

```
import pandas as pd

df=pd.read_csv("data.csv")
```

```
df.dtypes

Age          int64
BMI          float64
Glucose      int64
Insulin      float64
HOMA         float64
Leptin       float64
Adiponectin  float64
Resistin     float64
MCP.1        float64
Classification object
dtype: object
```

df.describe()

	Age	BMI	Glucose	Insulin	HOMA	Leptin	Adiponectin	Resistin	MCP.1
count	116.000000	116.000000	116.000000	116.000000	116.000000	116.000000	116.000000	116.000000	116.000000
mean	57.301724	27.582111	97.793103	10.012086	2.694988	26.615080	10.180874	14.725966	534.647000
std	16.112766	5.020136	22.525162	10.067768	3.642043	19.183294	6.843341	12.390646	345.912663
min	24.000000	18.370000	60.000000	2.432000	0.467409	4.311000	1.656020	3.210000	45.843000
25%	45.000000	22.973205	85.750000	4.359250	0.917966	12.313675	5.474283	6.881763	269.978250
50%	56.000000	27.662416	92.000000	5.924500	1.380939	20.271000	8.352692	10.827740	471.322500
75%	71.000000	31.241442	102.000000	11.189250	2.857787	37.378300	11.815970	17.755207	700.085000
max	89.000000	38.578759	201.000000	58.460000	25.050342	90.280000	38.040000	82.100000	1698.440000

```
df.groupby(by=['Age']).size()
```

```
Age
24    1
25    1
28    1
29    2
32    1
34    3
35    2
36    2
38    2
40    2
41    1
42    2
43    3
44    4
45    7
46    3
47    1
48    4
49    5
50    1
51    4
52    1
53    1
54    3
55    1
57    1
58    1
59    2
60    2
61    1
62    2
64    2
65    3
66    5
67    1
68    3
69    5
71    3
72    3
73    2
74    1
75    4
76    4
77    2
```

```
78    1
81    1
82    2
83    1
85    2
86    3
89    1
dtype: int64
```

```
df.isna().sum()
```

```
Age      0
BMI      0
Glucose  0
Insulin  0
HOMA     0
Leptin   0
Adiponectin  0
Resistin 0
MCP.1    0
Classification 0
dtype: int64
```

```
df.isnull()
```

```
Age      BMI  Glucose  Insulin  HOMA  Leptin  Adiponectin  Resistin  MCP.1  Classification
0  False  False    False    False  False  False      False    False  False          False
1  False  False    False    False  False  False      False    False  False          False
2  False  False    False    False  False  False      False    False  False          False
3  False  False    False    False  False  False      False    False  False          False
4  False  False    False    False  False  False      False    False  False          False
...     ...     ...      ...      ...     ...      ...      ...      ...      ...
111 False  False    False    False  False  False      False    False  False          False
112 False  False    False    False  False  False      False    False  False          False
113 False  False    False    False  False  False      False    False  False          False
114 False  False    False    False  False  False      False    False  False          False
115 False  False    False    False  False  False      False    False  False          False
```

116 rows × 10 columns

```
type("Age")
```

```
str
```

```
type("BMI")
```

```
str
```

```
df.Age.astype(float)
```

```
0      48.0
1      83.0
2      82.0
3      68.0
4      86.0
...
111    45.0
112    62.0
113    65.0
114    72.0
115    86.0
Name: Age, Length: 116, dtype: float64
```

```
df.Age=df.Age.astype(float)
```

```
df.describe()
```

```
df.dtypes
```

```
Age      float64
BMI      float64
```

```

Glucose      int64
Insulin      float64
HOMA         float64
Leptin       float64
Adiponectin  float64
Resistin     float64
MCP.1        float64
Classification object
dtype: object

```

```
df.sort_values('Age')
```



```
df.sort_values('Age',ascending=False)
```



```
df.rename(columns={'Age':'Year Old'})
```



```
df.sort_index()
```



```
df.reset_index()
```



```
df.drop(columns=['Age'])
```



```
pd.melt(df)
```



```
df.drop_duplicates()
```



```
df.head(5)
```



```
df.tail(5)
```



```
df.count()
```



```

Age          116
BMI          116
Glucose      116
Insulin      116
HOMA         116
Leptin       116
Adiponectin  116
Resistin     116
MCP.1        116
Classification 116
dtype: int64

```

```

for col in df.columns:
    print(f"Unique values in {col}: {df[col].unique()}")

```

```

# Convert non-numeric values to NaN for the 'Classification' column
df['Classification'] = pd.to_numeric(df['Classification'], errors='coerce')

```

```

# Calculate the median for numeric columns only
df.median()

