np.array([point]), axis=0)
# points where loc == "inline", p1, and pn is ignored since they can't form hull
    return left, right
def distance(p1, p2, px):
   A = p1[2]-p2[2]

B = p2[1]-p1[1]

C = p1[1]*p2[2]-p2[1]*p1[2]

return abs(A*px[1] + B*px[2] + C)/((A*A + B*B)**(1/2))
def angle(p1, pmax, pn):
    pA = np.array(p1)
    pB = np.array(pmax)
    pC = np.array(pn)
     vectBA = pA - pB
     vectBC = pC - pB
    return (np.degrees(np.arccos((vectBA @ vectBC))(np.linalg.norm(vectBA) * np.linalg.norm(vectBC)))))
def myConvexHull2(p1, pn, part, leftRightPos):
    # make an empty array to store hull simplices
cvHull = np.empty((0, 2))
     # if array of points is already empty, means there aren't any points other than p1 and pn
# p1 and pn is one of the hull's simplex
     if (not(np.size(part))):
         return [[p1[0], pn[0]]]
         dist_pmax = -1
         pmax = None
```

```
idx_pmax = 0
           ctr = 0
           for point in part:
                temp_dist = distance(p1, pn, point)
                 if (temp_dist > dist_pmax):
                     dist_pmax = temp_dist
                      pmax = point
                      idx pmax = ctr
                # if there are two/more points with same distance, choose by maximum angle gotten
                elif (temp_dist == dist_pmax and not(pmax is None)):
   if (angle(p1, point, pn) > angle(p1, pmax, pn)):
                           dist_pmax = temp_dist
                           pmax = point
                            idx_pmax = ctr
                ctr += 1
           if (not(pmax is None)):
                part = np.delete(part, idx_pmax, axis=0)
                # divide to two parts, only take the outer points
p1pmaxleft, p1pmaxright = divide(part, p1, pmax)
pmaxpnleft, pmaxpnright = divide(part, pmax, pn)
                # for points on left side of p1pn, take only left parts
if (leftRightPos == "left"):
                cvHull = np.append(cvHull, np.array(myConvexHull2(p1, pmax, p1pmaxleft, "left")), axis=0)
cvHull = np.append(cvHull, np.array(myConvexHull2(pmax, pn, pmaxpnleft, "left")), axis=0)
# for points on right side of p1pn, take only right parts
elif (leftRightPos == "right"):
                      cvHull = np.append(cvHull, np.array(myConvexHull2(p1, pmax, p1pmaxright, "right")), axis=0) cvHull = np.append(cvHull, np.array(myConvexHull2(pmax, pn, pmaxpnright, "right")), axis=0)
           return cvHull
def numTitik(points):
     # create new array to store numbered points
     tempPoints = np.empty((0, 3))
     # iterate for each point, add an identifier number
for i in range(len(points)):
          tempPoints = np.append(tempPoints, np.array([[i, points[i][0], points[i][1]]]), axis=0)
     return tempPoints
def myConvexHull(points):
     points = numTitik(points)
     cvHull = np.empty((0, 2))
     points = points[np.lexsort((points[:,2], points[:,1]))]
     # p1 and pn, leftmost and rightmost points respectively
p1 = points[0]
     pn = points[-1]
     if (not(np.size(points))):
           return [[p1[0], pn[0]]]
           left, right = divide(points, p1, pn)
           # call recursive func
          cvHull = np.append(cvHull, np.array(myConvexHull2(p1, pn, left, "left")), axis=0)
cvHull = np.append(cvHull, np.array(myConvexHull2(p1, pn, right, "right")), axis=0)
```

Kode berikut merupakan kode untuk memvisualisasikan hasil penggunakan Pustaka yang telah dibuat. Kode berikut terdapat di dalam file myConvexHull.ipynb.

