Python Table Management and Data Preprocessing

Pandas memiliki banyak kegunaan, pandas mampu menyajikan hal-hal pengolahan data yang rumit menjadi sederhana, membantu mempercepat proses penyajian data dan analasis data

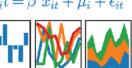
Pandas

Pandas adalah paket Python yang menyediakan struktur data yang cepat, fleksibel, dan ekspresif yang dirancang untuk bekerja dengan data terstruktur (tabular, multidimensi, berpotensi heterogen) dan deret waktu, mudah dan intuitif.



Hal ini bertujuan untuk menjadi blok penyusun tingkat tinggi yang fundamental untuk melakukan analisis data dunia nyata yang praktis dengan Python. Selain itu, ia memiliki tujuan yang lebih luas untuk menjadi alat analisis / manipulasi data open source yang paling kuat dan fleksibel yang tersedia dalam bahasa apa pun.





	BandName	WavelengthMax	WavelengthMin
0	CoastalAerosol	450	430
1	Blue	510	450
2	Green	590	530
3	Red	670	640
4	NearInfrared	880	850
5	ShortWaveInfrared_1	1650	1570
6	ShortWaveInfrared_2	2290	2110
7	Cirrus	1380	1360

Pandas DataFrame

Membuat Tabel dengan dataFrame

```
In [1]:
    import pandas as pd

    data = {
        "calories": [420, 380, 390],
        "duration": [50,40,45],
        "size": [80,75,60]
    }

    #Load data into a DataFrame object:
    df = pd.DataFrame(data)

    print(df)
```

```
calories duration size
0 420 50 80
1 380 40 75
2 390 45 60
```

Untuk melihat data pada baris ke 0

```
In [2]: print(df.loc[1])

calories 380
```

duration 40 size 75 Name: 1, dtype: int64

Untuk melihat baris ke 0 dan 1

```
In [3]:
         print(df.iloc[0:2])
            calories duration size
        0
                 420
                            50
                                  80
                            40
        1
                 380
                                  75
        Mengganti nama baris 0, 1 dan 2 menjadi day1, day2, dan day3
In [4]:
         df = pd.DataFrame(data, index = ["day1", "day2", "day3"])
         print(df)
               calories duration size
        day1
                   420
                         50
                                      80
        day2
                    380
                               40
                                      75
                    390
                               45
        day3
                                      60
        Memanggil data baris day2
In [5]:
         print(df.loc["day2"])
                     380
         calories
        duration
                      40
                      75
        size
        Name: day2, dtype: int64
        Memisahkan beberapa kolom dari tabel
In [6]:
         result = df.loc[:, ["calories","duration" ]]
         result
Out[6]:
              calories duration
                            50
         day1
                  420
         day2
                  380
                            40
                  390
                            45
         day3
```

Load File

2998

-117.12

34.10

Pastikan file data CSV ada di folder yang sama dengan file Python.

```
In [7]:
         import pandas as pd
         df = pd.read csv('california housing test.csv')
         print(df)
              longitude latitude housing_median_age total_rooms total_bedrooms \
        0
                -122.05
                         37.37
                                                  27.0
                                                             3885.0
                                                                              661.0
        1
                -118.30
                            34.26
                                                  43.0
                                                             1510.0
                                                                              310.0
        2
                -117.81
                            33.78
                                                  27.0
                                                             3589.0
                                                                              507.0
        3
                -118.36
                            33.82
                                                  28.0
                                                               67.0
                                                                               15.0
        4
                -119.67
                            36.33
                                                  19.0
                                                             1241.0
                                                                              244.0
        . . .
                    . . .
                             . . .
                                                  . . .
                                                               . . .
                                                                                . . .
                           34.42
                -119.86
                                                             1450.0
                                                                              642.0
        2995
                                                  23.0
        2996
                -118.14
                           34.06
                                                  27.0
                                                             5257.0
                                                                             1082.0
        2997
                -119.70
                            36.30
                                                  10.0
                                                             956.0
                                                                              201.0
```

