

Cont. Pandas - II

Session Objectives:

- Understand the key attributes of a DataFrame
- Use various DataFrame methods to explore and manipulate data
- Perform row and column operations
- Use joining, merging, and concatenation techniques across DataFrames
- Understand and perform statistical operations
- Conduct frequency and aggregation analysis
- Learn about sorting and indexing
- Export DataFrames to CSV/Excel files

```

import numpy as np
import pandas as pd
# Importing the DataSet.....
try:
    customers = pd.read_excel('anaconda_projects/CN-Python-Weekend/customers.xlsx')
    products = pd.read_excel('anaconda_projects/CN-Python-Weekend/products.xlsx')
    purchases = pd.read_excel('anaconda_projects/CN-Python-Weekend/purchases.xlsx')
except FileNotFoundError:
    print("File Not Found!")

File Not Found!

import os
os.getcwd()

'C:\\\\Users\\\\krish\\\\anaconda_projects\\\\CN-Python-Weekend'

import numpy as np
import pandas as pd
# Importing the DataSet.....
try:
    customers = pd.read_excel('C:\\\\Users\\\\krish\\\\anaconda_projects\\\\CN-Python-Weekend\\\\customers.xlsx')
    products = pd.read_excel('C:\\\\Users\\\\krish\\\\anaconda_projects\\\\CN-Python-Weekend\\\\products.xlsx')
    purchases = pd.read_excel('C:\\\\Users\\\\krish\\\\anaconda_projects\\\\CN-Python-Weekend\\\\purchases.xlsx')
except FileNotFoundError:
    print("File Not Found!")

```

customers.head()											
	id	first_name	last_name	email	gender	street_num	street_name	street_suffix	city	state	po:
0	1	Romain	Southcott	rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road	San Diego	California	9
1	2	Cosimo	Molyneaux	cmolyneaux1@wiley.com	Male	NaN	NaN	NaN	El Paso	Texas	
2	3	Bambi	Westrip	bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle	San Antonio	Texas	7
3	4	Roarke	Pankettman	rpankettman3@wiley.com	Male	74.0	Debs	Point	Memphis	Tennessee	3
4	5	Mikaela	Althorpe	malthorpe4@51.la	NaN	NaN	2nd	Drive	NaN	NaN	8

```
# DataFrame Attributes [.index] -> range
customers.index # range(Start=0,Stop=1000,Step=1)
```

```
RangeIndex(start=0, stop=1000, step=1)
```

```
products.index
```

```
RangeIndex(start=0, stop=60, step=1)
```

```
purchases.index
```

```
RangeIndex(start=0, stop=6000, step=1)
```

```
# Numpy 2D Array -> .shape(rows,cols)
# DataFrame.shape
customers.shape
```

```
(1000, 11)
```

```
products.shape
```

```
(60, 5)
```

```
purchases.shape
```

```
(6000, 7)
```

```
customers_rows = customers.shape[0]
customers_rows
```

```
1000
```

```
customers_cols = customers.shape[1]
customers_cols
```

```
11
```

```
# .columns [providing column headers details (label)]
customers.columns

Index(['id', 'first_name', 'last_name', 'email', 'gender', 'street_num',
       'street_name', 'street_suffix', 'city', 'state', 'postcode'],
      dtype='object')

products.columns

Index(['Unnamed: 0', 'id', 'product', 'cost', 'company'], dtype='object')

purchases.columns

Index(['Unnamed: 0', 'id', 'purch_date', 'customer_num', 'product_num',
       'amount', 'paid'],
      dtype='object')
```

```
# .dtypes [data types of each columns]
customers.dtypes

id           int64
first_name    object
last_name     object
email         object
gender        object
street_num    float64
street_name   object
street_suffix object
city          object
state          object
postcode      float64
dtype: object

products.dtypes

Unnamed: 0    int64
id           int64
product      object
cost          object
company      object
dtype: object
```

```
purchases.dtypes

Unnamed: 0    int64
id           int64
purch_date   object
customer_num int64
product_num  int64
amount        int64
paid          object
dtype: object

purchases.head()

  Unnamed: 0  id  purch_date  customer_num  product_num  amount  paid
```

```
purchases.head()
```

