"KONSEP & APLIKASI DATA MAINING"



Oleh:

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PROGRAM STUDI S1- TEKNOLOGI INFORMASI PROGRAM STUDI S1- SISTEM INFORMASI

KEMENTRIAN RISET DAN TEKNOLOGI PENDIDIKAN TINGGI SEKOLAH TINGGI MANAJEMEN INFORMATIKA KOMPUTER PRADNYA PARAMITA MALANG 2020

```
In [1]: import numpy as np
          import manapy as np
import pandas as pd
import matplotlib.pyplot as plt
  In [7]: df=pd.read_csv('D:/agnes/dataset_soalnol.csv',delimiter=';')
  In [8]: df.head()
  Out[8]:
              Age Income Student Credit_rating Class (buy_computer)
           0 <=30 High No Fair
          1 <=30 High
                                  Excellent
                                                      No
                           No
          2 31..40 High No
                                  Fair
                                                      Yes
           3 > 40 Medium No
                                 Fair
                                                      Yes
                                 Fair
           4 > 40 Low Yes
                                                      Yes
  In [9]: df.shape
  Out[9]: (51, 5)
 In [10]: #student
df['Student'].value_counts()
Out[10]: Yes 27
                 24
          Name: Student, dtype: int64
In [11]: PYes = 27/51
          PNo = 24/51
In [12]: print(PYes)
          0.5294117647058824
In [13]: print(PNo)
          0.47058823529411764
In [14]: #income with student
          pd.crosstab(df['Income'], df['Student'])
Out[14]:
           Student No Yes
           Income
           High 9 2
             Low 1 20
           Medium 14 5
```

```
In [15]: PHighNo = 9/24
         PLowNo = 1/24
         PMediumNo= 14/24
         PHighYes = 2/27
         PLowYes = 20/27
         PMediumYes = 5/27
         PHigh = 11/51
         PLow = 21/51
         PMedium = 19/51
         print(PHighNo)
         0.375
In [16]: print(PLowNo)
         0.04166666666666664
In [17]: print(PMediumNo)
         0.58333333333333334
In [18]: print(PHighYes)
         0.07407407407407407
In [19]: print(PLowYes)
         0.7407407407407407
In [20]: print(PMediumYes)
         0.18518518518518517
In [21]: print(PHigh)
         0.21568627450980393
In [22]: print(PLow)
         0.4117647058823529
In [23]: print(PMedium)
         0.37254901960784315
In [24]: #credit rating with student
         pd.crosstab(df['Credit_rating'], df['Student'])
```

```
Out[24]:
             Student No Yes
         Credit_rating
            Excellent 8 12
               Fair 16 15
In [25]: PExcellentNo = 8/24
         PFairNo = 16/24
         PExcellentYes = 12/27
         PFairYes = 15/27
         PExcellent = 20/51
         PFair = 31/51
         print(PExcellentNo)
         0.3333333333333333
In [26]: print(PFairNo)
         0.66666666666666
In [27]: print(PExcellentYes)
         0.444444444444444
In [28]: print(PFairYes)
          0.55555555555556
In [29]: print(PExcellent)
          0.39215686274509803
In [30]: print(PFair)
          0.6078431372549019
In [31]: #income with class(buy computer)
          pd.crosstab(df['Income'], df['Class (buy_computer)'])
Out[31]:
           Class (buy_computer) No Yes
                       Income
                                    5
                         High
                                   10
                         Low
                              11
                       Medium
                                   14
```

```
In [32]: PHighNo = 6/22
         PLowNo = 11/22
         PMediumNo= 5/22
         PHighYes = 5/29
         PLowYes = 10/29
         PMediumYes = 24/29
         PHigh = 11/51
         PLow = 21/51
         PMedium = 19/51
         print(PHighNo)
         0.2727272727272727
In [33]: print(PLowNo)
         0.5
In [34]: print(PMediumNo)
         0.22727272727272727
In [35]: print(PHighYes)
         0.1724137931034483
```

```
In [36]: print(PLowYes)
         0.3448275862068966
In [37]: print(PMediumYes)
         0.8275862068965517
In [38]: #credit rating with class(buy_computer)
         pd.crosstab(df['Credit_rating'], df['Class (buy_computer)'])
Out[38]:
          Class (buy_computer) No Yes
                Credit_rating
                   Excellent 8 12
                       Fair 14 17
In [39]: PExcellentNo = 8/22
         PFairNo = 14/22
         PExcellentYes = 12/29
         PFairYes = 17/29
         PExcellent = 20/51
         PFair = 31/51
            0.36363636363636365
  In [40]: print(PFairNo)
            0.6363636363636364
  In [41]: print(PExcellentYes)
            0.41379310344827586
  In [42]: print(PFairYes)
            0.5862068965517241
```