```
import pandas as pd
from sklearn import datasets
import matplotlib.pyplot as plt
from myConvexHull import myConvexHull
```

```
data = datasets.load_iris()

#create a DataFrame
df1 = pd.DataFrame(data.data, columns=data.feature_names)
df1['Target'] = pd.DataFrame(data.target)
print(df1.shape)
df1.head()
```

```
plt.figure(figsize = (10, 6))
colors = ['b','r','g']
plt.title('Petal Width vs Petal Length')
plt.xlabel(data.feature_names[2])
plt.ylabel(data.feature_names[3])
for i in range(len(data.target_names)):
    bucket = df1[df1['Target'] == i]
    bucket = bucket.iloc[:,[2,3]].values
    hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
hull = hull.astype(int)
    plt.scatter(bucket[:, 0], bucket[:, 1], label=data.target_names[i])
    for points in hull:
        plt.plot(bucket[points, 0], bucket[points, 1], colors[i])
plt.legend()
```

```
#create a DataFrame
df2 = pd.DataFrame(data.data, columns=data.feature_names)
df2['Tanget'] = pd.DataFrame(data.target)
print(df2.shape)
df2.head()
plt.figure(figsize = (10, 6))
colors = ['b','r','g']
plt.title('Sepal Width vs Sepal Length')
plt.xlabel(data.feature_names[0])
plt.ylabel(data.feature_names[1])
for i in range(len(data.target_names)):
    bucket = df2[df2['Tanget'] == i]
    bucket = bucket.iloc[:,[0,1]].values
    hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
    hull = hull.astype(int)
    plt.scatter(bucket[:, 0], bucket[:, 1], label=data.target_names[i])
    for points in hull:
        plt.plot(bucket[points, 0], bucket[points, 1], colors[i])
plt.legend()
```

```
rom sklearn.datasets import fetch_california_housing
data2 = fetch_california_housing()
#create a DataFrame
df3 = pd.DataFrame(data2.data, columns = data2.feature_names)
df3['Target'] = pd.DataFrame(data2.target)
print(df3.shape)
df3.head()
cols = [[0,8], [1,8], [2,8], [3,8], [4,8], [5,8], [6,8], [7,8]] for col in cols:
    plt.figure(figsize = (5, 3))
    plt.title(data2.feature_names[col[0]] + " vs Target")
    plt.xlabel(data2.feature_names[col[0]])
    plt.ylabel("Target")
bucket = df3[0:1000]
    bucket = bucket.iloc[:,[col[0],col[1]]].values
hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
    hull = hull.astype(int)
    plt.scatter(bucket[:, 0], bucket[:, 1], label=data2.feature_names[col[0]])
     for points in hull:
         plt.plot(bucket[points, 0], bucket[points, 1], 'b')
    plt.legend()
```

```
data3 = datasets.load_diabetes()

#create a DataFrame
df4 = pd.DataFrame(data3.data, columns = data3.feature_names)
df4['Target'] = pd.DataFrame(data3.target)
print(df4.shape)
df4.head()
cols = [[0,10], [1,10], [2,10], [3,10], [4,10], [5,10], [6,10], [7,10], [8,10], [9,10]]
for col in cols:
    plt.figure(figsize = (5, 3))
    plt.title(data3.feature_names[col[0]] + " vs Target")
    plt.ylabel(data3.feature_names[col[0]])
    plt.ylabel("Target")
    bucket = df4[0:1000]
    bucket = bucket.iloc[:,[col[0],col[1]]].values
    hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
    hull = hull.astype(int)
    plt.scatter(bucket[:, 0], bucket[:, 1], label=data3.feature_names[col[0]])
    for points in hull:
        plt.plot(bucket[points, 0], bucket[points, 1], 'b')
    plt.legend()
```

```
plt.figure(figsize = (10, 6))
colors = ['b','r','g', 'c', 'm', 'y', 'k', 'b', 'r', 'g']
plt.title("All vs Target")
plt.xlabel("All")
plt.ylabel("Target")
for i in range(len(cols)):
    bucket = df4[0:1000]
    bucket = bucket.iloc[:,[cols[i][0],cols[1][1]]].values
    hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
    hull = hull.astype(int)
    plt.scatter(bucket[:, 0], bucket[:, 1], label=data3.feature_names[cols[i][0]])
    for points in hull:
        plt.plot(bucket[points, 0], bucket[points, 1], colors[i])
plt.legend()
```