	Unnamed: 0	id	purch_date	customer_num	product_num	amount	paid
0	0	1	2019-01-03 00:00:00		823	27	12 \$568.92
1	1	2	2019-01-03 00:00:00		606	28	14 \$395.36
2	2	3	2019-01-03 00:00:00		955	9	17 \$510.17
3	3	4	2019-01-03 00:00:00		577	19	3 \$68.49
4	4	5	2019-01-03 00:00:00		429	8	18 \$759.42

```
# .describe() -> Statistical Summary [Continuous Vs Categorical]  
customers.describe() # Default cont.
```

	id	street_num	postcode
count	1000.000000	738.000000	843.000000
mean	500.500000	10536.439024	52669.548043
std	288.819436	23050.537603	28140.041026
min	1.000000	0.000000	214.000000
25%	250.750000	21.000000	29279.500000
50%	500.500000	445.500000	48232.000000
75%	750.250000	6918.500000	78337.500000
max	1000.000000	99918.000000	99812.000000

```
customers.describe(include = 'object')
```

	first_name	last_name	email	gender	street_name	street_suffix	city	state
count	1000	1000	878	957	963	963	921	920
unique	932	993	878	2	427	21	296	48
top	Berty	Sedworth	rsouthcott0@clickbank.net	Male	Arizona	Place	Washington	Texas
freq	4	3	1	485	6	57	29	107

```
customers.describe(include = 'O')
```

	first_name	last_name	email	gender	street_name	street_suffix	city	state
count	1000	1000	878	957	963	963	921	920
unique	932	993	878	2	427	21	296	48
top	Berty	Sedworth	rsouthcott0@clickbank.net	Male	Arizona	Place	Washington	Texas
freq	4	3	1	485	6	57	29	107

customers.describe(include = 'all')											
	id	first_name	last_name		email	gender	street_num	street_name	street_suffix	ci	ci
count	1000.000000	1000	1000		878	957	738.000000	963	963	9	9
unique		Nan	932	993		878	2	Nan	427	21	2
top		Nan	Berty	Sedworth	rsouthcott0@clickbank.net	Male		Nan	Arizona	Place	Washington
freq		Nan	4	3		1	485		Nan	6	57
mean	500.500000		Nan	Nan		Nan	Nan	10536.439024	Nan	Nan	Nan
std	288.819436		Nan	Nan		Nan	Nan	23050.537603	Nan	Nan	Nan
min	1.000000		Nan	Nan		Nan	Nan	0.000000	Nan	Nan	Nan
25%	250.750000		Nan	Nan		Nan	Nan	21.000000	Nan	Nan	Nan
50%	500.500000		Nan	Nan		Nan	Nan	445.500000	Nan	Nan	Nan
75%	750.250000		Nan	Nan		Nan	Nan	6918.500000	Nan	Nan	Nan
max	1000.000000		Nan	Nan		Nan	Nan	99918.000000	Nan	Nan	Nan

```
# .values []
customers.values
```

```
array([[1, 'Romain', 'Southcott', ..., 'San Diego', 'California',
       92127.0],
       [2, 'Cosimo', 'Molyneaux', ..., 'El Paso', 'Texas', nan],
       [3, 'Bambi', 'Westrip', ..., 'San Antonio', 'Texas', 78220.0],
       ...,
       [998, 'Tobit', 'Birt', ..., 'Waterbury', 'Connecticut', 6721.0],
       [999, 'Issiah', 'Standbrooke', ..., 'Savannah', 'Georgia',
        31405.0],
       [1000, 'Elmore', 'Malpas', ..., 'Atlanta', 'Georgia', 30386.0]],
      dtype=object)
```

```
products.values
```

```
array([[0, 1, 'Liners - Baking Cups', '$6.36', 'Skipfire'],
       [1, 2, 'Nori Sea Weed - Gold Label', '$85.74', 'Dynazzy'],
       [2, 3, 'Bar Bran Honey Nut', '$65.40', 'Ntag'],
       [3, 4, 'Soup - Campbells Beef Stew', '$68.16', 'Photojam'],
       [4, 5, 'Wine - Shiraz Wolf Blass Premium', '$87.39', 'Eare'],
       [5, 6, 'Wine - White, Riesling, Semi - Dry', '$99.22', 'Livepath'],
       [6, 7, 'Brandy - Bar', '$13.83', 'Oloo'],
       [7, 8, 'Onions - White', '$42.19', 'Oozz'],
       [8, 9, 'Lettuce - Baby Salad Greens', '$30.01', 'Meevee'],
       [9, 10, 'Sambuca - Ramazzotti', '$88.99', 'Livepath'],
       [10, 11, 'Coffee - Decafenated', '$21.47', 'Meezzy'],
       [11, 12, 'Lamb Leg - Bone - In Nz', '$42.56', 'Photofeed'],
       [12, 13, 'Lettuce - Spring Mix', '$19.90', 'Twitterworks'],
       [13, 14, 'Cookies Oatmeal Raisin', '$99.40', 'Dynabox'],
       [14, 15, 'Cookies Oatmeal Raisin', '$14.13', 'Vitz'],
       [15, 16, 'Sausage - Chorizo', '$55.45', nan],
       [16, 17, 'Glaze - Apricot', '$33.78', 'Browsezoom']])
```

```
purchases.values  
  
array([[0, 1, datetime.datetime(2019, 1, 3, 0, 0), ..., 27, 12,  
       '$568.92'],  
      [1, 2, datetime.datetime(2019, 1, 3, 0, 0), ..., 28, 14,  
       '$395.36'],  
      [2, 3, datetime.datetime(2019, 1, 3, 0, 0), ..., 9, 17, '$510.17'],  
      ...,  
      [5997, 5998, '06/20/2019', ..., 19, 9, '$205.47'],  
      [5998, 5999, '06/20/2019', ..., 11, 20, '$429.40'],  
      [5999, 6000, '06/20/2019', ..., 57, 4, '$274.52']], dtype=object)  
  
# .size -> .shape(1000,11) -> 1000 * 11 = 11000  
customers.size # -> Total Cell in Customer DataFrame  
  
11000  
  
products.size  
  
300  
  
purchases.size  
  
42000
```

```
# .empty [Boolean Return][T/F] -> DataFrame is empty or not  
customers.empty # False  
  
False  
  
products.empty # False  
  
False  
  
purchases.empty  
  
False  
  
empty_df = pd.DataFrame()  
print(empty_df)  
print(empty_df.empty) # True  
  
Empty DataFrame  
Columns: []  
Index: []  
True
```

```
# .ndim -> N-Dimension  
customers.ndim # DataFrame -> 2 Dimension  
  
2  
  
products.ndim  
  
2  
  
purchases.ndim  
  
2
```

```
# .head() & .tail()
customers.head(7) # default is 'First 5'
```