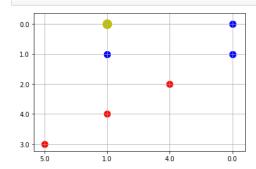
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                                                                                              2004
     In [1]: import pandas as pd
                  import numpy as np
                  import matplotlib.pyplot as plt
                  %matplotlib inline
      In [2]: pd.__version__
     Out[2]: '1.0.1'
     In [3]: df = pd.read excel('D:/agnes/dataset soalno2.xls')
     In [4]: df
     Out[4]:
                       Category Weather Holiday Game
                                                                    Qty
                   0
                            NaN
                                          V-1
                                                    V-2
                                                            V-3
                                                                   NaN
                    1
                                Α
                                           5
                                                              0 250.0
                               В
                                                              1 200.0
                    3
                               С
                                                      1
                                                              0 75.0
                                           1
                    4
                               D
                                           4
                                                              1 400.0
                    5
                               Ε
                                           4
                                                      0
                                                              0 150.0
                                           2
                                                              0 50.0
                                        1 400.0
                   Е
                           4
                                  0
                                        0 150.0
                                 0
                                       0 50.0
In [38]: #a.Apabila Cuaca buruk dengan nilai = 1, Weekday, dan Game = 0, maka berapa roti yang harus dibuat?
# misalkan hari misterius = H-M (Weekday)
data = np.array([[5.,3.,'Weather V-1'],[1.,4.,'Weather V-1'],[4.,2.,'Weather V-1'],[1.,1.,'Holiday V-2'],[1.,1.,'Holiday V-2']
query = [1.,0.,'Weekday H-M']
         4
In [39]: df = pd.DataFrame(data)
    df.columns = ['x','y','Qty']
Out[391:
                            Qty
          0 5.0 3.0 Weather V-1
           1 1.0 4.0 Weather V-1
           2 4.0 2.0 Weather V-1
           3 1.0 1.0 Holiday V-2
          4 1.0 1.0 Holiday V-2
           5 0.0 0.0 Holiday V-2
          6 0.0 1.0 Game V-3
```

```
5 0.0 0.0 Holiday V-2
6 0.0 1.0 Game V-3
7 0.0 1.0 Game V-3
8 0.0 0.0 Game V-3
9 1.0 0.0 Weekday H-M
```

```
[41]: for i in range(10):
    if(df.iloc[i]['Qty'] == 'Weather V-1'):
        plt.scatter(df.iloc[i]['x'], df.iloc[i]['y'], s=100, c='r')
    elif(df.iloc[i]['Qty'] == 'Weekday H-M'):
        plt.scatter(df.iloc[i]['x'], df.iloc[i]['y'], s=200, c='y')
    else:
        plt.scatter(df.iloc[i]['x'], df.iloc[i]['y'], s=100, c='b')

plt.grid()
plt.show()
```



```
In [42]: import math
    dis = []
    for i in range(10):
        dis.append(math.sqrt((float(df.iloc[i]['x']) - query[1]) **2 + (float(df.iloc[i]['y']) - query[0]) **2))

In [43]: df['dis'] = dis
    df
```

- - - -

```
In [43]: df['dis'] = dis
          df
Out[43]:
                                       dis
                              Qty
               x y
                       Weather V-1 5.385165
           0 5.0 3.0
           1 1.0 4.0
                       Weather V-1 3.162278
           2 4.0 2.0
                       Weather V-1 4.123106
           3 1.0 1.0
                       Holiday V-2 1.000000
           4 1.0 1.0
                       Holiday V-2 1.000000
           5 0.0 0.0
                       Holiday V-2 1.000000
           6 0.0 1.0
                       Game V-3 0.000000
           7 0.0 1.0
                       Game V-3 0.000000
           8 0.0 0.0
                       Game V-3 1.000000
           9 1.0 0.0 Weekday H-M 1.414214
In [44]: df.sort_values('dis')
In [43]: df['dis'] = dis
          df
Out[43]:
                             Qty
                                      dis
               x y
           0 5.0 3.0
                       Weather V-1 5.385165
           1 1.0 4.0
                       Weather V-1 3.162278
           2 4.0 2.0
                       Weather V-1 4.123106
                       Holiday V-2 1.000000
           3 1.0 1.0
           4 1.0 1.0
                       Holiday V-2 1.000000
           5 0.0 0.0
                       Holiday V-2 1.000000
                       Game V-3 0.000000
           6 0.0 1.0
           7 0.0 1.0
                       Game V-3 0.000000
           8 0.0 0.0
                       Game V-3 1.000000
           9 1.0 0.0 Weekday H-M 1.414214
In [44]: df.sort_values('dis')
```

