

Python Pandas

Pandas

Python library used for working with datasets. It is a python package that provides fast, flexible, and expressive data structures designed to make working with relational/labeled data both easy and initiative.

The most powerful and flexible open source analyzing, cleaning, exploring and manipulating data.

- · Easy handling of missing data
- Size mutability(columns can be inserted and deleted)

```
import pandas as pd
pd.__version__
```

DataFrame

• 2D array or a table with rows and column

iloc (integer-location based indexing)

Access data by specifying the rows and column positions as integers.

loc(Label-based indexing)

Access data by specifying the row and column labels.

```
#iloc
print(myvar.iloc[0:2,1])
print('\n')
#loc
print(myvar.loc[:,['Bike Category']])
```

```
O GSA 1250
1 Speed_Twin
Name: Bike Names, dtype: object

Bike Category
O Adventure
1 Street Bikes
2 Touring
3 Sports
```

Pandas Series

one-dimensional, labeled array capable of holding data of any type (integer, string, float, Python objects, etc.). It can be thought of as a column in a spreadsheet or a single column of a Pandas DataFrame.

```
list_d=[1,2,3,4,5,6]
list_series=pd.Series(list_d)
print('List Series \n',list_series)
print('\n')
dict d=({1:'A',2:'B',3:'C'})
dict_series=pd.Series(dict_d)
print('Dictionary Series \n',dict_series)
print('\n')
#custom Index
data=['a','b','c','d','e']
scalar_series=pd.Series(data,index=[1,2,3,4,5])
print('Scalar Series \n',scalar_series)
print('\n')
→ List Series
     0 1
         2
     2
         3
     3
         4
         5
         6
     dtype: int64
     Dictionary Series
     1
         Α
     2
         В
         C
     dtype: object
     Scalar Series
     1
         b
     3
         С
     4
         d
     dtype: object
# Load files into a DataFrame
#read a csv file as a data frame df
df=pd.read_csv("/content/drive/MyDrive/FSDS @Kodi Senapati/Datasets/Pandas.csv")
df.head()
```



You can print any number of rows by mentioning the number
df.tail(10)



#The info() method prints information about the DataFrame.

#The information contains the number of columns, column labels, column data types, memory usage, range index, and the number of cells in eac

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32 entries, 0 to 31
Data columns (total 5 columns):
Column Non-Null Count Dtype

Data columns (cotal 5 columns).						
#	Column	Non-Null Count	Dtype			
0	Duration	32 non-null	int64			
1	Date	31 non-null	object			
2	Pulse	32 non-null	int64			
3	Maxpulse	32 non-null	int64			
4	Calories	30 non-null	float64			
<pre>dtypes: float64(1), int64(3), object(1)</pre>						
memory usage: 1.4+ KB						

#find if any null values in df
print(df.isnull())

$\overline{\pm}$		Duration	Date	Pulse	Maxpulse	Calories
	0	False	False	False	False	False
	1	False	False	False	False	False
	2	False	False	False	False	False
	3	False	False	False	False	False
	4	False	False	False	False	False
	5	False	False	False	False	False
	6	False	False	False	False	False
	7	False	False	False	False	False
	8	False	False	False	False	False
	9	False	False	False	False	False
	10	False	False	False	False	False
	11	False	False	False	False	False
	12	False	False	False	False	False
	13	False	False	False	False	False
	14	False	False	False	False	False
	15	False	False	False	False	False
	16	False	False	False	False	False
	17	False	False	False	False	False
	18	False	False	False	False	True
	19	False	False	False	False	False
	20	False	False	False	False	False
	21	False	False	False	False	False
	22	False	True	False	False	False
	23	False	False	False	False	False
	24	False	False	False	False	False
	25	False	False	False	False	False
	26	False	False	False	False	False
	27	False	False	False	False	False
	28	False	False	False	False	True
	29	False	False	False	False	False
	30	False	False	False	False	False
	31	False	False	False	False	False

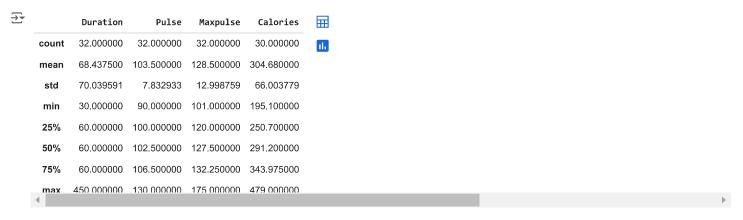
#By above one we cant find exact null value. #Summarize the count of null values column wise

print(df.isnull().sum())

