Double-click (or enter) to edit

Download the sampledata_orders.csv dataset. As the name says it is a .csv file. The file contains information regarding sales order details of United Shipping Agency. Open it in a text editor or excel and inspect it first. Write a python program to load the data stored in a csv data file sampledata_orders.csv in a pandas DataFrame and perform the following operations.

- i. Print the top 5 row and top 10 row
- ii. Print the all column names
- iii. Print the all row index ranges
- iv. Print the Order Quantity column (top 5 values)
- v. Print The Sales column (top 10 values)
- vi. Print the row with ID:50
- vii. Print the third row
- viii. Print the Order Quantity, Sales, Discount and Profit of the 2nd,4th, 6th and 8th row:
- ix. Print the Order Quantity, Sales, Discount and Profit of orders with discount > 10%
- x. Print the production cost.(production cost= Sales-Profit)

```
import pandas as pd
orders = pd.read_csv("/content/sampledata_orders.csv")
#print("First 5 entries:")
#print(orders.head())
#print(orders.head(10))
#print(orders.columns) #Print the all column names
#Print the all row index ranges
#print(orders.index)
#print(orders["Order Quantity"].head()) #Print the Order Quantity column (top 5 values)
#print(orders["Sales"].head(10))#Print The Sales column (top 10 values)
#print("The row with ID:50")
\#r50 = orders.loc[50]
                       #Print the row with ID:50
#print(r50)
#r3=orders.iloc[3]
#print(r3) #Print the third row
#print(orders[2:9:2][["Order Quantity", "Sales","Discount", "Profit"]])
#print(orders[orders["Discount"] > 0.1][["Order Quantity", "Sales", "Discount", "Profit"]])
pcost=orders["Sales"]-orders["Profit"]
print(pcost)
               474.7900
     1
              9665.2100
     2
               197.8600
     3
              3766.7895
               363.3300
     4
```

```
8394 1617.2200
8395 370.3200
8396 480.7300
8397 629.0775
8398 1445.5800
Length: 8399, dtype: float64
```

Download the diamonds.csv dataset. As the name says it is a .csv file. The file contains information regarding diamonds details of Laisha diamond shop. Open it in a text editor or excel and inspect it first. Write a python program to load the data stored in a csv file diamonds.csv in a pandas DataFrame and perform the following operations.

- I. find carat weight at least 0.3 and write this detail into cart.csv file
- II. find total price of cart weight at maximum 0.3
- III. Sort the 'carat' column by ascending and descending order.
- IV. Print the premium diamond details
- V. Find the most expensive diamond whose color is D

```
import pandas as pd
df = pd.read_csv('/content/diamonds.csv')
sample=df[df["carat"]<.3]
sample.to_csv('cart.csv')
tot=df[df["carat"]>.3]
tot["price"].sum()
result = df.sort_values('carat')
print(result)
result = df.sort_values('carat', ascending=False)
print(result)
pre=df.groupby("cut")
pre.get_group("Premium")
```

31593 31597 31596 31595 31594	carat 0.20 0.20 0.20 0.20 0.20	cut Premium Ideal Premium Ideal Premium	color E D F E	clarity VS2 VS2 VS2 VS2 VS2	depth 61.1 61.5 62.6 59.7 59.7	table 59.0 57.0 59.0 55.0 62.0	price 367 367 367 367 367	3.8 3.8 3.7 3.8	1 3 3 3 6 3	y 3.78 3.77 3.71 3.84 3.80	z 2.32 2.33 2.33 2.30 2.28
25999 25998 27130 27630 27415	4.01 4.01 4.13 4.50 5.01	Premium Premium Fair Fair Fair	T H I T	I1 I1 I1 I1 I1	62.5 61.0 64.8 65.8 65.5	62.0 61.0 61.0 58.0 59.0	15223 15223 17329 18531 18018	10.0 10.1 10.0 10.2 10.7	2 9 4 10 0 9 3 10	9.94 9.10 9.85 9.16 9.54	6.24 6.17 6.43 6.72 6.98
[53940 27415 27630 27130 25999 25998 31592 31591 31601 14 31596	rows x carat 5.01 4.50 4.13 4.01 4.01 0.20 0.20 0.20 0.20 0.20	10 colum cut Fair Fair Fair Premium	_	clarity I1 I1 I1 I1 VS2 VS2 VS2 VS2 VS2 VS2 VS2 VS2	depth 65.5 65.8 64.8 62.5 61.0 59.0 59.8 61.7 60.2 62.6	table 59.0 58.0 61.0 62.0 60.0 62.0 69.0 59.0	price 18018 18531 17329 15223 15223 367 367 367 345 367	10.7 10.2 10.0 10.0 10.1 3.8 3.7 3.7 3.7	3 10 0 9 2 9 4 10 	y 0.54 0.16 9.85 9.94 0.10 3.78 3.77 3.72 3.75 3.71	2 6.98 6.72 6.43 6.24 6.17 2.24 2.26 2.31 2.27 2.33
[53940		10 colum	_								
	carat	cut	color	clarity	depth	table	price	X	У	Z	_
1	0.21	Premium	Е	SI1	59.8	61.0	326	3.89	3.84	2.31	
3	0.29	Premium	- 1	VS2	62.4	58.0	334	4.20	4.23	2.63	
12	0.22	Premium	F	SI1	60.4	61.0	342	3.88	3.84	2.33	