```
Out[44]:
                x y
                                Qty
                                           dis
            6 0.0 1.0 Game V-3 0.000000
            7 0.0 1.0
                           Game V-3 0.000000
            3 1.0 1.0
                         Holiday V-2 1.000000
            4 1.0 1.0
                         Holiday V-2 1.000000
            5 0.0 0.0 Holiday V-2 1.000000
                          Game V-3 1.000000
            9 1.0 0.0 Weekday H-M 1.414214
            1 1.0 4.0 Weather V-1 3.162278
            2 4.0 2.0 Weather V-1 4.123106
            0 5.0 3.0 Weather V-1 5.385165
In [45]: df.to excel('D:/agnes/outputNo2(a).xls')
In [46]: #b.Apabila Cuaca baik dengan nilai 4, Weekend, dan Game =1, maka berapa roti yang harus dibuat?
## misalkan hari misterius = "H-M" (Weekenda)
data = np.array([[5.,3.,'Weather V-1'],[1.,4.,'Weather V-1'],[4.,2.,'Weather V-1'],[1.,1.,'Holiday V-2'],[1.,1.,'Holiday V-2']
query = [4.,1.,'Weekend H-M']
In [26]: for i in range(7):
                     if(df.iloc[i]['Qty'] == 'V1'):
                    \label{eq:plt.scatter} $$ plt.scatter(df.iloc[i]['x'], df.iloc[i]['y'], s=100, c='r') $$ elif(df.iloc[i]['Qty'] == 'V2'):
                          plt.scatter(df.iloc[i]['x'], df.iloc[i]['y'], s=100, c='y')
                      plt.scatter(df.iloc[i]['x'], df.iloc[i]['y'], s=200, c='b')
              plt.grid()
              plt.show()
                0.0
                1.0
                2.0
                4.0
                3.0
                       5.0
                                         1.0
                                                             4.0
                                                                                0.0
```

```
In [24]: #b.Apabila Cuaca baik dengan nilai 4, Weekend, dan Game =1, maka berapa roti yang harus dibuat?
        ## misalkan hari misterius = "NN"
        data = np.array([[5.,3.,'V1'],[1.,4.,'V1'],[4.,2.,'V1'],[1.,1.,'V2'],[1.,1.,'V2'],[0.,0.,'V2'],[4.,1.,'NN']])
        query = [4.,1.,'NN']
In [25]: df = pd.DataFrame(data)
        df.columns = ['x','y','Qty']
        df
Out[25]:
           x y Qty
        0 5.0 3.0 V1
        1 1.0 4.0 V1
        2 4.0 2.0 V1
        3 1.0 1.0 V2
        4 1.0 1.0 V2
        5 0.0 0.0 V2
        6 4.0 1.0 NN
 In [37]: import math
         dis = []
         for i in range(7):
            In [38]: df['dis'] = dis
         df
 Out[38]:
             x y Qty
         0 5.0 3.0 V1 2.236068
         1 1.0 4.0 V1 4.242641
         2 4.0 2.0 V1 1.000000
         3 1.0 1.0 V2 3.000000
         4 1.0 1.0 V2 3.000000
         5 0.0 0.0 V2 4.123106
         6 4.0 1.0 NN 0.000000
 In [39]: df.sort_values('dis')
In [39]: df.sort_values('dis')
Out[39]:
             x y Qty
                            dis
         6 4.0 1.0 NN 0.000000
          2 4.0 2.0 V1 1.000000
          0 5.0 3.0 V1 2.236068
          3 1.0 1.0 V2 3.000000
          4 1.0 1.0 V2 3.000000
          5 0.0 0.0 V2 4.123106
          1 1.0 4.0 V1 4.242641
In [40]: df.to excel('D:/agnes/outputNo2(b).xls')
In [ ]:
```

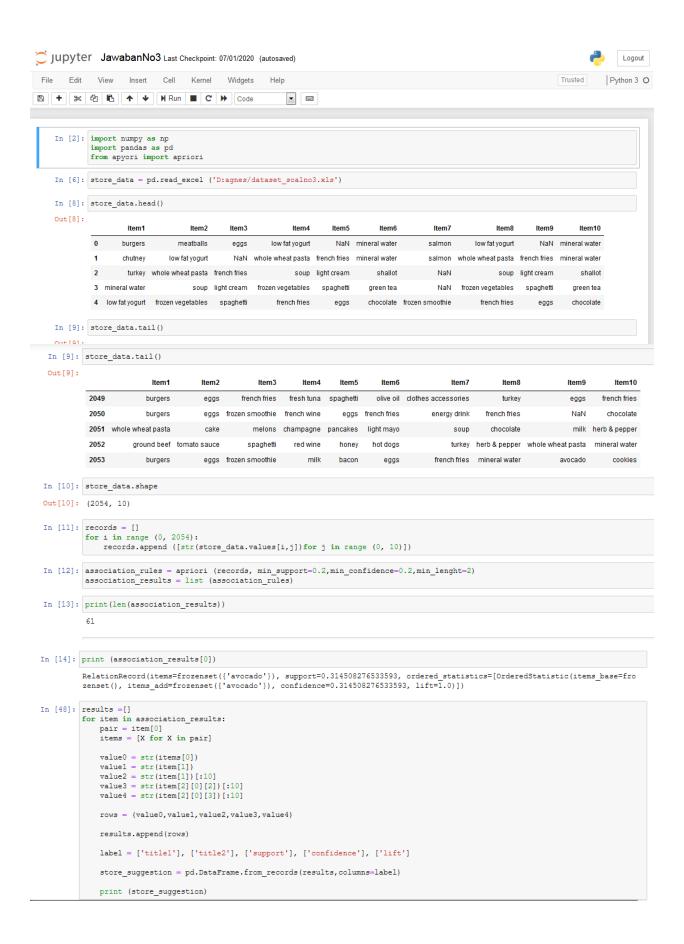
OUTPUT NO2

a.

	A1	•	(0	Ĵж		
4	Α	В	С	D	Е	F
1		X	у	Qty	dis	
2	0	5.0	3.0	Weather V	5.385165	
3	1	1.0	4.0	Weather V	3.162278	
4	2	4.0	2.0	Weather V	4.123106	
5	3	1.0	1.0	Holiday V-	1	
6	4	1.0	1.0	Holiday V-	1	
7	5	0.0	0.0	Holiday V-	1	
8	6	0.0	1.0	Game V-3	0	
9	7	0.0	1.0	Game V-3	0	
10	8	0.0	0.0	Game V-3	1	
11	9	1.0	0.0	Weekday I	1.414214	
12						
12						

b.

	A1	•	(0	f_{x}		
4	Α	В	С	D	Е	F
1		X	у	Qty	dis	
2	0	5.0	3.0	Weather V	4.123106	
3	1	1.0	4.0	Weather V	0	
4	2	4.0	2.0	Weather V	3.605551	
5	3	1.0	1.0	Holiday V-	3	
6	4	1.0	1.0	Holiday V-	3	
7	5	0.0	0.0	Holiday V-	4.123106	
8	6	0.0	1.0	Game V-3	3.162278	
9	7	0.0	1.0	Game V-3	3.162278	
10	8	0.0	0.0	Game V-3	4.123106	
11	9	4.0	1.0	Weekend I	4.242641	
12						
13						