```



```

Unique values in Age: [48. 83. 82. 68. 86. 49. 89. 76. 73. 75. 34. 29. 25. 24. 38. 44. 47. 61.
 64. 32. 36. 35. 54. 45. 50. 66. 53. 28. 43. 51. 67. 69. 60. 77. 71. 78.
 85. 42. 62. 59. 46. 72. 55. 41. 81. 65. 58. 40. 52. 74. 57.]
Unique values in BMI: [23.5      20.69049454 23.12467037 21.36752137 21.11111111 22.85445769
 22.7      23.8      22.      23.      21.47      23.01
 22.86      18.67      23.34      20.76      22.03      32.03895937]

```



```

34.5297228 36.51263743 28.57667585 31.97501487 32.27078777 30.27681661
30.48315806 37.03560819 38.57875854 31.44654088 35.2507611 34.17489
36.21227888 36.7901662 35.85581466 34.42217362 27.68877813 29.60676726
31.2385898 35.09270153 26.34929208 35.58792924 29.2184076 27.2
27.3 32.5 30.3 27.7 25.7 25.3
29.4 26.6 27.1 25.9 21.30394858 20.82999519
20.9566075 24.24242424 21.35991456 21.08281329 19.13265306 22.65625
22.4996371 21.51385851 22.89281998 22.83287935 23.14049587 24.21875
22.22222222 20.83 19.56 20.26 24.74 18.37
23.62 22.21 26.5625 31.25 26.66666667 26.6727633
28.67262608 31.64036818 32.46191136 25.51020408 29.296875 29.666548
28.125 29.15451895 30.83653053 31.21748179 30.8012487 31.23140988
29.77777778 27.88761707 27.63605442 27.91551882 28.44444444 28.65013774
30.91557669 29.13631634 34.83814777 37.109375 29.38475666 33.18
35.56 30.48 36.05 26.85 26.84 32.05
25.59 27.18 ]
Unique values in Glucose: [ 70 92 91 77 118 97 83 78 82 88 75 86 84 85 95 87 90 106
80 101 89 79 103 76 94 93 102 60 96 110 74 112 98 116 114 105
201 100 99 196 199 139 128 134 131 104 108 152 119 138]
Unique values in Insulin: [ 2.707 3.115 4.498 3.226 3.549 4.69 6.47 3.35 4.952 3.469
5.663 4.09 6.107 5.782 7.553 2.869 18.077 4.427 14.026 4.345
4.53 5.81 4.376 5.537 6.76 6.703 9.245 6.817 6.59 15.533
10.175 8.576 23.194 3.855 5.819 4.181 5.646 5.138 3.881 5.376
14.07 5.197 5.43 8.34 6.042 8.079 3.508 10.704 4.462 26.211
4.58 13.852 4.56 12.305 21.699 2.999 6.2 4.364 3.482 5.261
6.683 2.64 2.74 6.862 4.902 3.73 5.7 3.42 15.89 3.44
58.46 6.03 4.42 36.94 10.555 16.635 4.328 41.611 22.033 3.188
9.669 28.677 10.395 4.172 14.649 2.54 51.814 12.162 16.582 41.894
30.212 24.887 30.13 8.396 9.208 2.432 18.2 8.808 3.012 6.524
10.491 10.949 12.548 5.636 4.713 5.75 8.15 7.01 11.91 3.33
5.73 2.82 19.91 ]
Unique values in HOMA: [ 0.46740867 0.70689733 1.00965107 0.61272493 0.8053864 0.73208693
0.89078733 1.88320133 0.80154333 1.01383947 0.6674356 1.14543613
0.82727067 1.33 1.06967 1.6 0.59 3.79014433
1.03739367 3.0099796 0.92171933 0.972138 1.203832 0.9067072
1.229214 1.38399733 1.75261107 2.05239 1.513374 1.30042667
3.86978807 2.53493167 1.8404096 5.09185613 0.732193 1.13392913
0.84567693 1.4066068 1.30539453 0.72755813 1.1006464 3.262364
1.08963767 1.245642 2.098344 1.341324 1.8732508 0.519184
2.3498848 1.0566016 7.111918 0.96027333 3.4851632 0.832352
2.85311933 4.9242264 0.6879706 1.55992 1.0011016 0.79018187
1.23282767 1.84629013 0.507936 0.69614267 1.65877413 1.4026256
0.79125733 1.37788 0.742368 4.468268 0.78065067 15.28534133
1.56177 1.14478 7.83620533 2.62960233 3.775036 1.09960053
20.6307338 5.27176247 0.60550747 2.38502 7.0029234 2.871792
1.00851147 3.071407 0.56388 25.05034187 5.9699204 5.68541507
13.22733227 4.45899333 6.4834952 8.22598307 9.73600733 1.44970933
2.2485936 0.61789013 4.66890667 2.3464512 0.6538048 1.43223547
2.5101466 2.24162527 2.94041467 1.86288587 1.046286 1.30486667
2.63353667 2.62828267 3.495982 0.755688 1.1174 1.370998
0.570392 6.777364 ]
Unique values in Lentin: [ 8.8071 8.8438 17.9393 9.8827 6.6994 6.8317 6.964 4.311 4.47