↶ ↷ ⌂ ⌃ ⌄ ⌅ ⌆ ⌇ ⌈ ⌉ ⌊ ⌋

	id	first_name	last_name		email	gender	street_num	street_name	street_suffix	city	state	po
0	1	Romain	Southcott	rsouthcott0@clickbank.net	Male		1.0	Trailsway	Road	San Diego	California	9
1	2	Cosimo	Molyneaux	cmolyneaux1@wiley.com	Male		NaN	NaN	NaN	El Paso	Texas	
2	3	Bambi	Westrip	bwestrip2@symantec.com	Female		4057.0	Arkansas	Circle	San Antonio	Texas	7
3	4	Roarke	Pankettman	rpankettman3@wiley.com	Male		74.0	Debs	Point	Memphis	Tennessee	3
4	5	Mikaela	Althorpe	malthorpe4@51.la	NaN		NaN	2nd	Drive	NaN	NaN	8
5	6	Magdalena	Cullip	mcullip5@tiny.cc	Female		9190.0	Packers	Drive	Baltimore	Maryland	2
6	7	Marietta	Heball	mheball6@blog.com	Female		12.0	Mallory	Center	Carol Stream	Illinois	6

```
products.tail(7) # default is 'Last 5'
```

	Unnamed: 0	id		product	cost	company
53		53	54	Plate - Foam, Bread And Butter	\$37.79	Flipstorm
54		54	55	Beef - Bresaola	\$8.55	Digitube
55		55	56	Peach - Fresh	\$20.53	Zazio
56		56	57	Muffin Hinge - 211n	\$68.63	Fanoodle
57		57	58	Bar Bran Honey Nut	\$73.05	Yakijo
58		58	59	Nut - Almond, Blanched, Whole	\$74.28	Eazzy
59		59	60	Table Cloth 90x90 Colour	\$41.22	Quinu

```
# Case Study => Web Scrap data from .screener() -> Stock Analysis [.Transpose]
```