```
data4X, data4Y = datasets.load_linnerud(return_X_y=True)
df5X = pd.DataFrame(data4Y, columns = ["chins", "sit_ups", "jumps"])
df5Y = pd.DataFrame(data4X, columns = ["weight", "waist", "pulse"])
df5 = pd.merge(df5X, df5Y, left_index=True, right_index=True)
print(df5.shape)
df5.head()
plt.figure(figsize = (10, 6))
plt.title('Chins vs Sit Ups, Chins vs Jumps, Sit Ups vs Jumps')
plt.xlabel("Chins, Chins, Sit Ups")
plt.ylabel("Sit Ups, Jumps, Jumps")
# 1 : Chins vs Sit Ups
bucket = df5.iloc[:,[0,1]].values
hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
hull = hull.astype(int)
plt.scatter(bucket[:, 0], bucket[:, 1], label="Chins vs Sit Ups")
for points in hull:
    plt.plot(bucket[points, 0], bucket[points, 1], "b")
bucket = df5.iloc[:,[0,2]].values
hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
hull = hull.astype(int)
plt.scatter(bucket[:, 0], bucket[:, 1], label="Chins vs Jumps")
for points in hull:
    plt.plot(bucket[points, 0], bucket[points, 1], "r")
bucket = df5.iloc[:,[1,2]].values
hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
hull = hull.astype(int)
plt.scatter(bucket[:, 0], bucket[:, 1], label="Sit Ups vs Jumps")
for points in hull:
     plt.plot(bucket[points, 0], bucket[points, 1], "g")
plt.legend()
```

```
data6 = datasets.load_breast_cancer()
#create a DataFrame
df7 = pd.DataFrame(data6.data, columns=data6.feature_names)
```

BAB III EKSPERIMEN

TUGAS KECIL 2 IF2211 STRATEGI ALGORITMA

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STRUCTURE

- Title and Identity
- Structure
- Library ImportsDatasets Available
- Iris Plants Dataset

 1. Petal Length Petal Width

 2. Sepal Length Sepal Width

 3. Boston House Prices Dataset

 4. Diabetes Dataset

- 4. Diabetes Dataset
 5. Linerrud Dataset
 6. Wine Recognition Dataset
 7. Breast Cancer Wisconsin (Diagnosis) Dataset
 Acknowledgements

Iris Plant Dataset

1. Petal Length - Petal Width

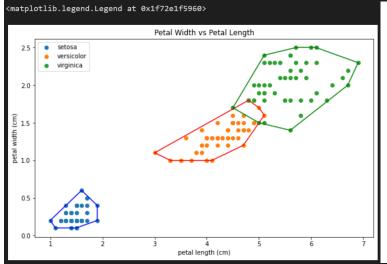
```
data = datasets.load_iris()
  #create a DataFrame
  df1 = pd.DataFrame(data.data, columns=data.feature_names)
  df1['Target'] = pd.DataFrame(data.target)
  print(df1.shape)
  df1.head()
✓ 0.4s
```

(150, 5)

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	Target
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

Library Imports

import matplotlib.pyplot as plt from myConvexHull import myConvexHull



2. Sepal Length - Sepal Width

```
data = datasets.load_iris()

#create a DataFrame

df2 = pd.DataFrame(data.data, columns=data.feature_names)

df2['Target'] = pd.DataFrame(data.target)

print(df2.shape)

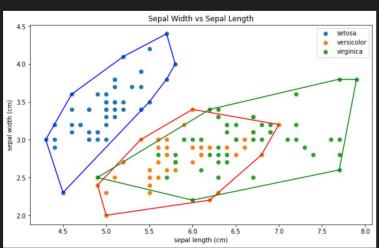
df2.head()

0.4s
```

(150, 5)

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)	Target
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

<matplotlib.legend.Legend at 0x1f72e3fafe0>



3. Boston House Dataset

```
from sklearn.datasets import fetch_california_housing
data2 = fetch_california_housing()

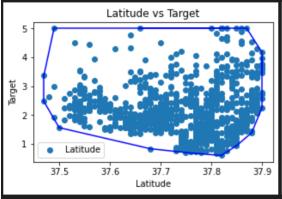
#create a DataFrame
df3 = pd.DataFrame(data2.data, columns = data2.feature_names)
df3['Target'] = pd.DataFrame(data2.target)
print(df3.shape)
df3.head()