```

df.quantile

```

<bound method DataFrame.quantile of
0 48.0 23.500000 70 2.707 0.467409 8.8071 9.702400
1 83.0 20.690495 92 3.115 0.706897 8.8438 5.429285
2 82.0 23.124670 91 4.498 1.009651 17.9393 22.432040
3 68.0 21.367521 77 3.226 0.612725 9.8827 7.169560
4 86.0 21.111111 92 3.549 0.805386 6.6994 4.819240
.. ... ..
111 45.0 26.850000 92 3.330 0.755688 54.6800 12.100000
112 62.0 26.840000 100 4.530 1.117400 12.4500 21.420000
113 65.0 32.050000 97 5.730 1.370998 61.4800 22.540000
114 72.0 25.590000 82 2.820 0.570392 24.9600 33.750000
115 86.0 27.180000 138 19.910 6.777364 90.2800 14.110000

```

```

Resistin MCP.1 Classification
0 7.99585 417.114 1.0
1 4.06405 468.786 1.0
2 9.27715 554.697 1.0
3 12.76600 928.220 1.0
4 10.57635 773.920 1.0
.. ... ..
111 10.96000 268.230 2.0
112 7.32000 330.160 2.0
113 10.33000 314.050 2.0
114 3.27000 392.460 2.0
115 4.35000 90.090 2.0