	0	1	2	3	4	5	6	7	8	9	...	50	51
Unnamed:	0	1	2	3	4	5	6	7	8	9	...	50	51
id	1	2	3	4	5	6	7	8	9	10	...	51	52
product	Liners - Baking Cups	Nori Sea Weed - Gold Label	Bar Bran Honey Nut	Soup - Campbells Beef Stew	Wine - Shiraz Wolf Blass Premium	Wine - White, Riesling, Semi - Dry	Brandy - Bar	Onions - White	Lettuce - Baby Salad Greens	Sambuca - Ramazzotti	...	Ecolab - Power Fusion	Soup - Knorr, Country Bean
cost	\$6.36	\$85.74	\$65.40	\$68.16	\$87.39	\$99.22	\$13.83	\$42.19	\$30.01	\$88.99	...	\$75.32	\$98.74
company	Skipfire	Dynazzy	Ntag	Photojam	Eare	Livepath	Oloo	Oizz	Meevee	Livepath	...	Fliptune	Eimbee

5 rows × 60 columns

products.transpose()														
	0	1	2	3	4	5	6	7	8	9	...	50	51	
Unnamed:	0	0	1	2	3	4	5	6	7	8	9	...	50	51
	id	1	2	3	4	5	6	7	8	9	10	...	51	52
product	Liners - Baking Cups	Nori Sea Weed - Gold Label	Bar Bran Honey Nut	Soup - Campbells Beef Stew	Wine - Shiraz Wolf Blass Premium	Wine - White, Riesling, Semi - Dry	Brandy - Bar	Onions - White	Lettuce - Baby Salad Greens	Sambuca - Ramazzotti	...	Ecolab - Power Fusion	Soup - Knorr, Country Bean	Th
cost	\$6.36	\$85.74	\$65.40	\$68.16	\$87.39	\$99.22	\$13.83	\$42.19	\$30.01	\$88.99	...	\$75.32	\$98.74	
company	Skipfire	Dynazzy	Ntag	Photojam	Eare	Livepath	Oloo	Oozz	Meevee	Livepath	...	Fliptune	Eimbee	To
5 rows × 60 columns														

# products.T.T products.transpose().transpose()													
	Unnamed: 0	id		product	cost	company							
0	0	1		Liners - Baking Cups	\$6.36	Skipfire							
1	1	2		Nori Sea Weed - Gold Label	\$85.74	Dynazzy							
2	2	3		Bar Bran Honey Nut	\$65.40	Ntag							
3	3	4		Soup - Campbells Beef Stew	\$68.16	Photojam							
4	4	5		Wine - Shiraz Wolf Blass Premium	\$87.39	Eare							
5	5	6		Wine - White, Riesling, Semi - Dry	\$99.22	Livepath							
6	6	7		Brandy - Bar	\$13.83	Oloo							
7	7	8		Onions - White	\$42.19	Oozz							
8	8	9		Lettuce - Baby Salad Greens	\$30.01	Meevee							

# Operations on Rows And Columns in DataFrame [Feature Engineering]											
# Create a new column name 'Country' -> 'United States'											
customers['country'] = 'United States'											
customers.head()											
name	last_name		email	gender	street_num	street_name	street_suffix	city	state	postcode	country
John	Southcott		rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road	San Diego	California	92127.0	United States
John	Molyneaux		cmolyneaux1@wiley.com	Male	NaN	NaN	NaN	El Paso	Texas	NaN	United States
John	Westrip		bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle	San Antonio	Texas	78220.0	United States
John	Pankettman		rpankettman3@wiley.com	Male	74.0	Debs	Point	Memphis	Tennessee	38150.0	United States
John	Althorpe		malthorpe4@51.la	NaN	NaN	2nd	Drive	NaN	NaN	83732.0	United States

```
# Membership_status - ['Diamond','Platinum','Gold','Silver','Bronze']
membership_status = ['Diamond','Platinum','Gold','Silver','Bronze']
customers['membership_status'] = membership_status
# ValueError: Length of values (5) does not match length of index (1000)
```

```
# .loc[rows , cols]
membership_status = ['Diamond','Platinum','Gold','Silver','Bronze']
customers.loc[:4,'membership_status'] = membership_status
# Rest 995 Records are NaN
customers
```