Duration 0
Date 1
Pulse 0
Maxpulse 0
Calories 2
dtype: int64

df.describe()

#Gives overview of data of numbered columns



Slicing df[1:10:3]

1 to 10 with step 3

1 60 '2020/12/02' 117 145 479.0 4 45 '2020/12/05' 117 148 406.0 7 450 '2020/12/08' 104 134 253.3	₹		Duration	Date	Pulse	Maxpulse	Calories	
7 450 '2020/12/08' 104 134 253.3		1	60	'2020/12/02'	117	145	479.0	ıl.
		4	45	'2020/12/05'	117	148	406.0	
		7	450	'2020/12/08'	104	134	253.3	

Get column wise details

print(df[['Duration','Pulse','Maxpulse']])

\rightarrow		Duration	Pulse	Maxpulse
	0	60	110	130
	1	60	117	145
	2	60	103	135
	3	45	109	175
	4	45	117	148
	5	60	102	127
	6	60	110	136
	7	450	104	134
	8	30	109	1 33
	9	60	98	124
	10	60	103	147
	11	60	100	120
	12	60	100	120
	13	60	106	128
	14	60	104	132
	15	60	98	123
	16	60	98	120
	17	60	100	120
	18	45	90	112
	19	60	103	123
	20	45	97	125
	21	60	108	131
	22	45	100	119
	23	60	130	101
	24	45	105	132
	25	60	102	126
	26	60	100	120
	27	60	92	118
	28	60	103	132
	29	60	100	132
	30	60	102	129
	31	60	92	115

 $\ensuremath{\text{\#}}$ Find data types of each column

print(df.dtypes)

₹	Duration	int64
	Date	object
	Pulse	int64
	Maxpulse	int64

```
Calories
                float64
    dtype: object
# Find the count of columns
print(len(df.dtypes))
print('\n')
print(len(df.columns))
→ 5
#split the categorial and number columns
print(df.info())
df_int=df.select_dtypes(include=['int64'])
df_float=df.select_dtypes(include=['float64'])
df_object=df.select_dtypes(include=['object'])
print('\n')
print(len(df.columns))
print('\n')
print(len(df_int.columns))
print(len(df_float.columns))
print(len(df_object.columns))
<class 'pandas.core.frame.DataFrame'>
    RangeIndex: 32 entries, 0 to 31
    Data columns (total 5 columns):
     # Column Non-Null Count Dtype
     0 Duration 32 non-null
                                   int64
                   31 non-null
                                  object
     1 Date
        Pulse
                   32 non-null
                                   int64
     3 Maxpulse 32 non-null
                                   int64
     4 Calories 30 non-null
                                  float64
    dtypes: float64(1), int64(3), object(1)
    memory usage: 1.4+ KB
    None
    5
    3
    1
```

Data Cleaning

which means fixing bad data into clean data.

Bad data could be:

- Empty Cell
- Data in wrong format
- Wrong Data
- Duplicates

Cleaning Empty Cells

```
0
        Duration 32 non-null
                                   int64
         Date
                   31 non-null
                                  object
         Pulse
                   32 non-null
                                  int64
         Maxpulse 32 non-null
                                  int64
     4 Calories 30 non-null
                                  float64
    dtypes: float64(1), int64(3), object(1)
    memory usage: 1.4+ KB
df_new.isnull().sum()
₹
               0
      Duration 0
       Date
       Pulse
               0
     Maxpulse 0
      Calories
```

df.dropna()

• this method returns a new DataFrame, and will not change the original.

df.dropna(inplace=True)

• this method will NOT return a new DataFrame, but it will remove all rows containing NULL values from the original DataFrame.

df.fillna(100,inplace=True)

· this method allows us to replace empty cells with a value

df["Duration"].fillna(100,inplace=True)

• Replace NULL values in the "Duration" columns with the number 100.

df["Calories"].fillna(df["Calories"].mean(), inplace=True)

• Replace NULL values in the "Calories" column with the mean value of that column.

df["Calories"].fillna(df["Calories"].median(), inplace=True)