1. WAP program to read a sampledata_orders.csv file

and perform descriptive statics and print the result.

```
import pandas as pd
orders = pd.read_csv("/content/sampledata_orders.csv")
print("A description of the numerical values:")
print(orders.describe())
orders.min()
orders.max()
orders.std()
orders.var()
orders.mean()
orders.median()
orders.mode()
orders.sum()
orders.count()
orders.rank()
```

```
A description of the numerical values:
                        Order ID Order Quantity
            Row ID
                                                          Sales
                                                                    Discount \
       8399.000000
                     8399.000000
                                      8399.000000
                                                    8399.000000
                                                                 8399.000000
count
                                                    1775.878179
       4200.000000
                    29965.179783
                                        25.571735
                                                                    0.049671
mean
std
       2424.726789
                    17260.883447
                                        14.481071
                                                    3585.050525
                                                                    0.031823
          1.000000
                        3.000000
                                        1.000000
                                                       2.240000
                                                                    0.000000
min
25%
       2100.500000
                    15011.500000
                                        13.000000
                                                     143.195000
                                                                    0.020000
                                        26.000000
50%
       4200.000000
                    29857.000000
                                                     449.420000
                                                                    0.050000
75%
       6299.500000 44596.000000
                                        38.000000
                                                    1709.320000
                                                                    0.080000
                                                                    0.250000
       8399.000000
                   59973.000000
                                        50.000000
                                                   89061.050000
max
             Profit
                                  Shipping Cost Product Base Margin
                      Unit Price
count
        8399.000000 8399.000000
                                    8399.000000
                                                          8336.000000
        181.184423
                       89.346259
                                       12.838557
                                                             0.512513
mean
std
        1196.653372
                      290.354383
                                       17.264052
                                                             0.135589
      -14140.700000
                                                             0.350000
min
                        0.990000
                                        0.490000
25%
         -83.315000
                        6.480000
                                        3.300000
                                                             0.380000
50%
          -1.500000
                       20.990000
                                        6.070000
                                                             0.520000
75%
         162.750000
                       85.990000
                                       13.990000
                                                             0.590000
       27220.690000 6783.020000
                                      164.730000
                                                             0.850000
max
<ipython-input-7-ec6ac894d235>:7: FutureWarning: Dropping of nuisance columns i
  orders.std()
<ipython-input-7-ec6ac894d235>:8: FutureWarning: Dropping of nuisance columns i
  orders.var()
<ipython-input-7-ec6ac894d235>:9: FutureWarning: Dropping of nuisance columns i
  orders.mean()
<ipython-input-7-ec6ac894d235>:10: FutureWarning: Dropping of nuisance columns
  orders.median()
<ipython-input-7-ec6ac894d235>:15: FutureWarning: Dropping of nuisance columns
  orders.skew()
<ipython-input-7-ec6ac894d235>:16: FutureWarning: Dropping of nuisance columns
  orders.kurt()
Row ID
                        -1.200000
Order ID
                        -1.178317
Order Quantity
                        -1.208020
                        60.928376
Sales
Discount
                        -0.959411
Profit
                        67.349705
Unit Price
                       271.168733
Shipping Cost
                         7.751587
Product Base Margin
                        -0.660870
dtype: float64
```