	email	gender	street_num	street_name	street_suffix	city	state	postcode	country	membership_status
@clickbank.net	Male	1.0	Trailsway	Road	San Diego	California	92127.0	United States	Diamond	
ux1@wiley.com	Male	Nan	Nan	Nan	El Paso	Texas	Nan	United States	Platinum	
psymantec.com	Female	4057.0	Arkansas	Circle	San Antonio	Texas	78220.0	United States	Gold	
on3@wiley.com	Male	74.0	Debs	Point	Memphis	Tennessee	38150.0	United States	Silver	
thorpe4@51.la	Nan	Nan	2nd	Drive	Nan	Nan	83732.0	United States	Bronze	

```
# 1000 records // 5 count of membership_status -> 200
# np.tile(data , repetition)
# Construct an array by repeating A the number of times given by reps.
membership_status = ['Diamond','Platinum','Gold','Silver','Bronze']
customers['Membership_Status'] = np.tile(membership_status , len(customers) // len(membership_status))
# Rest 995 Records are NaN
customers
```

r	street_num	street_name	street_suffix	city	state	postcode	country	membership_status	Membership_Status
e	1.0	Trailsway	Road	San Diego	California	92127.0	United States	Diamond	Diamond
e	Nan	Nan	Nan	El Paso	Texas	Nan	United States	Platinum	Platinum
e	4057.0	Arkansas	Circle	San Antonio	Texas	78220.0	United States	Gold	Gold
e	74.0	Debs	Point	Memphis	Tennessee	38150.0	United States	Silver	Silver
N	Nan	2nd	Drive	Nan	Nan	83732.0	United States	Bronze	Bronze
...

```
customers['Membership_Status'].value_counts()
```

Membership_Status

```
Diamond    200  
Platinum   200  
Gold       200  
Silver     200  
Bronze     200  
Name: count, dtype: int64
```

Assigning Based on Conditions

```
customers['membershipStatus'] = customers['state'].apply(  
    lambda x : 'Gold' if x == 'Texas' else 'Silver'  
)  
customers
```

t_name	street_suffix	city	state	postcode	country	membership_status	Membership_Status	membershipStatus
Tailsway	Road	San Diego	California	92127.0	United States	Diamond	Diamond	Silver
NaN	NaN	El Paso	Texas	NaN	United States	Platinum	Platinum	Gold
Kansas	Circle	San Antonio	Texas	78220.0	United States	Gold	Gold	Gold
Debs	Point	Memphis	Tennessee	38150.0	United States	Silver	Silver	Silver
2nd	Drive	NaN	NaN	83732.0	United States	Bronze	Bronze	Silver
...

Assigning Based on Conditions

```
customers['membershipStatus'] = customers['state'].apply(  
    lambda x : 'Gold' if (x == 'Texas') or (x == 'California') or (x == 'Georgia') else 'Silver'  
)  
customers
```

t_name	street_suffix	city	state	postcode	country	membership_status	Membership_Status	membershipStatus
Tailsway	Road	San Diego	California	92127.0	United States	Diamond	Diamond	Gold
NaN	NaN	El Paso	Texas	NaN	United States	Platinum	Platinum	Gold
Kansas	Circle	San Antonio	Texas	78220.0	United States	Gold	Gold	Gold
Debs	Point	Memphis	Tennessee	38150.0	United States	Silver	Silver	Silver
2nd	Drive	NaN	NaN	83732.0	United States	Bronze	Bronze	Silver
...

```
customers.columns
```

```
Index(['id', 'first_name', 'last_name', 'email', 'gender', 'street_num',  
       'street_name', 'street_suffix', 'city', 'state', 'postcode', 'country',  
       'membership_status', 'Membership_Status', 'membershipStatus'],  
      dtype='object')
```

```
customers.dtypes
```

```
id          int64
first_name    object
last_name     object
email         object
gender        object
street_num    float64
street_name   object
street_suffix object
city          object
state         object
postcode      float64
country       object
membership_status  object
Membership_Status  object
membershipStatus  object
dtype: object
```

```
products.columns
```

```
Index(['Unnamed: 0', 'id', 'product', 'cost', 'company'], dtype='object')
```

```
products.dtypes
```

```
Unnamed: 0    int64
id           int64
product      object
cost          object
company      object
dtype: object
```

```
# Row Insertions = INSERT Command in SQL
```

```
products.loc['60'] = [60, 61, 'Iphone 17 Pro Max', '$1599', 'Apple']
products.loc['61'] = [61, 62, 'Nokia Lumia', '$400', 'Nokia']
products.loc['62'] = [62, 63, 'Nothing Phone 2', '$1000', 'Nothing']
products.loc['63'] = [63, 64, 'Moto G', '$700', 'Motorola']
products.loc['64'] = [64, 65, 'Redmi 15 Pro', '$800', 'Redmi']
products
```