```

[116 rows x 10 columns]>

df.min()

```

Age 24.000000
BMI 18.370000
Glucose 60.000000
Insulin 2.432000

```

```
HOMA          0.467409
Leptin         4.311000
Adiponectin    1.656020
Resistin       3.210000
MCP.1          45.843000
Classification  1.000000
dtype: float64
```

```
df.max()
```

```
➡ Age          89.000000
   BMI         38.578759
   Glucose     201.000000
   Insulin      58.460000
   HOMA         25.050342
   Leptin       90.280000
   Adiponectin  38.040000
   Resistin     82.100000
   MCP.1       1698.440000
   Classification 2.000000
dtype: float64
```

```
for col in df.columns:
    print(f"Unique values in {col}: {df[col].unique()}")
```

```
df.mean()
```

```
➡ Unique values in Age: [48. 83. 82. 68. 86. 49. 89. 76. 73. 75. 34. 29. 25. 24. 38. 44. 47. 61.
 64. 32. 36. 35. 54. 45. 50. 66. 53. 28. 43. 51. 67. 69. 60. 77. 71. 78.
 85. 42. 62. 59. 46. 72. 55. 41. 81. 65. 58. 40. 52. 74. 57.]
Unique values in BMI: [23.5      20.69049454 23.12467037 21.36752137 21.11111111 22.85445769
 22.7      23.8      22.      23.      21.47      23.01
 22.86     18.67     23.34     20.76     22.03     32.03895937
 34.5297228 36.51263743 28.57667585 31.97501487 32.27078777 30.27681661
 30.48315806 37.03560819 38.57875854 31.44654088 35.2507611 34.17489
 36.21227888 36.7901662 35.85581466 34.42217362 27.68877813 29.60676726
 31.2385898 35.09270153 26.34929208 35.58792924 29.2184076 27.2
 27.3      32.5      30.3      27.7      25.7      25.3
 29.4      26.6      27.1      25.9      21.30394858 20.82999519
 20.9566075 24.24242424 21.35991456 21.08281329 19.13265306 22.65625
 22.49963271 21.51385851 22.89281998 22.83287935 23.14049587 24.21875
 22.22222222 20.83     19.56     20.26     24.74     18.37
 23.62     22.21     26.5625     31.25     26.66666667 26.6727633
 28.67262608 31.64036818 32.46191136 25.51020408 29.296875 29.666548
 28.125     29.15451895 30.83653053 31.21748179 30.8012487 31.23140988
 29.77777778 27.88761707 27.63605442 27.91551882 28.44444444 28.65013774
 30.91557669 29.13631634 34.83814777 37.109375 29.38475666 33.18
 35.56     30.48     36.05     26.85     26.84     32.05
 25.59     27.18     ]
Unique values in Glucose: [ 70 92 91 77 118 97 83 78 82 88 75 86 84 85 95 87 90 106
 80 101 89 79 103 76 94 93 102 60 96 110 74 112 98 116 114 105
 201 100 99 196 199 139 128 134 131 104 108 152 119 138]
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 5.663 4.09 6.107 5.782 7.553 2.869 18.077 4.427 14.026 4.345
 4.53 5.81 4.376 5.537 6.76 6.703 9.245 6.817 6.59 15.533
 10.175 8.576 23.194 3.855 5.819 4.181 5.646 5.138 3.881 5.376
 14.07 5.197 5.43 8.34 6.042 8.079 3.508 10.704 4.462 26.211
 4.58 13.852 4.56 12.305 21.699 2.999 6.2 4.364 3.482 5.261
 6.683 2.64 2.74 6.862 4.902 3.73 5.7 3.42 15.89 3.44
 58.46 6.03 4.42 36.94 10.555 16.635 4.328 41.611 22.033 3.188
 9.669 28.677 10.395 4.172 14.649 2.54 51.814 12.162 16.582 41.894
 30.212 24.887 30.13 8.396 9.208 2.432 18.2 8.808 3.012 6.524
 10.491 10.949 12.548 5.636 4.713 5.75 8.15 7.01 11.91 3.33
 5.73 2.82 19.91 ]
Unique values in HOMA: [ 0.46740867 0.70689733 1.00965107 0.61272493 0.8053864 0.73208693
 0.89078733 1.88320133 0.80154333 1.01383947 0.6674356 1.14543613
 0.82727067 1.33 1.06967 1.6 0.59 3.79014433
 1.03739367 3.0099796 0.92171933 0.972138 1.203832 0.9067072
 1.229214 1.38399733 1.75261107 2.05239 1.513374 1.30042667
 3.86978807 2.53493167 1.8404096 5.09185613 0.732193 1.13392913
 0.84567693 1.4066068 1.30539453 0.72755813 1.1006464 3.262364
 1.08963767 1.245642 2.098344 1.341324 1.8732508 0.519184
 2.3498848 1.0566016 7.111918 0.96027333 3.4851632 0.832352
 2.85311933 4.9242264 0.6879706 1.55992 1.0011016 0.79018187
 1.23282767 1.84629013 0.507936 0.69614267 1.65877413 1.4026256
 0.79125733 1.37788 0.742368 4.468268 0.78065067 15.28534133
 1.56177 1.14478 7.83620533 2.62960233 3.775036 1.09960053
 20.6307338 5.27176247 0.60550747 2.38502 7.0029234 2.871792
 1.00851147 3.071407 0.56388 25.05034187 5.9699204 5.68541507
 13.22733227 4.45899333 6.4834952 8.22598307 9.73600733 1.44970933
 2.2485936 0.61789013 4.66890667 2.3464512 0.6538048 1.43223547
 2.5101466 2.24162527 2.94041467 1.86288587 1.046286 1.30486667
 2.63353667 2.62828267 3.495982 0.755688 1.1174 1.370998
 0.570392 6.777364 ]
Unique values in Leptin: [ 8.8071 8.8438 17.9393 9.8827 6.6994 6.8317 6.964 4.311 4.47
```

```
# Convert all numeric columns
df = df.apply(pd.to_numeric, errors='coerce')
df.std()
```

```
Age          16.112766
BMI           5.020136
Glucose       22.525162
Insulin       10.067768
HOMA          3.642043
Leptin        19.183294
Adiponectin   6.843341
Resistin      12.390646
MCP.1         345.912663
Classification 0.498406
dtype: float64
```

```
pd.get_dummies(df['Age'])
```

```
pd.get_dummies(df['Glucose'])
```

```
df.iloc[2:50]
```

```
print(pd.get_dummies(df['Glucose']))
```

```

60    70    74    75    76    77    78    79    80    82  \
0  False  True  False  False  False  False  False  False  False  False
1  False  False  False  False  False  False  False  False  False  False
2  False  False  False  False  False  False  False  False  False  False
3  False  False  False  False  False  True  False  False  False  False
4  False  False  False  False  False  False  False  False  False  False
..    ...    ...    ...    ...    ...    ...    ...    ...    ...
111  False  False  False  False  False  False  False  False  False  False
112  False  False  False  False  False  False  False  False  False  False
113  False  False  False  False  False  False  False  False  False  False
114  False  False  False  False  False  False  False  False  False  True
115  False  False  False  False  False  False  False  False  False  False

...    119    128    131    134    138    139    152    196    199    201
0  ...  False  False  False  False  False  False  False  False  False  False
1  ...  False  False  False  False  False  False  False  False  False  False
2  ...  False  False  False  False  False  False  False  False  False  False
3  ...  False  False  False  False  False  False  False  False  False  False
4  ...  False  False  False  False  False  False  False  False  False  False
..    ...    ...    ...    ...    ...    ...    ...    ...    ...    ...
111  ...  False  False  False  False  False  False  False  False  False  False
112  ...  False  False  False  False  False  False  False  False  False  False
113  ...  False  False  False  False  False  False  False  False  False  False
114  ...  False  False  False  False  False  False  False  False  False  False
115  ...  False  False  False  False  True  False  False  False  False  False

[116 rows x 50 columns]
```