40.0

96.0

14.0

```
population households median_income median_house_value
          0
                    1537.0
                                  606.0
                                                6.6085
                                                                   344700.0
          1
                     809.0
                                  277.0
                                                3.5990
                                                                   176500.0
          2
                    1484.0
                                  495.0
                                                5.7934
                                                                   270500.0
          3
                      49.0
                                  11.0
                                                6.1359
                                                                   330000.0
          4
                     850.0
                                  237.0
                                                2.9375
                                                                    81700.0
                       . . .
                                   . . .
                    1258.0
          2995
                                  607.0
                                                1.1790
                                                                   225000.0
          2996
                    3496.0
                                 1036.0
                                                3.3906
                                                                   237200.0
          2997
                     693.0
                                  220.0
                                                2.2895
                                                                    62000.0
          2998
                      46.0
                                  14.0
                                                3.2708
                                                                   162500.0
          2999
                     753.0
                                  260.0
                                                8.5608
                                                                   500001.0
          [3000 rows x 9 columns]
 In [8]:
          print(df.head(10))
             longitude latitude housing_median_age total_rooms total_bedrooms \
               -122.05
                           37.37
                                                             3885.0
         0
                                                 27.0
                                                                               661.0
         1
               -118.30
                           34.26
                                                 43.0
                                                             1510.0
                                                                               310.0
          2
               -117.81
                           33.78
                                                 27.0
                                                             3589.0
                                                                               507.0
          3
               -118.36
                           33.82
                                                 28.0
                                                               67.0
                                                                               15.0
          4
               -119.67
                           36.33
                                                 19.0
                                                             1241.0
                                                                               244.0
          5
               -119.56
                           36.51
                                                 37.0
                                                             1018.0
                                                                               213.0
          6
               -121.43
                           38.63
                                                 43.0
                                                             1009.0
                                                                               225.0
          7
               -120.65
                           35.48
                                                 19.0
                                                             2310.0
                                                                               471.0
          8
               -122.84
                           38.40
                                                 15.0
                                                             3080.0
                                                                               617.0
          9
               -118.02
                           34.08
                                                 31.0
                                                             2402.0
                                                                               632.0
             population households median_income median_house_value
                                             6.6085
         0
                              606.0
                 1537.0
                                                                344700.0
         1
                  809.0
                              277.0
                                             3.5990
                                                                176500.0
          2
                 1484.0
                              495.0
                                             5.7934
                                                                270500.0
                                                                330000.0
          3
                   49.0
                               11.0
                                             6.1359
          4
                              237.0
                  850.0
                                             2.9375
                                                                 81700.0
          5
                  663.0
                              204.0
                                             1.6635
                                                                 67000.0
                              218.0
          6
                  604.0
                                             1.6641
                                                                 67000.0
          7
                              441.0
                 1341.0
                                             3.2250
                                                                166900.0
          8
                 1446.0
                              599.0
                                             3.6696
                                                                194400.0
          9
                 2830.0
                              603.0
                                             2.3333
                                                                164200.0
 In [9]:
          print(df.tail())
                longitude latitude housing_median_age total_rooms total_bedrooms
          2995
                  -119.86
                               34.42
                                                     23.0
                                                                1450.0
                                                                                  642.0
          2996
                  -118.14
                               34.06
                                                     27.0
                                                                5257.0
                                                                                 1082.0
          2997
                  -119.70
                               36.30
                                                     10.0
                                                                 956.0
                                                                                  201.0
          2998
                  -117.12
                              34.10
                                                     40.0
                                                                  96.0
                                                                                   14.0
          2999
                  -119.63
                              34.42
                                                     42.0
                                                                1765.0
                                                                                  263.0
                population households median income median house value
          2995
                    1258.0
                                 607.0
                                                1.1790
                                                                   225000.0
          2996
                    3496.0
                                 1036.0
                                                3.3906
                                                                   237200.0
          2997
                     693.0
                                  220.0
                                                2.2895
                                                                    62000.0
          2998
                      46.0
                                  14.0
                                                3.2708
                                                                   162500.0
          2999
                                  260.0
                     753.0
                                                8.5608
                                                                   500001.0
In [10]:
          print(df.info())
```

42.0

1765.0

263.0

2999

-119.63

34.42

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 3000 entries, 0 to 2999 Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
0	longitude	3000 non-null	float64
1	latitude	3000 non-null	float64
2	housing_median_age	3000 non-null	float64
3	total_rooms	3000 non-null	float64
4	total_bedrooms	3000 non-null	float64
5	population	3000 non-null	float64
6	households	3000 non-null	float64
7	median_income	3000 non-null	float64
8	median_house_value	3000 non-null	float64

dtypes: float64(9)
memory usage: 211.1 KB

None

In [11]: df.describe()