	Unnamed: 0	id	product	cost	company
0	0	1	Liners - Baking Cups	\$6.36	Skipfire
1	1	2	Nori Sea Weed - Gold Label	\$85.74	Dynazzy
2	2	3	Bar Bran Honey Nut	\$65.40	Ntag
3	3	4	Soup - Campbells Beef Stew	\$68.16	Photojam
4	4	5	Wine - Shiraz Wolf Blass Premium	\$87.39	Eare
...
60	60	61	Iphone 17 Pro Max	\$1599	Apple
61	61	62	Nokia Lumia	\$400	Nokia
62	62	63	Nothing Phone 2	\$1000	Nothing
63	63	64	Moto G	\$700	Motorola
64	64	65	Redmi 15 Pro	\$800	Redmi

65 rows × 5 columns

```
products.loc['65'] = None
products
```

	Unnamed: 0	id	product	cost	company
0	0.0	1.0	Liners - Baking Cups	\$6.36	Skipfire
1	1.0	2.0	Nori Sea Weed - Gold Label	\$85.74	Dynazzy
2	2.0	3.0	Bar Bran Honey Nut	\$65.40	Ntag
3	3.0	4.0	Soup - Campbells Beef Stew	\$68.16	Photojam
4	4.0	5.0	Wine - Shiraz Wolf Blass Premium	\$87.39	Eare
...
61	61.0	62.0	Nokia Lumia	\$400	Nokia
62	62.0	63.0	Nothing Phone 2	\$1000	Nothing
63	63.0	64.0	Moto G	\$700	Motorola
64	64.0	65.0	Redmi 15 Pro	\$800	Redmi
65	NaN	NaN	NaN	NaN	NaN

66 rows × 5 columns

```
# Dropping a row -> axis = 0 [row dropping][Horizontal Axis]
products.drop(index = '65' , axis = 0 , inplace = True)
products
```

	Unnamed: 0	id	product	cost	company
0	0.0	1.0	Liners - Baking Cups	\$6.36	Skipfire
1	1.0	2.0	Nori Sea Weed - Gold Label	\$85.74	Dynazzy
2	2.0	3.0	Bar Bran Honey Nut	\$65.40	Ntag
3	3.0	4.0	Soup - Campbells Beef Stew	\$68.16	Photojam
4	4.0	5.0	Wine - Shiraz Wolf Blass Premium	\$87.39	Eare
...
60	60.0	61.0	Iphone 17 Pro Max	\$1599	Apple
61	61.0	62.0	Nokia Lumia	\$400	Nokia
62	62.0	63.0	Nothing Phone 2	\$1000	Nothing
63	63.0	64.0	Moto G	\$700	Motorola
64	64.0	65.0	Redmi 15 Pro	\$800	Redmi

65 rows × 5 columns

```
products.iloc[-5:]
```

	Unnamed: 0	id	product	cost	company
55	55.0	56.0	Peach - Fresh	\$20.53	Zazio
56	56.0	57.0	Muffin Hinge - 211n	\$68.63	Fanoode
57	57.0	58.0	Bar Bran Honey Nut	\$73.05	Yakijo
58	58.0	59.0	Nut - Almond, Blanched, Whole	\$74.28	Eazzy
59	59.0	60.0	Table Cloth 90x90 Colour	\$41.22	Quinu

```
drop_slicing = products.iloc[-5:].index
```

```
drop_slicing
```

```
Index(['60', '61', '62', '63', '64'], dtype='object')
```

```
# # Dropping a row -> axis = 0 [row dropping][Horizontal Axis]
# products.drop(index = ['63' , '64'] , axis = 0 , inplace = True)
drop_slicing = products.iloc[-5:].index
products.drop(index = drop_slicing , axis = 0 , inplace = True)
products
```