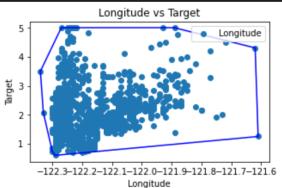
0.6s
```

(20640, 9)

	MedInc	HouseAge	AveRooms	AveBedrms	Population	AveOccup	Latitude	Longitude	Target
0	8.3252	41.0	6.984127	1.023810	322.0	2.555556	37.88	-122.23	4.526
1	8.3014	21.0	6.238137	0.971880	2401.0	2.109842	37.86	-122.22	3.585
2	7.2574	52.0	8.288136	1.073446	496.0	2.802260	37.85	-122.24	3.521
3	5.6431	52.0	5.817352	1.073059	558.0	2.547945	37.85	-122.25	3.413
4	3.8462	52.0	6.281853	1.081081	565.0	2.181467	37.85	-122.25	3.422

```
cols = [[0,8], [1,8], [2,8], [3,8], [4,8], [5,8], [6,8], [7,8]]
                                plt.figure(figsize = (5, 3))
                                plt.title(data2.feature_names[col[0]] + " vs Target")
                                plt.xlabel(data2.feature_names[col[0]])
                                plt.ylabel("Target")
                                bucket = df3[0:1000]
                                bucket = bucket.iloc[:,[col[0],col[1]]].values
                                \textbf{hull} = \textbf{-myConvexHull(bucket)} + \textbf{\#bagian} \cdot \textbf{ini} \cdot \textbf{diganti} \cdot \textbf{dengan} \cdot \textbf{hasil} \cdot \textbf{implementasi} \cdot \textbf{ConvexHull} \cdot \textbf{Divide} \cdot \& \cdot \textbf{Conquermontal} \cdot \textbf{ConvexHull} \cdot \textbf{Divide} \cdot \textbf{Conquermontal} \cdot \textbf{ConvexHull} \cdot 
                                hull = hull.astype(int)
                                plt.scatter(bucket[:, 0], bucket[:, 1], label=data2.feature_names[col[0]])
                                 for points in hull:
                                      plt.plot(bucket[points, 0], bucket[points, 1], 'b')
                                 plt.legend()
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Target
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                                                                                                                                                                                                                                                                                                                                                                                                                                                                           2.5
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                                                                                                             Population
                                                                                                                                                                                                                                                                                                                                                                         AveOccup
```





4. Diabetes Dataset

```
data3 = datasets.load_diabetes()

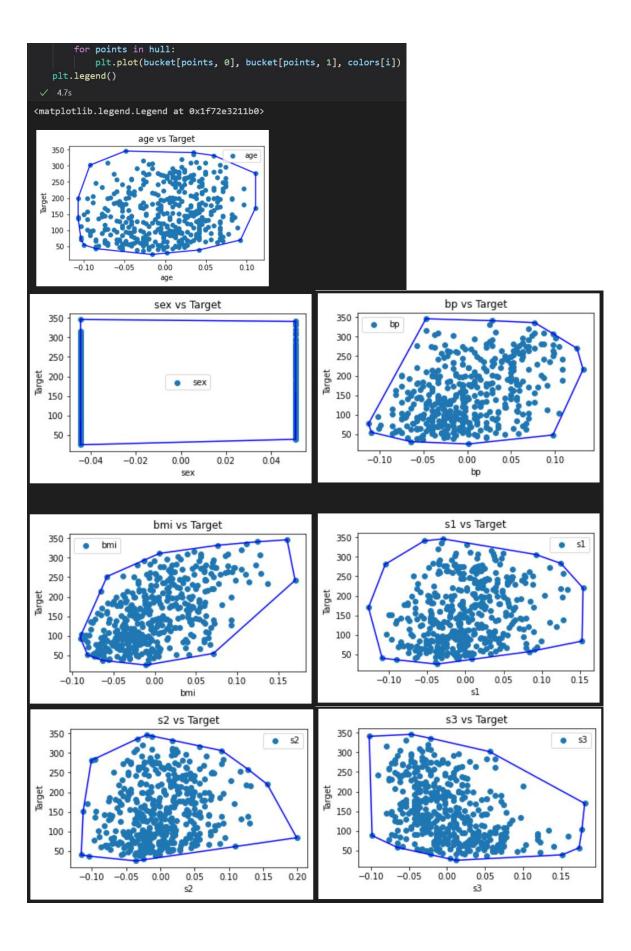
#create a DataFrame
df4 = pd.DataFrame(data3.data, columns = data3.feature_names)
df4['Target'] = pd.DataFrame(data3.target)
print(df4.shape)
df4.head()