```
(support,) (confidence,) (lift,)
           (title1,)
                             (title2,)
           avocado 0.314508276533593 0.31450827
                                                    0.31450827
                                                                    1.0
          (title1,)
                              (title2,) (support,) (confidence,) (lift,)
                                                      0.31450827
                                                                    1.0
            avocado
                       0.314508276533593 0.31450827
            burgers 0.24294060370009737 0.24294060
                                                      0.24294060
                                                                      1.0
                                (title2,) (support,) (confidence,) (lift,)
            (title1,)
                        0.314508276533593 0.31450827
            avocado
                                                      0.31450827
                                                                    1.0
             burgers 0.24294060370009737 0.24294060
                                                        0.24294060
                                                                       1.0
         2 chocolate
                      0.4756572541382668 0.47565725
                                                       0.47565725
                                                                       1.0
                                 (title2,) (support,) (confidence,) (lift,) 0.314508276533593 0.31450827 0.31450827 1.0
                     (title1,)
         0
                       avocado
                       burgers 0.24294060370009737 0.24294060
                                                                  0.24294060
                     chocolate 0.4756572541382668 0.47565725
                                                                  0.47565725
                                                                                 1.0
          clothes accessories 0.33982473222979553 0.33982473
                                                                  0.33982473
         3
                                                                                1.0
                                                    (support,) (confidence,) (lift,)
                                          (title2.)
                     (title1.)
                                  0.314508276533593
        0
                       avocado
                                                   0.31450827
                                                                  0.31450827
                                                                                1.0
        1
                       burgers 0.24294060370009737 0.24294060
                                                                  0.24294060
                                                                                1.0
         2
                                0.4756572541382668
                                                    0.47565725
                                                                  0.47565725
                     chocolate
           clothes accessories 0.33982473222979553 0.33982473
                                                                  0.33982473
In [49]: store_suggestion.describe()
                        chocolate 0.23661148977604674 0.23661148
                                                                        0.23661148
          59
                                                                                        1.0
                         (title1,)
                                               (title2,)
                                                          (support,) (confidence,) (lift,)
          0
                                      0.314508276533593 0.31450827
                                                                      0.31450827
                                                                                        1.0
                          avocado
          1
                          burgers 0.24294060370009737 0.24294060
                                                                       0.24294060
                                                                                        1.0
                        chocolate
                                    0.4756572541382668 0.47565725
                                                                        0.47565725
                                                                                        1.0
              clothes accessories 0.33982473222979553 0.33982473
                                                                        0.33982473
          3
                                                                                        1.0
          4
                          cookies 0.3588120740019474 0.35881207
                                                                        0.35881207
                              milk 0.20837390457643623 0.20837390
          56
                                                                        0.20837390
                                                                                        1.0
          57
                                    0.2249269717624148 0.22492697
                    mineral water
                                                                        0.22492697
                                                                                        1.0
                     french fries 0.22249269717624148 0.22249269
          58
                                                                        0.22249269
                                                                                        1.0
                        chocolate 0.23661148977604674 0.23661148
          59
                                                                         0.23661148
                                                                                        1.0
          60 clothes accessories 0.24196689386562803 0.24196689
                                                                        0.24196689
                                                                                        1.0
          [61 rows x 5 columns]
In [49]: store_suggestion.describe()
Out[49]:
                    (title1,)
                                      (title2,)
                                               (support,) (confidence,) (lift,)
                       61
           count
           unique
                       15
                                          53
                                                    53
                                                               53
                                                                     1
             top french fries 0.24294060370009737 0.24294060
                                                        0.24294060
                                                                    1.0
                       12
             frea
                                                     4
                                                                4 61
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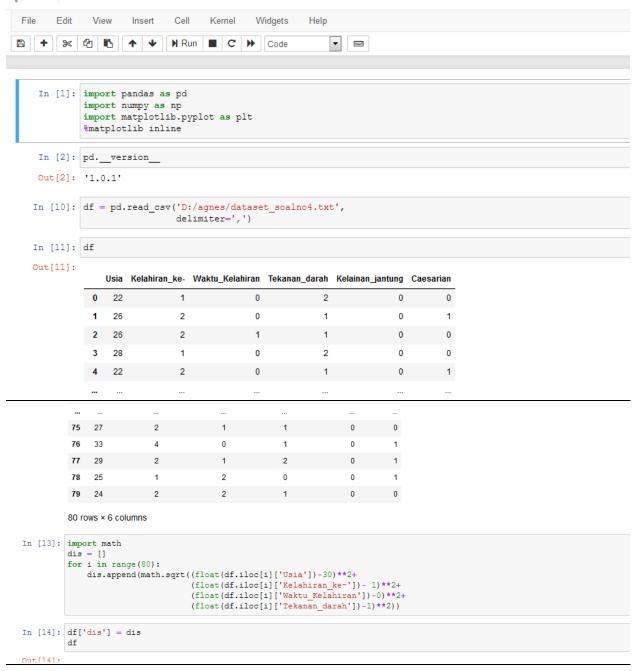
In [50]: store suggestion.to excel('D:/agnes/outputno3.xls')