2. Write a python program to read a sampledata_orders.csv file perform descriptive statics for the first 25 rows of the sales column and print the result.

```
import pandas as pd
orders = pd.read_csv("sampledata_orders.csv")
```

```
orders[0:25:1][["Sales"]].min()
orders[0:25:1][["Sales"]].max()
orders[0:25:1][["Sales"]].std()
orders[0:25:1][["Sales"]].var()
orders[0:25:1][["Sales"]].mean()
orders[0:25:1][["Sales"]].mode()
orders[0:25:1][["Sales"]].sum()
orders[0:25:1][["Sales"]].count()
orders[0:25:1][["Sales"]].rank()
orders[0:25:1][["Sales"]].skew()
orders[0:25:1][["Sales"]].skew()
orders[0:25:1][["Sales"]].kurt()
```

Horizontal bar chart

```
import numpy as np
import pandas as pd
#!pip install matplotlib==3.4
from matplotlib import pyplot as plt
plt.style.use('seaborn-whitegrid')
import pandas as pd
import numpy as np
fig = plt.figure()
ax = plt.axes()
```



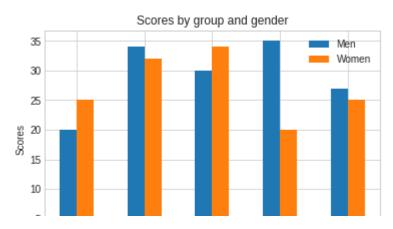
```
from matplotlib import pyplot as plt
plt.style.use('seaborn-whitegrid')
fig = plt.figure()
ax = plt.axes()
data = {'Apple': 10,
        'Orange': 15,
        'Banana': 13,
        'Pear': 9,
        'Pomegrante': 11,
        'Lemon': 5,
        'Lime': 20}
names = list(data.keys())
values = list(data.values())
ax = plt.axes()
ax.barh(names, values, color='c', height=0.80)
                                                         # Horizontal bar-chart
ax.set_xlabel('Quantity')
```

```
ax.set_ylabel('Fruit')
ax.set_title('Fruit Quantities');
```

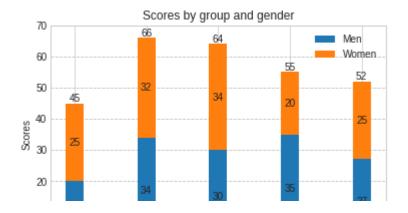
plt.show()



```
labels = ['G1', 'G2', 'G3', 'G4', 'G5']
men_means = [20, 34, 30, 35, 27]
women_means = [25, 32, 34, 20, 25]
x = np.arange(len(labels))  # the label locations
width = 0.25  # the width of the bars
fig, ax = plt.subplots()
rects1 = ax.bar(x - width/2, men_means, width, label='Men')
rects2 = ax.bar(x + width/2, women_means, width, label='Women')
ax.set_ylabel('Scores')
ax.set_title('Scores by group and gender')
ax.set_xticks(x)
ax.set_xticklabels('Groups')
ax.set_xticklabels(labels)
ax.legend();
```



```
ax.bar_label(p2, label_type='center')
ax.bar_label(p2)
ax.set_ylim(0,70)
```



flights Data set

import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
flights = pd.read_csv("/content/flights.csv")
flights.head()

	year	month	day	dep_time	dep_delay	arr_time	arr_delay	carrier	tailnum	1
0	2013	1	1	517.0	2.0	830.0	11.0	UA	N14228	
1	2013	1	1	533.0	4.0	850.0	20.0	UA	N24211	
2	2013	1	1	542.0	2.0	923.0	33.0	AA	N619AA	
3	2013	1	1	554.0	-6.0	812.0	-25.0	DL	N668DN	
4	2013	1	1	554.0	-4.0	740.0	12.0	UA	N39463	

flights.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 160754 entries, 0 to 160753
Data columns (total 16 columns):