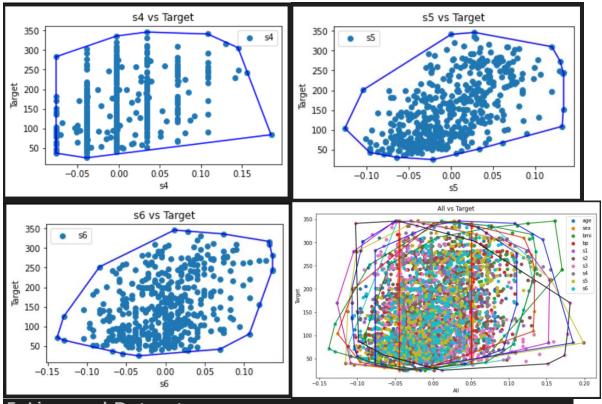
    0.4s
```

(442, 11)

```
bmi
                                    bр
                                                         s2
                                                                    s3
                                                                              s4
                                                                                        s5
                                                                                                   s6 Target
               sex
    age
0.038076
          0.050680
                     0.061696
                               0.021872 -0.044223 -0.034821 -0.043401 -0.002592
                                                                                   0.019908 -0.017646
-0.001882 -0.044642
                    -0.051474 -0.026328
                                         -0.008449 -0.019163
                                                              0.074412 -0.039493
                                                                                  -0.068330 -0.092204
                                                                                                         75.0
0.085299
                                                                                                        141.0
                     0.044451 -0.005671
                                        -0.045599 -0.034194
                                                             -0.032356 -0.002592
                                                                                   0.002864 -0.025930
          0.050680
-0.089063 -0.044642 -0.011595 -0.036656
                                         0.012191
                                                    0.024991 -0.036038 0.034309
                                                                                   0.022692 -0.009362
                                                                                                        206.0
0.005383 -0.044642 -0.036385
                               0.021872
                                         0.003935 0.015596
                                                              0.008142 -0.002592 -0.031991 -0.046641
                                                                                                        135.0
```

```
cols = [[0,10], [1,10], [2,10], [3,10], [4,10], [5,10], [6,10], [7,10], [8,10], [9,10]]
for col in cols:
   plt.figure(figsize = (5, 3))
   plt.title(data3.feature_names[col[0]] + " vs Target")
   plt.xlabel(data3.feature_names[col[0]])
   plt.ylabel("Target")
   bucket = df4[0:1000]
   bucket = bucket.iloc[:,[col[0],col[1]]].values
   hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
   hull = hull.astype(int)
   plt.scatter(bucket[:, 0], bucket[:, 1], label=data3.feature_names[col[0]])
    for points in hull:
       plt.plot(bucket[points, 0], bucket[points, 1], 'b')
   plt.legend()
plt.figure(figsize = (10, 6))
colors = ['b','r','g', 'c', 'm', 'y', 'k', 'b', 'r', 'g']
plt.title("All vs Target")
plt.xlabel("All")
plt.ylabel("Target")
for i in range(len(cols)):
   bucket = df4[0:1000]
   bucket = bucket.iloc[:,[cols[i][0],cols[1][1]]].values
    hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
   hull = hull.astype(int)
   plt.scatter(bucket[:, 0], bucket[:, 1], label=data3.feature_names[cols[i][0]])
```



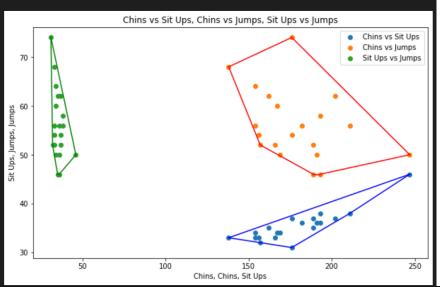


5. Linnerud Dataset