OUTPUTNO3

4	Α	В	С	D	Е	F
1		('title1',)	('title2',)	('support',)	onfidence	('lif
2	0	avocado	0.3145082	0.3145082	0.3145082	1.0
3	1	burgers	0.2429406	0.2429406	0.2429406	1.0
4	2				0.4756572	
5	3				0.3398247	
6	4	cookies	0.3588120	0.3588120	0.3588120	1.0
7	5				0.4099318	
8	6	energy drir	0.3213242	0.3213242	0.3213242	1.0
9	7	french fries	0.6548198	0.6548198	0.6548198	1.0
10	8				0.3042843	
11	9				0.4079844	1.0
12	10	mineral wa	0.4527750	0.4527750	0.4527750	1.0
13	11				0.6285296	
14	12				0.2151898	
15	13	turkey	0.5272638	0.5272638	0.5272638	1.0
16	14				0.2653359	
17	15				0.2030185	
18	16	mineral wa	0.3037974	0.3037974	0.3037974	1.0
19	17				0.2921129	
20	18				0.2521908	
21	19	chocolate	0.3042843	0.3042843	0.3042843	1.0
22	20				0.2711781	
23	21				0.3763388	1.0
24	22	clothes ac	0.2512171	0.2512171	0.2512171	1.0
25	23				0.3237585	
26	24				0.2108081	
27	25	clothes ac	0.3281402	0.3281402	0.3281402	1.0
28	26	french fries	0.3213242	0.3213242	0.3213242	1.0
29	27	mineral wa	0.2005842	0.2005842	0.2005842	1.0
30	28	nan	0.2429406	0.2429406	0.2429406	1.0