• Replace NULL values in the "Calories" column with the median value of that column.

df["Calories"].fillna(df["Calories"].mode(), inplace=True)

• Replace NULL values in the "Calories" column with the mode value of that column.

```
# Drop rows if any column value is NULL
df_c=df.dropna()
# The above one returns a new DataFrame without changing the original dataFrame
print('df_c:\n',df_c.isnull().sum())
print("\n")
print('df:\n',df.isnull().sum())
#df.dropna(inplace=True)
# The above one returns a new DataFrame changing the original dataFrame
→
     df_c:
                  0
      Duration
     Date
                 0
     Pulse
     Maxpulse
                 0
     Calories
     dtype: int64
     df:
      Duration
                  0
     Date
                 1
     Pulse
                 0
     Maxpulse
                 0
     Calories
```

```
dtype: int64
df_new["Calories"].fillna(df_new["Calories"].mean(), inplace=True)
    <ipython-input-85-33da02143b9d>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignm
     The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting value
     For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].me
       df_new["Calories"].fillna(df_new["Calories"].mean(), inplace=True)
print(df_new.isnull().sum())
→ Duration
     Date
     Pulse
                 0
     Maxpulse
                 0
     Calories
                 0
     dtype: int64
Convert into a Correct Format
# convert all cells in the 'Date' column into dates
df_new['Date']=pd.to_datetime(df_new['Date'],errors='coerce')
df_new['Date'].fillna(df['Date'].mean(),inplace=True)
print(df_new.isnull().sum())
Duration
                 0
     Date
                 0
     Pulse
                 0
     Maxpulse
     Calories
     dtype: int64
     <ipython-input-89-135ecb763a5e>:3: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignm
     The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting value
     For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].me
       df_new['Date'].fillna(df['Date'].mean(),inplace=True)
# Drop duplicate rows
df_new.drop_duplicates()
print(df_new.shape)
```

Finding Relationships

 \rightarrow (32, 5)

- The corr() method calculates the relationship between each column in your data set.
- · The corr() method ignores "not numeric" columns.
- Perfect Correlation: We can see that "Duration" and "Duration" got the number 1.000000, which makes sense, each column always has a perfect relationship with itself.
- Good Correlation: "Duration" and "Calories" got a 0.922721 correlation, which is a very good correlation, and we can predict that the longer you work out, the more calories you burn, and the other way around: if you burned a lot of calories, you probably had a long work out.
- Bad Correlation: "Duration" and "Maxpulse" got a 0.009403 correlation, which is a very bad correlation, meaning that we can not predict the max pulse by just looking at the duration of the work out, and vice versa.

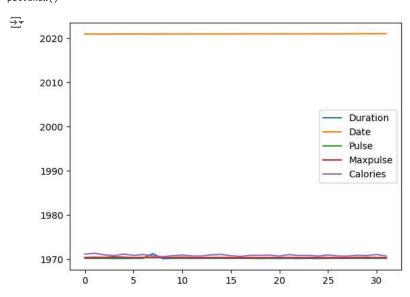
```
# Data Correlations
df new.corr()
```



Plotting

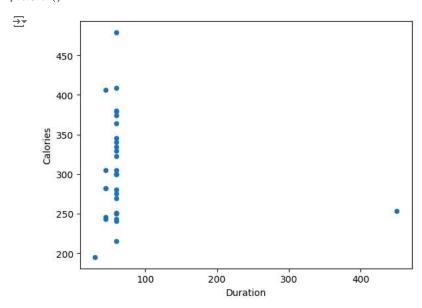
import matplotlib.pyplot as plt

df_new.plot()
plt.show()

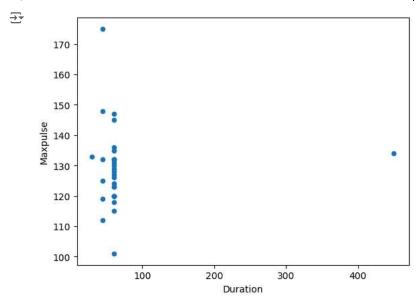


Scatter Plot

df_new.plot(kind='scatter',x='Duration',y='Calories')
plt.show()



df_new.plot(kind='scatter',x='Duration',y='Maxpulse')
plt.show()



Histogram

df['Calories'].plot(kind='hist')
plt.show()