#	Column	Non-Null Count	Dtype
0	year	160754 non-null	int64
1	month	160754 non-null	int64
2	day	160754 non-null	int64
3	dep_time	158418 non-null	float64
4	dep_delay	158418 non-null	float64
5	arr_time	158275 non-null	float64
6	arr_delay	157927 non-null	float64
7	carrier	160754 non-null	object
8	tailnum	159321 non-null	object
9	flight	160754 non-null	int64

	origin	160754	non-null	object				
11	dest	160754	non-null	object				
12	air_time	157927	non-null	float64				
13	distance	160754	non-null	int64				
14	hour	158418	non-null	float64				
15	minute	158418	non-null	float64				
<pre>dtypes: float64(7), int64(5), object(4)</pre>								
memo	ry usage: 1	9.6+ MB						

Select the rows that have at least one missing value flights[flights.isnull().any(axis=1)]

	year	month	day	dep_time	dep_delay	arr_time	arr_delay	carrier	tailr
330	2013	1	1	1807.0	29.0	2251.0	NaN	UA	N314
403	2013	1	1	NaN	NaN	NaN	NaN	AA	N3EH
404	2013	1	1	NaN	NaN	NaN	NaN	AA	N3E\
855	2013	1	2	2145.0	16.0	NaN	NaN	UA	N122
858	2013	1	2	NaN	NaN	NaN	NaN	AA	٨
			•••				•••	•••	
159681	2013	9	28	1214.0	-11.0	1801.0	NaN	AA	N488
159854	2013	9	28	NaN	NaN	NaN	NaN	AA	N320
159855	2013	9	28	NaN	NaN	NaN	NaN	US	٨
160185	2013	9	29	1734.0	23.0	2159.0	NaN	UA	N463
160286	2013	9	29	NaN	NaN	NaN	NaN	UA	Ν

2827 rows × 16 columns

2. Filter all the rows where arr_delay value is missing:

flights1 = flights[flights['arr_delay'].notnull()] flights1.head(10)

	year	month	day	dep_time	dep_delay	arr_time	arr_delay	carrier	tailnum 1
0	2013	1	1	517.0	2.0	830.0	11.0	UA	N14228
1	2013	1	1	533.0	4.0	850.0	20.0	UA	N24211
2	2013	1	1	542.0	2.0	923.0	33.0	AA	N619AA
3	2013	1	1	554.0	-6.0	812.0	-25.0	DL	N668DN
4	2013	1	1	554.0	-4.0	740.0	12.0	UA	N39463

3. Remove all the observations with missing values

	year	month	day	dep_time	dep_delay	arr_time	arr_delay	carrier	tailr
0	2013	1	1	517.0	2.0	830.0	11.0	UA	N142
1	2013	1	1	533.0	4.0	850.0	20.0	UA	N24;
2	2013	1	1	542.0	2.0	923.0	33.0	AA	N619
3	2013	1	1	554.0	-6.0	812.0	-25.0	DL	N668
4	2013	1	1	554.0	-4.0	740.0	12.0	UA	N394
•••	•••	•••	•••						
160749	2013	9	30	2105.0	-1.0	2329.0	-25.0	UA	N477
160750	2013	9	30	2121.0	21.0	2349.0	-25.0	DL	N193
160751	2013	9	30	2140.0	0.0	10.0	-30.0	AA	N335
160752	2013	9	30	2149.0	-7.0	2245.0	-23.0	UA	N813
160753	2013	9	30	2233.0	80.0	112.0	42.0	UA	N578

157927 rows × 16 columns

4. Fill missing values with zeros for dep_delay

nomiss =flights['dep_delay'].fillna(0)
nomiss

nomiss.isnull().any()

False

- 5. Let's compute summary statistic per a group':
- 6. Using agg() methods for aggregation fetch min,max and mean for columns dep_delay and arr_delay:

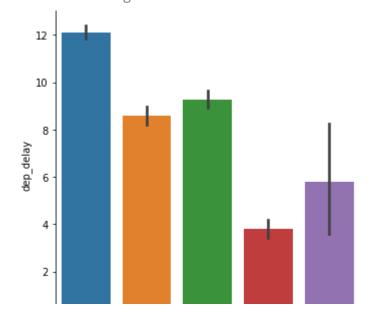
flights[['dep_delay','arr_delay']].agg(['min','mean','max'])

	dep_delay	arr_delay
min	-33.000000	-75.000000
mean	9.463773	2.094537
max	1014.000000	1007.000000