33 31 eggs			_	_	_	_		
35 33 french fries 0.3033106 0.3033106 0.3033106 1.0 36 34 nan 0.3033106 0.3033106 0.3033106 1.0 37 35 milk 0.2804284 0.2804284 0.2804284 1.0 38 36 mineral wa 0.2702044 0.2702044 0.2702044 1.0 39 37 french fries 0.4819863 0.4819863 0.4819863 1.0 40 38 french fries 0.3709834 0.3709834 0.3709834 1.0 41 39 herb & per 0.2478091 0.2478091 0.2478091 1.0 42 40 milk 0.2702044 0.2702044 1.0 43 41 mineral wa 0.3476144 0.3476144 1.0 44 42 nan 0.2502434 0.2502434 0.2502434 1.0 45 43 mineral wa 0.2891918 0.2891918 0.2891918 1.0 46 44 chocolate 0.2429406 0.2429406 0.2429406 1.0 47 45 chocolate 0.2453748 0.2453748 1.0 48 46 chocolate 0.2453748 0.2453748 1.0 49 47 clothes ac 0.2429406 0.2429406 0.2429406 1.0 50 48 clothes ac 0.2429406 0.2429406 0.2429406 1.0 51 49 clothes ac 0.2497565 0.2497565 1.0 52 50 clothes ac 0.204247 0.202044 7 0.202044 1.0 53 51 clothes ac 0.20447 0.202044 7 0.202044 1.0 54 52 french fries 0.215189 0.215189 0.215189 1.0 55 53 french fries 0.215189 0.215189 0.215189 1.0 56 54 french fries 0.215189 0.215189 0.215189 1.0 57 55 french fries 0.2215189 0.2215189 0.2215189 1.0 58 56 milk 0.2033739 0.2033739 1.0 59 57 mineral wa 0.2249269 0.2249269 0.2249269 1.0 50 clothes ac 0.2366114 0.2366114 0.2366114 1.0 51 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 52 60 clothes ac 0.2249266 0.2249266 0.2249266 1.0 58 56 french fries 0.2249266 0.2249266 0.2249266 1.0 59 chocolate 0.2366114 0.2366114 0.2366114 1.0	33	31	eggs	0.2409931 (0.2409931	0.2409931	1.0	
36 34 nan	34	32						
37 35 milk 0.2804284 0.2804284 0.2804284 1.0 38 36 mineral wa 0.2702044 0.2702044 1.0 39 37 french fries 0.4819863 0.4819863 0.4819863 1.0 40 38 french fries 0.3709834 0.3709834 1.0 41 39 herb & per 0.2478091 0.2478091 0.2478091 1.0 42 40 milk 0.2702044 0.2702044 1.0 43 41 mineral wa 0.3476144 0.3476144 1.0 44 42 nan 0.2502434 0.2502434 1.0 45 43 mineral wa 0.2891918 0.2891918 1.0 46 44 chocolate 0.2429406 0.2429406 1.0 47 45 chocolate 0.2429406 0.2429406 1.0 48 46 chocolate 0.2453748 0.2453748 0.2453748 1.0 49 47 clothes ac 0.2429406 0.2429406 1.0 50 48 clothes ac 0.2429406 0.2429406 1.0 51 49 clothes ac 0.2429406 0.2429406 0.2429406 1.0 52 50 clothes ac 0.2429406 0.2429406 0.2429406 1.0 53 51 clothes ac 0.2020447 0.2020447 0.2020447 1.0 54 52 french fries 0.215189 0.215189 0.215189 1.0 55 53 french fries 0.2158978 0.2155978 1.0 56 54 french fries 0.215978 0.215189 0.215189 1.0 57 55 french fries 0.293739 0.2083739 1.0 58 56 milk 0.2083739 0.2083739 0.2083739 1.0 59 57 mineral wa 0.2492668 0.2249266 0.2249265 1.0 50 58 french fries 0.2366114 0.2366114 1.0 51 59 chocolate 0.2366114 0.2366114 1.0 52 60 clothes ac 0.22492668 0.2249266 0.2249265 1.0 53 51 clothes ac 0.2249406 0.215189 0.2215189 1.0 54 55 french fries 0.2248266 0.2249266 0.2249266 1.0 55 53 french fries 0.2366114 0.2366114 0.2366114 1.0	35	33						
38 36 mineral wa ⁰ .2702044 ⁶ .2702044 ⁷ .2702044 ⁷ .10 39 37 french fries ⁶ .4819863 ⁷ .4819863 ⁷ .4819863 ⁷ .10 40 38 french fries ⁶ .3709834 ⁷ .3709834 ⁷ .3709834 ⁷ .10 41 39 herb & per ⁶ .2478091 ⁷ .2478091 ⁷ .2478091 ⁷ .0 42 40 milk	36	34	nan	0.3033106	0.3033106	0.3033106	1.0	
39 37 french fries 0.4819863 0.4819863 1.0 4819863 1.0 38 french fries 0.3709834 0.3709834 1.0 3709834 1.0 41 39 herb & per 0.2478091 0.2478091 0.2478091 1.0 42 40 milk 0.2702044 0.2702044 0.2702044 1.0 43 41 mineral wa 0.3476144 0.3476144 0.3476144 1.0 44 42 nan 0.2502434 0.2502434 0.2502434 1.0 44 42 nan 0.2502434 0.2502434 0.2502434 1.0 45 43 mineral wa 0.2891918 0.2891918 0.2891918 1.0 46 44 chocolate 0.2429406 0.2429406 0.2429406 1.0 47 45 chocolate 0.2453748 0.2453748 0.2453748 1.0 46 46 chocolate 0.2833495 0.2833495 1.0 47 clothes ac 0.2429406 0.2429406 0.2429406 1.0 48 clothes ac 0.2497565 0.2497565 0.2497565 1.0 49 clothes ac 0.2497565 0.2497565 0.2497565 1.0 52 50 clothes ac 0.2020447 0.2020447 0.2020447 1.0 52 51 clothes ac 0.2020447 0.2020447 0.2020447 1.0 52 52 french fries 0.2215189 0.2215189 0.2215189 1.0 53 51 clothes ac 0.2020447 0.2020447 0.2020447 1.0 54 52 french fries 0.2185978 0.2185978 1.0 55 53 french fries 0.2185978 0.2185978 0.2185978 1.0 56 54 french fries 0.2911392 0.2911392 0.2911392 0.591392 0.593359 0.5653359 1.0 57 55 french fries 0.2249269 0.2249269 0.2249269 1.0 58 french fries 0.2249269 0.2249269 0.2249269 1.0 58 french fries 0.2249268 0.2249269 0.2249269 1.0 58 french fries 0.2249268 0.224926 0.2249269 1.0 58 french fries 0.2366114 0.2366114 0.2366114 1.0 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 50 clothes ac 0.2419668 0.2419668 0.2419668 1.0	37	35	milk	0.2804284 (0.2804284	0.2804284	1.0	