```
data4X, data4Y = datasets.load_linnerud(return_X_y=True)
   #create a DataFrame
   # dataset consists of 3 exercises (data)
   # and 3 physiological (target)
   df5X = pd.DataFrame(data4Y, columns = ["chins", "sit_ups", "jumps"])
   df5Y = pd.DataFrame(data4X, columns = ["weight", "waist", "pulse"])
   df5 = pd.merge(df5X, df5Y, left_index=True, right_index=True)
   print(df5.shape)
   df5.head()
✓ 0.4s
(20, 6)
           sit_ups jumps weight waist
    chins
                                          pulse
 0 191.0
              36.0
                     50.0
                               5.0
                                   162.0
                                           60.0
    189.0
             37.0
                     52.0
                              2.0
                                  110.0
                                           60.0
 2 193.0
              38.0
                     58.0
                             12.0
                                   101.0
                                          101.0
 3
    162.0
             35.0
                     62.0
                             12.0
                                   105.0
                                           37.0
    189.0
              35.0
                     46.0
                                  155.0
                                           58.0
                              13.0
```

```
plt.figure(figsize = (10, 6))
plt.title('Chins vs Sit Ups, Chins vs Jumps, Sit Ups vs Jumps')
plt.xlabel("Chins, Chins, Sit Ups")
plt.ylabel("Sit Ups, Jumps, Jumps")
# 1 : Chins vs Sit Ups
bucket = df5.iloc[:,[0,1]].values
hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
hull = hull.astype(int)
plt.scatter(bucket[:, 0], bucket[:, 1], label="Chins vs Sit Ups")
for points in hull:
   plt.plot(bucket[points, 0], bucket[points, 1], "b")
bucket = df5.iloc[:,[0,2]].values
hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
hull = hull.astype(int)
plt.scatter(bucket[:, 0], bucket[:, 1], label="Chins vs Jumps")
for points in hull:
    plt.plot(bucket[points, 0], bucket[points, 1], "r")
bucket = df5.iloc[:,[1,2]].values
hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
hull = hull.astype(int)
plt.scatter(bucket[:, 0], bucket[:, 1], label="Sit Ups vs Jumps")
for points in hull:
    plt.plot(bucket[points, 0], bucket[points, 1], "g")
plt.legend()
```

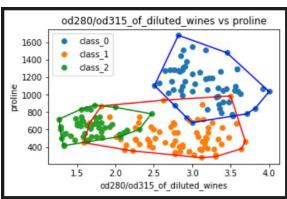
<matplotlib.legend.Legend at 0x1f72e955e70>



6. Wine Recognition Dataset