Out[11]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	ł
count	3000.000000	3000.00000	3000.000000	3000.000000	3000.000000	3000.000000	
mean	-119.589200	35.63539	28.845333	2599.578667	529.950667	1402.798667	
std	1.994936	2.12967	12.555396	2155.593332	415.654368	1030.543012	
min	-124.180000	32.56000	1.000000	6.000000	2.000000	5.000000	
25%	-121.810000	33.93000	18.000000	1401.000000	291.000000	780.000000	
50%	-118.485000	34.27000	29.000000	2106.000000	437.000000	1155.000000	
75%	-118.020000	37.69000	37.000000	3129.000000	636.000000	1742.750000	
max	-114.490000	41.92000	52.000000	30450.000000	5419.000000	11935.000000	
4							•

Data pre-Processing

Dalam pre processing data terdapat beberapa aspek berikut ini:

- 1. missing values
- 2. data standardization
- 3. data normalization
- 4. data binning

Import Data

```
import pandas as pd
    df = pd.read_csv('hepatitis_csv.csv')
    df.head(10)
```

Out[12]:		age	sex	steroid	antivirals	fatigue	malaise	anorexia	liver_big	liver_firm	spleen_palpable
	0	30	male	False	False	False	False	False	False	False	False
	1	50	female	False	False	True	False	False	False	False	False
	2	78	female	True	False	True	False	False	True	False	False

	age	sex	steroid	antivirals	fatigue	malaise	anorexia	liver_big	liver_firm	spleen_palpable
3	31	female	NaN	True	False	False	False	True	False	False
4	34	female	True	False	False	False	False	True	False	False
5	34	female	True	False	False	False	False	True	False	False
6	51	female	False	False	True	False	True	True	False	True
7	23	female	True	False	False	False	False	True	False	False
8	39	female	True	False	True	False	False	True	True	False
9	30	female	True	False	False	False	False	True	False	False
4										•

Identifikasi Missing Value

```
In [13]:
          df.isna().sum()
                               0
          age
Out[13]:
          sex
                               0
                               1
          steroid
          antivirals
                               0
          fatigue
                               1
          malaise
                               1
          anorexia
                              1
          liver_big
                              10
                              11
          liver_firm
          spleen_palpable
                              5
                               5
          spiders
                               5
          ascites
                               5
          varices
          bilirubin
                               6
                              29
          alk_phosphate
          sgot
                              4
          albumin
                              16
          protime
                              67
          histology
                               0
          class
                               0
          dtype: int64
```

Cetak dalam bentuk presentase

```
In [14]:
          df.isna().sum()/len(df)*100
                              0.000000
         age
Out[14]:
                              0.000000
          sex
          steroid
                              0.645161
          antivirals
                              0.000000
                              0.645161
          fatigue
         malaise
                              0.645161
          anorexia
                              0.645161
          liver_big
                              6.451613
          liver_firm
                              7.096774
          spleen_palpable
                              3.225806
          spiders
                              3.225806
          ascites
                              3.225806
          varices
                              3.225806
          bilirubin
                              3.870968
          alk_phosphate
                             18.709677
          sgot
                              2.580645
```