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42 40 milk	40	38						
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44 42 nan	42	40						
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46 44 chocolate 0.2429406 0.2429406 0.2429406 1.0 47 45 chocolate 0.2453748 0.2453748 0.2453748 1.0 48 46 chocolate 0.2833495 0.2833495 1.0 49 47 clothes ac 0.2429406 0.2429406 0.2429406 1.0 50 48 clothes ac 0.2497565 0.2497565 0.2497565 1.0 51 49 clothes ac 0.20426 0.2064264 0.2064264 1.0 52 50 clothes ac 0.3154819 0.3154819 1.0 53 51 clothes ac 0.2020447 0.2020447 1.0 54 52 french fries 0.2215189 0.2215189 0.2215189 1.0 55 53 french fries 0.2185978 0.2185978 1.0 56 54 french fries 0.2185978 0.2185978 0.2185978 1.0 57 55 french fries 0.2911392 0.2911392 0.2911392 1.0 58 56 milk 0.2083739 0.2083739 0.2083739 1.0 59 57 mineral wa 0.2249269 0.2249269 0.2249269 1.0 60 58 french fries 0.22366114 0.2366114 0.2366114 1.0 61 59 chocolate 0.2366114 0.2366114 0.2366114 1.0	44	42	nan	0.2502434	0.2502434	0.2502434	1.0	
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48	46	44	chocolate	0.2429406	0.2429406	0.2429406	1.0	
49 47 clothes ac 0.2429406 0.2429406 0.2429406 1.0 50 48 clothes ac 0.2497565 0.2497565 1.0 51 49 clothes ac 0.2064264 0.2064264 0.2064264 1.0 52 50 clothes ac 0.3154819 0.3154819 1.0 53 51 clothes ac 0.2020447 0.2020447 0.2020447 1.0 54 52 french fries 0.2215189 0.2215189 0.2215189 1.0 55 53 french fries 0.215189 0.215189 0.215189 1.0 56 54 french fries 0.2653359 0.2653359 1.0 57 55 french fries 0.291392 0.2911392 0.2911392 1.0 58 56 milk 0.2083739 0.2083739 0.2083739 1.0 59 57 mineral wa 0.2249269 0.2249269 0.2249269 1.0 60 58 french fries 0.2236114 0.2366114 0.2366114 1.0 61 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 62 60 clothes ac 0.2419668 0.2419668 1.0	47	45	chocolate	0.2453748	0.2453748	0.2453748	1.0	
50 48 clothes ac 0.2497565 0.2497565 0.2497565 1.0 51 49 clothes ac 0.2064264 0.2064264 0.2064264 1.0 52 50 clothes ac 0.3154819 0.3154819 0.3154819 1.0 53 51 clothes ac 0.2020447 0.2020447 1.0 54 52 french fries 0.2215189 0.2215189 0.2215189 1.0 55 53 french fries 0.2185978 0.2185978 0.2185978 1.0 56 54 french fries 0.2653359 0.2653359 1.0 57 55 french fries 0.2911392 0.2911392 0.2911392 1.0 58 56 milk 0.2083739 0.2083739 0.2083739 1.0 59 57 mineral wa 0.2249269 0.2249269 1.249269 1.0 60 58 french fries 0.2224926 0.2224926 1.0 61 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 62 60 clothes ac 0.2419668 0.2419668 0.2419668 1.0	48	46	chocolate	0.2833495 0).2833495	0.2833495	1.0	
51	49	47	clothes ac	0.2429406	0.2429406	0.2429406	1.0	
52 50 clothes ac 0.3154819 0.3154819 0.3154819 1.0 53 51 clothes ac 0.2020447 2.020447 2.020447 1.0 54 52 french fries 0.2215189 0.2215189 2.215189 1.0 55 53 french fries 0.2185978 2.215189 2.215189 1.0 56 54 french fries 0.2653359 2.2653359 2.2653359 2.0 57 55 french fries 0.2911392 2.02911392 2.02911392 1.0 58 56 milk 0.2083739 0.2083739 0.2083739 1.0 59 57 mineral wa 0.2249269 0.2249269 2.249269 1.0 60 58 french fries 0.2224926 0.2224926 1.0 61 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 62 60 clothes ac 0.2419668 0.2419668 0.2419668 1.0	50	48	clothes ac	0.2497565	0.2497565	0.2497565	1.0	
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56 54 french fries 0.2653359 0.2653359 0.2653359 1.0 57 55 french fries 0.2911392 0.2911392 0.2911392 1.0 58 56 milk 0.2083739 0.2083739 0.2083739 1.0 59 57 mineral wa 0.2249269 0.2249269 0.224926 1.0 60 58 french fries 0.2224926 0.2224926 0.2224926 1.0 61 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 62 60 clothes ac 0.2419668 0.2419668 0.2419668 1.0	54	52						
57 55 french fries 0.2911392 0.2911392 0.2911392 1.0 58 56 milk 0.2083739 0.2083739 0.2083739 1.0 59 57 mineral wa 0.2249269 0.2249269 0.2249269 1.0 60 58 french fries 0.2224926 0.2224926 0.2224926 1.0 61 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 62 60 clothes ac 0.2419668 0.2419668 0.2419668 1.0	55	53						
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60 58 french fries 0.2224926 0.2224926 0.2224926 1.0 61 59 chocolate 0.2366114 0.2366114 1.0 62 60 clothes ac 0.2419668 0.2419668 0.2419668 1.0	58	56	milk	0.2083739 (0.2083739	0.2083739	1.0	
61 59 chocolate 0.2366114 0.2366114 0.2366114 1.0 62 60 clothes ac 0.2419668 0.2419668 0.2419668 1.0	59	57						
62 60 clothes ac 0.2419668 0.2419668 0.2419668 1.0	60	58						
	61	59	chocolate	0.2366114 (0.2366114	0.2366114	1.0	
4 4 b bl Shoot1 / \$77	62	60	clothes ac	0.2419668	.2419668	0.2419668	1.0	
	4 4	▶ H Sh	oot1 🐬	/				