7. Using seaborn plot factorplot

```
import seaborn as sns
sns.catplot(x='carrier',y='dep_delay', data=flights, kind='bar')
```

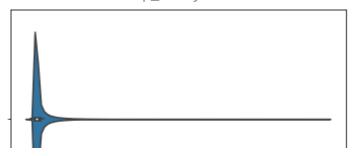
<seaborn.axisgrid.FacetGrid at 0x7fbdeb0ed760>



Violin plot

sns.violinplot(x = "dep_delay", data=flights)

<Axes: xlabel='dep_delay'>



```
sns.scatterplot(x='carrier', y='dep_delay', data=flights)
sns.jointplot(x='carrier', y='dep_delay', data=flights)
sns.boxplot(x='carrier',y='dep_delay', data=flights)
sns.swarmplot(x='carrier',y='dep_delay', data=flights)
```

```
KeyboardInterrupt
                                                 Traceback (most recent call last)
     <ipython-input-18-85e526d2b4d0> in <module>
     ---> 1 sns.swarmplot(x='carrier',v='dep delay', data=flights)
                                        14 frames
     /usr/local/lib/python3.9/dist-packages/seaborn/categorical.py in
     first_non_overlapping_candidate(self, candidates, neighbors)
                          dx = neighbors_x - x_i
        3502
        3503
                          dy = neighbors_y - y_i
     -> 3504
                          sq_distances = np.square(dx) + np.square(dy)
        3505
                          sep_needed = np.square(neighbors_r + r_i)
        3506
     KevboardInterrupt:
      SEARCH STACK OVERFLOW
     Error in callback <function flush figures at 0x7fbe1ea9f0d0> (for post execute)
     KevboardInterrupt
                                                Traceback (most recent call last)
     /usr/local/lib/python3.9/dist-packages/ipykernel/pylab/backend_inline.py in
     flush_figures()
         119
                      # ignore the tracking, just draw and close all figures
         120
     --> 121
                          return show(True)
         122
                      except Exception as e:
         123
                          # safely show traceback if in IPython, else raise
                                       18 frames
     <decorator-gen-2> in __call__(self, obj)
     /usr/local/lib/python3.9/dist-packages/seaborn/categorical.py in
     first_non_overlapping_candidate(self, candidates, neighbors)
        3509
                          # squared distance between candidate and any of the
     neighbors has
        3510
                          # to be at least square of the summed radii
     -> 3511
                          good_candidate = np.all(sq_distances >= sep_needed)
        3512
        3513
                          if good candidate:
     KeyboardInterrupt:
      SEARCH STACK OVERFLOW
sns.pairplot(flights)
Salaries data set
import pandas as pd
df = pd.read_csv("/content/Salaries.csv")
```

```
df.head(10)
```

Rename the columns

```
df_new =df.rename(columns={
          'discipline': 'subject',
          'sex': 'gender'
     })
df_new
```

Create a new column

```
df = df.assign( salary_k = lambda x: x.salary/1000.0)
df.head(10)
```

	rank	discipline	phd	service	sex	salary	salary_k
0	Prof	В	56	49	Male	186960	186.960
1	Prof	А	12	6	Male	93000	93.000
2	Prof	А	23	20	Male	110515	110.515
3	Prof	А	40	31	Male	131205	131.205
4	Prof	В	20	18	Male	104800	104.800
5	Prof	А	20	20	Male	122400	122.400
6	AssocProf	А	20	17	Male	81285	81.285
7	Prof	А	18	18	Male	126300	126.300
8	Prof	А	29	19	Male	94350	94.350
9	Prof	А	51	51	Male	57800	57.800

Check how many unique values in a column

Get frequency table for a categorical or binary column

```
df['rank'].value_counts()
```

Calculate the mean salary for men and women.

```
df.groupby('sex')['salary'].mean()
```