 albumin
 10.322581

 protime
 43.225806

 histology
 0.000000

 class
 0.000000

dtype: float64

Drop Missing Values

Menghampus semua kolom yang memiliki missing value

0.000000

0.000000

protime histology

```
In [15]:
            df.dropna( )
Out[15]:
                              steroid antivirals fatigue malaise anorexia liver_big liver_firm spleen_palpab
                 age
             5
                  34
                      female
                                 True
                                            False
                                                    False
                                                              False
                                                                        False
                                                                                   True
                                                                                             False
                                                                                                               Fal:
                                                              False
             10
                  39
                      female
                                 False
                                            True
                                                     False
                                                                        False
                                                                                  False
                                                                                              True
                                                                                                               Fal:
             11
                  32
                      female
                                 True
                                            True
                                                     True
                                                              False
                                                                        False
                                                                                              True
                                                                                                               Fal:
                                                                                   True
             12
                  41
                      female
                                 True
                                            True
                                                     True
                                                              False
                                                                        False
                                                                                   True
                                                                                              True
                                                                                                               Fal:
             13
                  30 female
                                 True
                                            False
                                                     True
                                                              False
                                                                        False
                                                                                   True
                                                                                              True
                                                                                                               Fal:
           139
                  45
                      female
                                 True
                                            True
                                                     False
                                                              False
                                                                        False
                                                                                   True
                                                                                             False
                                                                                                               Fal:
                  49
                                                                                                               Trι
           143
                      female
                                 False
                                            False
                                                     True
                                                              True
                                                                        False
                                                                                   True
                                                                                             False
                  31 female
                                            False
                                                              False
                                                                        False
                                                                                                               Fal:
           145
                                 False
                                                     True
                                                                                   True
                                                                                             False
           153
                  53
                        male
                                 False
                                            False
                                                     True
                                                              False
                                                                        False
                                                                                   True
                                                                                             False
                                                                                                               Trι
           154
                  43 female
                                            False
                                                     True
                                                              False
                                                                        False
                                                                                   True
                                                                                             False
                                                                                                               Trι
                                 True
          80 rows × 20 columns
          Drop baris dari spesifik kolom
In [16]:
            df = pd.read_csv('hepatitis_csv.csv')
            df.dropna(subset=['protime'],axis=0,inplace=True)
            df.isna().sum()/len(df)*100
                                  0.000000
           age
Out[16]:
                                  0.000000
           sex
           steroid
                                  1.136364
           antivirals
                                  0.000000
           fatigue
                                  0.000000
           malaise
                                  0.000000
           anorexia
                                  0.000000
           liver_big
                                  2.272727
           liver_firm
                                  2.272727
           spleen_palpable
                                  1.136364
           spiders
                                  1.136364
           ascites
                                  1.136364
           varices
                                  1.136364
           bilirubin
                                  0.000000
           alk_phosphate
                                  4.545455
           sgot
                                  0.000000
           albumin
                                  1.136364
```

class 0.000000

dtype: float64

In [17]: df.head(10)

Out[17]:

	age	sex	steroid	antivirals	fatigue	malaise	anorexia	liver_big	liver_firm	spleen_palpable
3	31	female	NaN	True	False	False	False	True	False	False
5	34	female	True	False	False	False	False	True	False	False
10	39	female	False	True	False	False	False	False	True	False
11	32	female	True	True	True	False	False	True	True	False
12	41	female	True	True	True	False	False	True	True	False
13	30	female	True	False	True	False	False	True	True	False
15	38	female	False	False	True	True	True	True	False	Fals€
17	40	female	False	False	True	False	False	True	True	False
18	38	female	True	False	False	False	False	True	False	False
19	38	female	False	True	False	False	False	False	True	False
4										>

Mengganti data Missing Values

Strategi yang baik saat menangani nilai yang hilang melibatkan penggantiannya dengan nilai lain. Biasanya, strategi berikut ini:

- 1. untuk nilai numerik ganti nilai yang hilang dengan nilai rata-rata kolom
- 2. untuk nilai kategorial ganti nilai yang hilang dengan nilai kolom yang paling sering
- 3. gunakan fungsi lain

Untuk mengganti nilai yang hilang, bisanya digunakan tiga fungsi seperti: fillna(), replace() dan interpolate().

```
In [18]:
          df = pd.read_csv('hepatitis_csv.csv')
          df.dtypes
                             int64
         age
Out[18]:
                            object
         sex
         steroid
                            object
         antivirals
                              bool
         fatigue
                            object
         malaise
                            object
                            object
         anorexia
         liver_big
                            object
         liver_firm
                            object
         spleen_palpable
                            object
         spiders
                            object
         ascites
                            object
         varices
                            object
         bilirubin
                           float64
         alk_phosphate
                           float64
                           float64
         sgot
         albumin
                           float64
                           float64
         protime
         histology
                              bool
```

Kolom Numerik

```
import numpy as np
numeric = df.select_dtypes(include=np.number)
numeric_columns = numeric.columns
```

Isi kolom numerik dengan nilai rataan

```
In [20]:
    df[numeric_columns] = df[numeric_columns].fillna(df.mean())
```

C:\Users\USER\AppData\Local\Temp/ipykernel_10040/3464485706.py:1: FutureWarning: Dro pping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns be fore calling the reduction.

df[numeric_columns] = df[numeric_columns].fillna(df.mean())