Jupyter JawabanNo4 Last Checkpoint: 7 minutes ago (autosaved)



Usia Kelahiran_ke- Waktu_Kelahiran Tekanan_darah Kelainan_jantung Caesarian dis 0 8.062258 1 4.123106 0 4.242641 0 2.236068 1 8.062258 0 3.316625 1 4.242641 1 2.000000 1 5.477226 0 6.403124 80 rows × 7 columns In [15]: df.sort_values('dis') In [15]: df.sort_values('dis') Out[15]: Usia Kelahiran_ke- Waktu_Kelahiran Tekanan_darah Kelainan_jantung Caesarian dis 30 0.000000

1.000000 0 1.414214 1 1.414214 1 1.732051 1 11.000000 1 11.000000 0 12.000000 1 12.083046 1 13.038405

80 rows × 7 columns

In [16]: y = df.sort_values('dis').head(5)
y

```
Out[16]:
             Usia Kelahiran_ke- Waktu_Kelahiran Tekanan_darah Kelainan_jantung Caesarian
                                                                             0.000000
          38
                                         0
                                                      1
                                                                    0
                                                                             0 1.000000
              29
                           2
                                                                             0 1.414214
          54
                           2
                                         0
                                                      1
                                                                     1
                                                                             1 1.414214
                           2
                                                                             1 1.732051
In [17]: z = y["Caesarian"]
Out[17]: 27
         38
         67
               0
         54
         59
         Name: Caesarian, dtype: int64
In [18]: np.mean(z)
Out[18]: 0.4
In [19]: df.to_excel('D:/agnes/outputNo4.xls')
In [ ]:
```

OUTPUTNO4

4	Α	В	С	D	Е	F	G	Н
1		Usia	Kelahiran_ke-	Waktu_Kelahiran	Tekanan_darah	Kelainan_jantung	Caesarian	dis
2	0	22	1	0	2	0	0	8.062257748
3	1	26	2	0	1	0	1	4.123105626
4	2	26	2	1	1	0	0	4.242640687
5	3	28	1	0	2	0	0	2.236067977
6	4	22	2	0	1	0	1	8.062257748
7	5	26	1	1	0	0	0	4.242640687
3	6	27	2	0	1	0	0	3.16227766
9	7	32	3	0	1	0	1	2.828427125
0	8	28	2	0	1	0	0	2.236067977
1	9	27	1	1	1	0	1	3.16227766
2	10	36	1	0	1	0	0	6
3	11	33	1	1	0	0	1	3.31662479
4	12	23	1	1	1	0	0	7.071067812
5	13	20	1	0	1	1	0	10
6	14	29	1	2	0	1	1	2.449489743
7	15	25	1	2	0	0	0	5.477225575
8	16	25	1	0	1	0	0	5
9	17	20	1	2	2	0	1	10.24695077
:0	18	37	3	0	1	1	1	7.280109889
!1	19	24	1	2	0	1	1	6.403124237
2	20	26	1	1	1	0	0	4.123105626
:3	21	33	2	0	0	1	1	3.31662479
4	22	25	1	1	2	0	0	5.196152423
!5	23	27	1	0	0	1	1	3.16227766
16	24	20	1	0	2	1	1	10.04987562
.7	25	18	1	0	1	0	0	12
18	26	18	1	1	2	1	1	12.08304597
!9	27	30	1	0	1	0	0	0
0	28	32	1	0	2	1	1	2.236067977
	v 11 01		1200					

50	33	3	2	1	1	0	4.123105626
51	21	2	1	0	1	1	9.16515139
52	30	3	2	2	0	0	3
53	35	1	1	0	0	0	5.196152423
54	29	2	0	1	1	1	1.414213562
55	25	2	0	1	0	0	5.099019514
56	32	3	1	0	1	1	3.16227766
57	21	1	0	0	0	1	9.055385138
58	26	1	0	2	0	1	4.123105626
59	30	2	1	2	1	1	1.732050808
60	22	1	2	2	0	0	8.306623863
61	19	1	0	1	0	1	11
62	32	2	0	0	0	1	2.449489743
63	32	2	0	1	1	1	2.236067977
64	31	1	2	2	1	0	2.449489743
65	35	2	0	1	0	1	5.099019514
66	28	3	0	1	0	1	2.828427125
67	29	2	0	1	1	0	1.414213562
68	25	1	0	0	0	1	5.099019514
69	27	2	2	0	0	0	3.872983346
70	17	1	0	0	0	1	13.03840481
71	29	1	2	0	1	1	2.449489743
72	28	2	0	1	0	0	2.236067977
73	32	3	0	1	1	0	2.828427125
74	38	3	2	2	1	1	8.544003745
75	27	2	1	1	0	0	3.31662479
76	33	4	0	1	0	1	4.242640687
77	29	2	1	2	0	1	2
78	25	1	2	0	0	1	5.477225575
79	24	2	2	1	0	0	6.403124237