```
df6 = pd.DataFrame(data5.data, columns=data5.feature_names)
   df6['Target'] = pd.DataFrame(data5.target)
print(df6.shape)
   df6.head()
(178, 14)
                                                                                                              proanthocyanins color_intensity hue od280/od315
    alcohol malic_acid ash alcalinity_of_ash magnesium total_phenols flavanoids nonflavanoid_phenols
                   1.71 2.43
                                                                                                                          2.29
                                                                                                                                               1.04
      13.20
                    1.78 2.14
                                                       100.0
                                                                                                                           1.28
                   2.36 2.67
                                           18.6
                                                       101.0
                                                                      2.80
                                                                                  3.24
                                                                                                         0.30
      13.24
                   2.59 2.87
                                                       118.0
                                                                                                                                          4.32 1.04
```

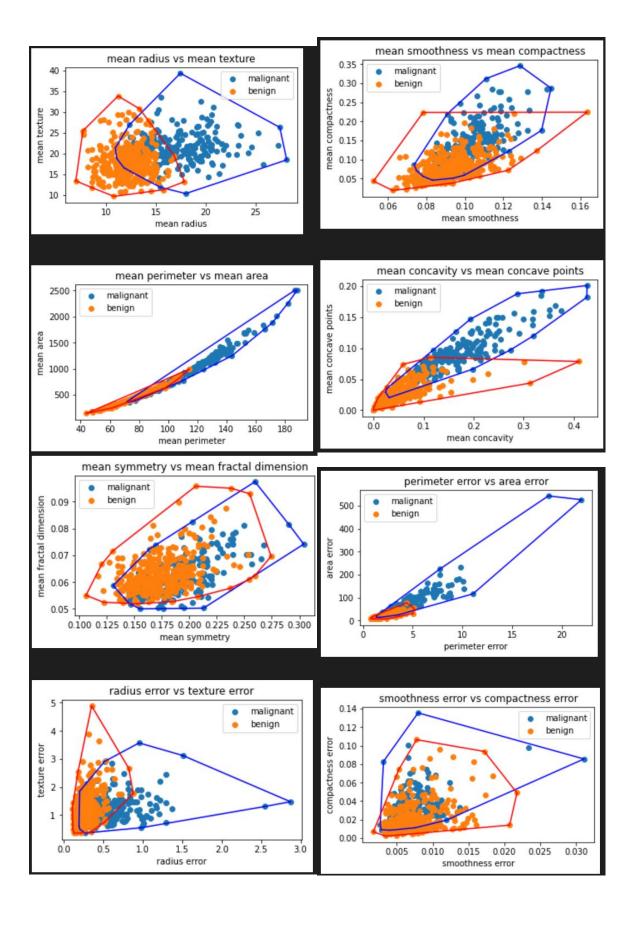
```
colors = ['b','r','g']
   cols = [[0,1], [2,3], [4,5], [6,7], [8,9], [10,11], [11,12]]
       plt.figure(figsize = (5, 3))
       plt.title(data5.feature_names[col[0]] + " vs " + data5.feature_names[col[1]])
       plt.xlabel(data5.feature_names[col[0]])
       plt.ylabel(data5.feature_names[col[1]])
       for i in range(len(data5.target_names)):
           bucket = df6[df6['Target'] == i]
           bucket = bucket.iloc[:,[col[0], col[1]]].values
           hull = hull.astype(int)
           plt.scatter(bucket[:, 0], bucket[:, 1], label=data5.target_names[i])
           for points in hull:
               plt.plot(bucket[points, 0], bucket[points, 1], colors[i])
       plt.legend()
✓ 1.9s
                                                                      magnesium vs total phenols
                 alcohol vs malic_acid
                                                         4.0
                                                                                                   dass_0
                                          dass_0
                                                         3.5
                                                                                                   dass 1
                                          dass_1
   5
                                                                                                   dass 2
                                          dass_2
                                                         3.0
                                                      total phenols
   4
malic acid
                                                         2.5
  3
                                                         2.0
   2
                                                         1.5
                                                         1.0
                                                                                    120
                                                                                                      160
                                                                           100
                                                                                             140
         11.5
               12.0 12.5
                          13.0
                               13.5 14.0 14.5 15.0
     11.0
                                                                               magnesium
                        alcohol
                                                                  flavanoids vs nonflavanoid_phenols
                ash vs alcalinity_of_ash
   30
                                                                                                  dass_0
            class 0
                                                         0.6
                                                                                                   dass_1
            dass_1
                                                      nonflavanoid phenols
alcalinity of ash
                                                                                                   dass 2
            dass 2
                                                         0.5
                                                         0.4
                                                         0.3
                                                         0.2
   10
                                                                                                       5
                                                                                               4
                         2.25
                               2.50 2.75 3.00 3.25
                           ash
           proanthocyanins vs color_intensity
                                                                hue vs od280/od315 of diluted wines
                                                         4.0
                                                      dass 0
                                                                                                  dass 0
  12
                                           dass_1
                                                                                                  dass_1
  10
                                           dass_2
                                                                                                  dass_2
color intensity
   8
   6
   4
   2
       0.5
                                        3.0
              1.0
                           2.0
                                 2.5
                                              3.5
                                                                        0.8
                                                                               1.0
                                                                                     12
                                                                                            1.4
                                                                                                   1.6
                     proanthocyanins
```

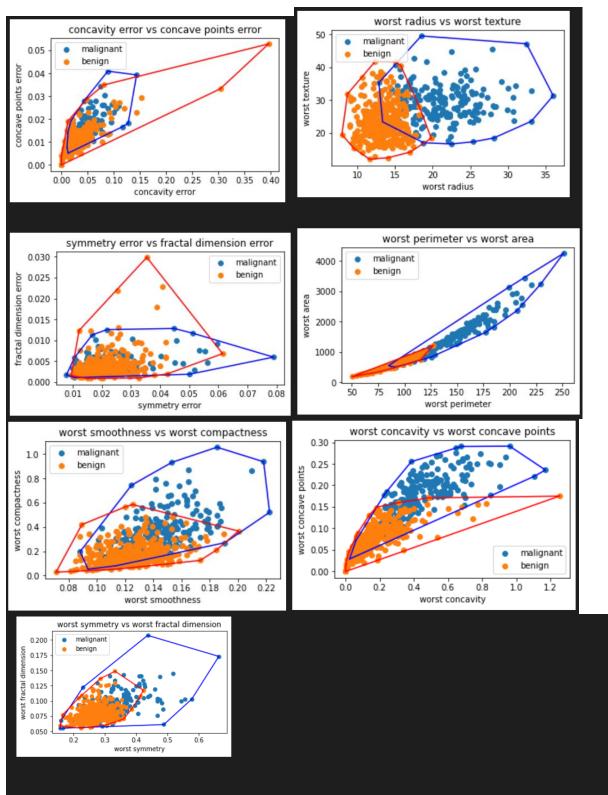


✓ 2.6s