Melihat hasilnya adalah

```
In [21]:
          df.isna().sum()/len(df)*100
                            0.000000
Out[21]:
                            0.000000
         steroid
                            0.645161
         antivirals
                            0.000000
         fatigue
                            0.645161
         malaise
                           0.645161
         anorexia
                           0.645161
         liver_big
                           6.451613
                          7.096774
         liver_firm
         spleen_palpable 3.225806
         spiders
                            3.225806
         ascites
                           3.225806
         varices
                           3.225806
         bilirubin
                           0.000000
         alk_phosphate
                            0.000000
         sgot
                            0.000000
         albumin
                            0.000000
         protime
                            0.000000
         histology
                            0.000000
                            0.000000
         class
         dtype: float64
```

Kolom Kategorial

```
boolean_columns = df.select_dtypes(include=np.object).columns.tolist()
boolean_columns.remove('class')
df[boolean_columns] = df[boolean_columns].astype('bool')
```

C:\Users\USER\AppData\Local\Temp/ipykernel_10040/2487696438.py:1: DeprecationWarnin
g: `np.object` is a deprecated alias for the builtin `object`. To silence this warni
ng, use `object` by itself. Doing this will not modify any behavior and is safe.
Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/r
elease/1.20.0-notes.html#deprecations
 boolean_columns = df.select_dtypes(include=np.object).columns.tolist()

Gunakan fungsi mode() untuk mencari nilai yang sering muncul.

```
In [23]:
```

```
df[boolean_columns].fillna(df.mode())
```

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\cup	uι	- 1	\angle	0		
		-			-	

: _		sex	steroid	fatigue	malaise	anorexia	liver_big	liver_firm	spleen_palpable	spiders	ascites
	0	True	False	False	False	False	False	False	False	False	False
	1	True	False	True	False	False	False	False	False	False	False
	2	True	True	True	False	False	True	False	False	False	Fals€
	3	True	True	False	False	False	True	False	False	False	False
	4	True	True	False	False	False	True	False	False	False	False
	•••										
	150	True	True	True	True	True	True	False	False	True	Tru€
	151	True	True	True	False	False	True	True	False	False	False
	152	True	False	True	True	False	False	True	False	True	False
	153	True	False	True	False	False	True	False	True	True	False
	154	True	True	True	False	False	True	False	True	True	Tru€

155 rows × 11 columns

←

Melihat hasilnya

```
In [24]:
           df.isna().sum()/len(df)*100
                               0.0
          age
Out[24]:
                               0.0
          sex
          steroid
         antivirals
fatigue
malaise
anorexia
liver_big
liver_firm
                              0.0
                            0.0
                             0.0
                             0.0
                             0.0
                             0.0
                              0.0
          spleen_palpable 0.0
          spiders
                              0.0
          ascites
                              0.0
          varices
                              0.0
          varices 0.0 bilirubin 0.0 alk_phosphate 0.0
                              0.0
          sgot
          albumin
                             0.0
```

dtype: float64

0.0

0.0

0.0

Interpolasi

protime histology

class

Solusi lain untuk mengganti nilai yang hilang melibatkan penggunaan fungsi lain, seperti interpolasi linier. Dalam kasus ini, misalnya, kita bisa mengganti nilai yang hilang di atas kolom, dengan interpolasi antara yang sebelumnya dan yang berikutnya. Ini dapat dicapai melalui penggunaan fungsi interpolate ().

```
In [25]: df = pd.read_csv('hepatitis_csv.csv')
```

```
df.isna().sum()/len(df)*100
          age
                                0.000000
Out[25]:
          sex
                                0.000000
          steroid
                              0.645161
          antivirals
                              0.000000
          fatigue
                              0.645161
                               0.645161
          malaise
                              0.645161
          anorexia
          anorexia
liver_big 6.451013
liver_firm 7.096774
spleen_palpable 3.225806
3.225806
                              3.225806
          ascites
          varices 3.225806
bilirubin 3.870968
alk_phosphate 18.709677
          sgot
                               2.580645
                             10.322581
          albumin
                              43.225806
          protime
          histology
                               0.000000
                               0.000000
          class
          dtype: float64
          Pilih kolom yang bertipe numerik
In [26]:
           numeric = df.select_dtypes(include=np.number)
           numeric_columns = numeric.columns
           df.head(10)
```

5]:	age	sex	steroid	antivirals	fatigue	malaise	anorexia	liver_big	liver_firm	spleen_palpable
0	30	male	False	False	False	False	False	False	False	False
1	50	female	False	False	True	False	False	False	False	False
2	78	female	True	False	True	False	False	True	False	False
3	31	female	NaN	True	False	False	False	True	False	False
4	34	female	True	False	False	False	False	True	False	False
5	34	female	True	False	False	False	False	True	False	False
6	51	female	False	False	True	False	True	True	False	True
7	23	female	True	False	False	False	False	True	False	False
8	39	female	True	False	True	False	False	True	True	False
9	30	female	True	False	False	False	False	True	False	False
4										>