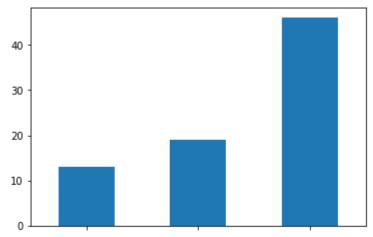
Group using 2 variables - sex and rank:

```
df.groupby(['rank','sex'], sort=True)[['salary']].mean()

%matplotlib inline
import matplotlib.pyplot as plt
plt.hist(df['salary'],bins=8, density=True)
```

df.groupby(['rank'])['salary'].count().plot(kind='bar')





Data Cleaning

```
import pandas as pd
data =pd.read_csv('/content/data.csv')
print(data)

data.describe()

data.rank()

data.isnull()

from numpy import NaN
data.replace({NaN:0.00})
```

```
data.fillna(3)
data.fillna(method='pad')
data.fillna(method='backfill') #Example of replacing missing values by filling backward
data.drop_duplicates()
del data['YOB']
print(data)
print(data.rename(columns={'Name':'FirstName','Surname':'LastName'}))
 Simple bar chart
data = {'Apple': 10, 'Orange': 15, 'Lemon': 5, 'Lime': 20}
names = list(data.keys())
values = list(data.values())
fig = plt.figure()
ax = plt.axes()
# ax.grid(linestyle='--', color='0.85') # Color: grayscale between 0 and 1
ax.bar(names, values);
data = {'Apple': 10, 'Orange': 15, 'Lemon': 5, 'Lime': 20, 'Bannan': 30, 'WM': 18, 'mango'
names = list(data.keys())
values = list(data.values())
fig = plt.figure(figsize=(10,7))
ax = plt.axes()
ax.bar(names, values, color='r', width = 0.40);
ax.set_xlabel('Fruit')
ax.set_ylabel('Quantity')
ax.set_title('Fruit Quantities');
data = {'Apple': 10,
        'Orange': 15,
        'Banana': 13,
        'Pear': 9,
        'Pomegrante': 11,
        'Lemon': 5,
        'Lime': 20}
names = list(data.keys())
values = list(data.values())
ax = plt.axes()
ax.barh(names, values, color='c', height=0.80)
                                                          # Horizontal bar-chart
```

```
ax.set_xlabel('Quantity')
ax.set_ylabel('Fruit')
ax.set_title('Fruit Quantities');
labels = ['G1', 'G2', 'G3', 'G4', 'G5']
men_means = [20, 34, 30, 35, 27]
women_means = [25, 32, 34, 20, 25]
x = np.arange(len(labels)) # the label locations
width = 0.40 # the width of the bars
fig, ax = plt.subplots()
rects1 = ax.bar(x - width/2, men_means, width, label='Men')
rects2 = ax.bar(x + width/2, women_means, width, label='Women')
# # Add some text for labels, title and custom x-axis tick labels, etc.
ax.set_ylabel('Scores')
ax.set_xlabel('Groups')
ax.set_title('Scores by group and gender')
ax.set_xticks(x)
ax.set_xticklabels(labels)
ax.legend();
plt.show()
 Scatter Plot
x = np.random.randn(20)
v = np.random.randn(20)
fig, ax = plt.subplots()
ax.scatter(x, y);
                            # a figure with a single Axes
fig, ax = plt.subplots()
ax.scatter(x, y, marker = "v");
fig, axs = plt.subplots(2, 3, sharex=True, sharey=True, figsize=(16,12));
# marker symbol
axs[0, 0].scatter(x, y, s=80, marker=">")
axs[0, 0].set_title("Right Triangle ")
# marker from TeX
axs[0, 1].scatter(x, y, s=80, marker=r'$\alpha$')
axs[0, 1].set_title(r"marker=r'\$\alpha\$'")
# marker from path
verts = [[-1, -1], [1, -1], [1, 1], [-1, -1]]
axs[0, 2].scatter(x, y, s=80, marker=verts)
axs[0, 2].set_title("marker=verts")
# regular polygon marker
axs[1, 0].scatter(x, y, s=80, marker=(5, 0))
axs[1, 0].set_title("polygon")
# regular star marker
axs[1, 1].scatter(x, y, s=80, marker=(5, 1))
axs[1, 1].set_title("star")
# regular asterisk marker
axs[1, 2].scatter(x, y, s=80, marker=(5, 2))
axs[1, 2].set_title("asterisk");
```