```
7. Breast cancer wisconsin (diagnostic) Dataset
    data6 = datasets.load_breast_cancer()
    df7 = pd.DataFrame(data6.data, columns=data6.feature_names)
df7['Target'] = pd.DataFrame(data6.target)
print(df7.shape)
                                                                                                     mean
fractal
                                                                                                                  worst
                                                                             concave
     radius
            texture perimeter
                                 area smoothness compactness concavity
                                                                                      symmetry
                                                                                                                texture perimeter
                                                                                                                                     area smoothness com
                                                                              points
      17.99
               10.38
                         122.80 1001.0
                                            0.11840
                                                         0.27760
                                                                     0.3001
                                                                             0.14710
                                                                                         0.2419
                                                                                                                  17.33
                                                                                                                            184.60 2019.0
                                                                                                                                                0.1622
                                                                                                                                                 0.1238
                         130.00 1203.0
                                            0.10960
                                                         0.15990
                                                                                         0.2069
                                                                                                    0.05999
                                                                                                                            152.50 1709.0
                                                                                                                                                 0.1444
                                            0.14250
                                                                                                                                                 0.2098
               20.38
                                 386.1
                                                         0.28390
                                                                                                    0.09744
                                                                                                                  26.50
                                                                                                                             98.87
                                                                                                    0.05883
    colors = ['b','r','g']

cols = [[0,1], [2,3], [4,5], [6,7], [8,9], [10,11], [12,13], [14,15], [16,17], [18,19], [20,21], [22,23], [24,25], [26,27], [28,29]]
     for col in cols:
         plt.figure(figsize = (5, 3))
         plt.title(data6.feature_names[col[0]] + " vs " + data6.feature_names[col[1]])
         plt.xlabel(data6.feature_names[col[0]])
         plt.ylabel(data6.feature_names[col[1]])
         for i in range(len(data6.target_names)):
             bucket = df7[df7['Target'] == i]
bucket = bucket.iloc[:,[col[0], col[1]]].values
              hull = myConvexHull(bucket) #bagian ini diganti dengan hasil implementasi ConvexHull Divide & Conquer
              hull = hull.astype(int)
              plt.scatter(bucket[:, 0], bucket[:, 1], label=data6.target_names[i])
                 plt.plot(bucket[points, 0], bucket[points, 1], colors[i])
         plt.legend()
```





Acknowledgements

References:

- Munir, R., Maulidevi, N. U. 2022. Algoritma Divide and Conquer (Bagian 4). Bahan Kuliah IF2211 Strategi Algoritma.
 Anany V. Levitin. 2002. Introduction to the Design and Analysis of Algorithms. Addison-Wesley Longman Publishing Co., Inc., USA.
 Scikit-Learn Toy Datasets Webpage. Accessed from: https://scikit-learn.org/stable/datasets/toy_dataset.html#linnerrud-dataset on

BAB IV

CHECKLIST

Poin	Ya	Tidak		
1. Pustaka <i>myConvexHull</i> berhasil dibuat dan tidak ada	/			
kesalahan	V			
2. Convex hull yang dihasilkan sudah benar	√			
3. Pustaka <i>myConvexHull</i> dapat digunakan untuk menampilkan	✓			
convex hull setiap label dengan warna yang berbeda.				
4. Bonus : program dapat menerima input dan menuliskan	/			
output untuk dataset lainnya.	v			