Sekarang kita dapat menerapkan fungsi interpolate () ke kolom numerik, dengan mengatur juga arah batas ke depan. Artinya interpolasi linier diterapkan mulai dari baris pertama hingga baris terakhir.

```
In [27]: df[numeric_columns] = df[numeric_columns].interpolate(method ='linear', limit_direct
In [28]: df.head(10)
```

Out[28]: sex steroid antivirals fatigue malaise anorexia liver_big liver_firm spleen_palpable age

30 male **False False** False **False** False False False False 1 50 False female False True False False False False False 2 78 female True False False False True False False True 3 31 female NaN True **False** False False True False False 4 34 female True False **False** False **False** True **False** False 5 34 female True False False False False True False False 6 51 female **False False** True False True True **False** True 7 23 female True False **False** False False True **False** False 8 39 female True **False** True **False False** True True False

False

False

True

False

False

False

False

In [29]:

9

30 female

df.isna().sum()/len(df)*100

True

Out[29]:

0.000000 0.000000 steroid 0.645161 antivirals 0.000000 fatigue 0.645161 malaise 0.645161 anorexia 0.645161 6.451613 liver_big liver_firm 7.096774 spleen_palpable 3.225806 spiders 3.225806 ascites 3.225806 varices 3.225806 bilirubin 0.000000 alk_phosphate 0.000000 sgot 0.000000 albumin 0.000000 protime 1.935484 histology 0.000000 0.000000

dtype: float64

Excerise

class

Life Expectancy (WHO)

https://www.kaggle.com/datasets/augustus0498/life-expectancy-who

In [30]: import pandas as pd df = pd.read_csv('Life Expectancy Data.csv') df

Out[30]:

	Country	Year	Status	Life expectancy	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B
0	Afghanistan	2015	Developing	65.0	263.0	62	0.01	71.279624	65.0

	Country	Year	Status	Life expectancy	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B
1	Afghanistan	2014	Developing	59.9	271.0	64	0.01	73.523582	62.0
2	Afghanistan	2013	Developing	59.9	268.0	66	0.01	73.219243	64.0
3	Afghanistan	2012	Developing	59.5	272.0	69	0.01	78.184215	67.0
4	Afghanistan	2011	Developing	59.2	275.0	71	0.01	7.097109	68.0
•••									
2933	Zimbabwe	2004	Developing	44.3	723.0	27	4.36	0.000000	68.0
2934	Zimbabwe	2003	Developing	44.5	715.0	26	4.06	0.000000	7.0
2935	Zimbabwe	2002	Developing	44.8	73.0	25	4.43	0.000000	73.0
2936	Zimbabwe	2001	Developing	45.3	686.0	25	1.72	0.000000	76.0
2937	Zimbabwe	2000	Developing	46.0	665.0	24	1.68	0.000000	79.0

2938 rows × 22 columns

In [31]: print(

```
print(df.info())
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2938 entries, 0 to 2937
Data columns (total 22 columns):

#	Column	Non-Null Count	Dtype
0	Country	2938 non-null	object
1	Year	2938 non-null	int64
2	Status	2938 non-null	object
3	Life expectancy	2928 non-null	float64
4	Adult Mortality	2928 non-null	float64
5	infant deaths	2938 non-null	int64
6	Alcohol	2744 non-null	float64
7	percentage expenditure	2938 non-null	float64
8	Hepatitis B	2385 non-null	float64
9	Measles	2938 non-null	int64
10	BMI	2904 non-null	float64
11	under-five deaths	2938 non-null	int64
12	Polio	2919 non-null	float64
13	Total expenditure	2712 non-null	float64
14	Diphtheria	2919 non-null	float64
15	HIV/AIDS	2938 non-null	float64
16	GDP	2490 non-null	float64
17	Population	2286 non-null	float64
18	thinness 1-19 years	2904 non-null	float64
19	thinness 5-9 years	2904 non-null	float64
20	Income composition of resources	2771 non-null	float64
21	Schooling	2775 non-null	float64
dtvn	es: float64(16), int64(4), object	(2)	

dtypes: float64(16), int64(4), object(2)

memory usage: 505.1+ KB

None

In [32]:

df.describe()

	Year	Life expectancy	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B
count	2938.000000	2928.000000	2928.000000	2938.000000	2744.000000	2938.000000	2385.000000
mean	2007.518720	69.224932	164.796448	30.303948	4.602861	738.251295	80.940461
std	4.613841	9.523867	124.292079	117.926501	4.052413	1987.914858	25.070016
min	2000.000000	36.300000	1.000000	0.000000	0.010000	0.000000	1.000000
25%	2004.000000	63.100000	74.000000	0.000000	0.877500	4.685343	77.000000
50%	2008.000000	72.100000	144.000000	3.000000	3.755000	64.912906	92.000000
75 %	2012.000000	75.700000	228.000000	22.000000	7.702500	441.534144	97.000000
max	2015.000000	89.000000	723.000000	1800.000000	17.870000	19479.911610	99.000000
4							>

Data pre-Processing

Identifikasi Missing Value

Alcohol

```
In [33]:
          df.isna().sum()
                                                0
         Country
Out[33]:
          Year
                                                0
          Status
                                                0
          Life expectancy
                                               10
         Adult Mortality
                                               10
          infant deaths
                                                0
         Alcohol
                                              194
          percentage expenditure
                                              553
         Hepatitis B
         Measles
                                                0
          BMI
                                               34
          under-five deaths
                                                0
         Polio
                                               19
         Total expenditure
                                              226
         Diphtheria
                                               19
          HIV/AIDS
                                                0
         GDP
                                              448
                                              652
         Population
          thinness 1-19 years
                                               34
                                               34
          thinness 5-9 years
          Income composition of resources
                                              167
          Schooling
                                              163
          dtype: int64
         Cetak dalam bentuk presentase
In [34]:
          df.isna().sum()/len(df)*100
         Country
                                               0.000000
Out[34]:
          Year
                                               0.000000
          Status
                                               0.000000
          Life expectancy
                                               0.340368
         Adult Mortality
                                               0.340368
          infant deaths
                                               0.000000
```

6.603131

```
percentage expenditure
                                    0.000000
Hepatitis B
                                   18.822328
Measles
                                    0.000000
BMI
                                    1.157250
under-five deaths
                                    0.000000
Polio
                                    0.646698
Total expenditure
                                    7.692308
Diphtheria
                                    0.646698
HIV/AIDS
                                    0.000000
GDP
                                   15.248468
Population
                                   22.191967
thinness 1-19 years
                                    1.157250
thinness 5-9 years
                                    1.157250
Income composition of resources
                                   5.684139
Schooling
                                    5.547992
dtype: float64
```

Drop Missing Values

Drop baris dari spesifik kolom

```
In [35]:
    df.dropna(subset=['Total expenditure'],axis=0,inplace=True)
    df.isna().sum()/len(df)*100
```

0.000000 Country Out[35]: Year 0.000000 0.000000 Status Life expectancy 0.368732 Adult Mortality 0.368732 infant deaths 0.000000 Alcohol 0.184366 percentage expenditure 0.000000 19.026549 Hepatitis B Measles 0.000000 BMI 0.737463 under-five deaths 0.000000 Polio 0.294985 Total expenditure 0.000000 Diphtheria 0.294985 HIV/AIDS 0.000000 **GDP** 14.011799 Population 21.423304 thinness 1-19 years 0.737463 thinness 5-9 years 0.737463 Income composition of resources 4.682891 Schooling 4.535398

dtype: float64

In [36]: df.head(10)

Out[36]:

	Country	Year	Status	Life expectancy	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B	N
0	Afghanistan	2015	Developing	65.0	263.0	62	0.01	71.279624	65.0	
1	Afghanistan	2014	Developing	59.9	271.0	64	0.01	73.523582	62.0	
2	Afghanistan	2013	Developing	59.9	268.0	66	0.01	73.219243	64.0	
3	Afghanistan	2012	Developing	59.5	272.0	69	0.01	78.184215	67.0	
4	Afghanistan	2011	Developing	59.2	275.0	71	0.01	7.097109	68.0	

	Country	Year	Status	Life expectancy	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B	N
5	Afghanistan	2010	Developing	58.8	279.0	74	0.01	79.679367	66.0	
6	Afghanistan	2009	Developing	58.6	281.0	77	0.01	56.762217	63.0	
7	Afghanistan	2008	Developing	58.1	287.0	80	0.03	25.873925	64.0	
8	Afghanistan	2007	Developing	57.5	295.0	82	0.02	10.910156	63.0	
9	Afghanistan	2006	Developing	57.3	295.0	84	0.03	17.171518	64.0	

10 rows × 22 columns

Mengganti data Missing Values

```
In [37]:
          df.dtypes
                                               object
         Country
Out[37]:
                                                int64
         Year
         Status
                                               object
         Life expectancy
                                              float64
                                              float64
         Adult Mortality
         infant deaths
                                                int64
         Alcohol
                                              float64
         percentage expenditure
                                              float64
         Hepatitis B
                                              float64
         Measles
                                                int64
          BMI
                                              float64
         under-five deaths
                                                int64
                                              float64
         Polio
         Total expenditure
                                              float64
         Diphtheria
                                              float64
          HIV/AIDS
                                              float64
         GDP
                                              float64
         Population
                                              float64
          thinness 1-19 years
                                              float64
          thinness 5-9 years
                                              float64
         Income composition of resources
                                              float64
                                              float64
         Schooling
         dtype: object
```

Kolom Numerik

```
In [38]:
          import numpy as np
          numeric = df.select dtypes(include=np.number)
          numeric_columns = numeric.columns
```

```
In [39]:
          df[numeric_columns] = df[numeric_columns].fillna(df.mean())
```

C:\Users\USER\AppData\Local\Temp/ipykernel_10040/3464485706.py:1: FutureWarning: Dro pping of nuisance columns in DataFrame reductions (with 'numeric only=None') is depr ecated; in a future version this will raise TypeError. Select only valid columns be fore calling the reduction.

df[numeric_columns] = df[numeric_columns].fillna(df.mean())

Melihat hasilnya adalah

```
In [40]: df.isna().sum()/len(df)*100
                                             0.0
         Country
Out[40]:
         Year
                                             0.0
         Status
                                             0.0
         Life expectancy
                                             0.0
         Adult Mortality
                                             0.0
         infant deaths
                                             0.0
         Alcohol
                                             0.0
                                             0.0
         percentage expenditure
         Hepatitis B
                                             0.0
         Measles
                                             0.0
          BMT
                                             0.0
         under-five deaths
                                             0.0
         Polio
                                             0.0
                                             0.0
         Total expenditure
         Diphtheria
                                             0.0
          HIV/AIDS
                                             0.0
         GDP
                                             0.0
         Population
                                             0.0
          thinness 1-19 years
                                             0.0
          thinness 5-9 years
                                             0.0
         Income composition of resources
                                             0.0
         Schooling
                                             0.0
         dtype: float64
         Kolom Kategorial
In [41]:
          boolean_columns = df.select_dtypes(include=np.object).columns.tolist()
          boolean_columns.remove('Status')
          df[boolean_columns] = df[boolean_columns].astype('bool')
         C:\Users\USER\AppData\Local\Temp/ipykernel_10040/1529296272.py:1: DeprecationWarnin
         g: `np.object` is a deprecated alias for the builtin `object`. To silence this warni
         ng, use `object` by itself. Doing this will not modify any behavior and is safe.
         Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/r
         elease/1.20.0-notes.html#deprecations
           boolean_columns = df.select_dtypes(include=np.object).columns.tolist()
In [42]:
```

df[boolean_columns].fillna(df.mode())

```
Out[42]:
                  Country
               0
                      True
               1
                      True
               2
                      True
                      True
               4
                      True
            2933
                      True
            2934
                      True
            2935
                      True
            2936
                      True
            2937
                      True
```

Melihat hasilnya

```
In [43]:
          df.isna().sum()/len(df)*100
                                            0.0
         Country
Out[43]:
         Year
                                            0.0
                                            0.0
         Status
         Life expectancy
                                            0.0
         Adult Mortality
                                            0.0
         infant deaths
                                            0.0
         Alcohol
                                            0.0
         percentage expenditure
                                            0.0
         Hepatitis B
                                            0.0
         Measles
                                            0.0
          BMI
                                            0.0
         under-five deaths
                                            0.0
         Polio
                                            0.0
         Total expenditure
                                            0.0
         Diphtheria
                                            0.0
         HIV/AIDS
                                            0.0
         GDP
                                            0.0
         Population
                                            0.0
          thinness 1-19 years
                                            0.0
          thinness 5-9 years
                                            0.0
         Income composition of resources
                                            0.0
         Schooling
                                            0.0
         dtype: float64
```