	Unnamed: 0	id	product	cost	company
0	0.0	1.0	Liners - Baking Cups	\$6.36	Skipfire
1	1.0	2.0	Nori Sea Weed - Gold Label	\$85.74	Dynazzy
2	2.0	3.0	Bar Bran Honey Nut	\$65.40	Ntag
3	3.0	4.0	Soup - Campbells Beef Stew	\$68.16	Photojam
4	4.0	5.0	Wine - Shiraz Wolf Blass Premium	\$87.39	Eare
5	5.0	6.0	Wine - White, Riesling, Semi - Dry	\$99.22	Livepath
6	6.0	7.0	Brandy - Bar	\$13.83	Oloo
7	7.0	8.0	Onions - White	\$42.19	Oozz
8	8.0	9.0	Lettuce - Baby Salad Greens	\$30.01	Meevee

```
# Row Insertions = INSERT Command in SQL
products.loc[60] = [60, 61, 'Iphone 17 Pro Max', '$1599', 'Apple']
products.loc[61] = [61, 62, 'Nokia Lumia', '$400', 'Nokia']
products.loc[62] = [62, 63, 'Nothing Phone 2', '$1000', 'Nothing']
products.loc[63] = [63, 64, 'Moto G', '$700', 'Motorola']
products.loc[64] = [64, 65, 'Redmi 15 Pro', '$800', 'Redmi']
products
```

	Unnamed: 0	id	product	cost	company
0	0.0	1.0	Liners - Baking Cups	\$6.36	Skipfire
1	1.0	2.0	Nori Sea Weed - Gold Label	\$85.74	Dynazzy
2	2.0	3.0	Bar Bran Honey Nut	\$65.40	Ntag
3	3.0	4.0	Soup - Campbells Beef Stew	\$68.16	Photojam
4	4.0	5.0	Wine - Shiraz Wolf Blass Premium	\$87.39	Eare
...
60	60.0	61.0	Iphone 17 Pro Max	\$1599	Apple
61	61.0	62.0	Nokia Lumia	\$400	Nokia
62	62.0	63.0	Nothing Phone 2	\$1000	Nothing
63	63.0	64.0	Moto G	\$700	Motorola
64	64.0	65.0	Redmi 15 Pro	\$800	Redmi

```
products.iloc[-5::2].index # This will drop [alternative] records from last 5 rows
Index([60, 62, 64], dtype='int64')

# .between(X,Y) X & Y Both are inclusive
products[products['id'].between(61,65)]
```

	Unnamed: 0	id	product	cost	company
60	60.0	61.0	Iphone 17 Pro Max	\$1599	Apple
61	61.0	62.0	Nokia Lumia	\$400	Nokia
62	62.0	63.0	Nothing Phone 2	\$1000	Nothing
63	63.0	64.0	Moto G	\$700	Motorola
64	64.0	65.0	Redmi 15 Pro	\$800	Redmi

```
idx_range = products[products['id'].between(61,65)].index
idx_range
```

```
Index([60, 61, 62, 63, 64], dtype='int64')
```

```
# Impact Analysis -> Out of 1000 records , 157 are missing [<5%] XX
customers.drop(index = postcode_idx , axis = 0 , inplace = True)
customers
```

```
# Imputing the missing Values
customers['postcode'] = customers['postcode'].fillna(0).astype(int)
customers.head()
```

first_name	last_name	email	gender	street_num	street_name	street_suffix	city	state	postcode
Romain	Southcott	rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road	San Diego	California	92127
Cosimo	Molyneaux	cmolyneaux1@wiley.com	Male	NaN	NaN	NaN	El Paso	Texas	0
Bambi	Westrip	bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle	San Antonio	Texas	78220
Roarke	Pankettman	rpankettman3@wiley.com	Male	74.0	Debs	Point	Memphis	Tennessee	38150
Mikaela	Althorpe	malthorpe4@51.la	NaN	NaN	2nd	Drive	NaN	NaN	83732

```
customers.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 1000 entries, 0 to 999
Data columns (total 15 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   id               1000 non-null    float64
 1   first_name       1000 non-null    object 
 2   last_name        1000 non-null    object 
 3   email            878 non-null    object 
 4   gender           957 non-null    object 
 5   street_num       738 non-null    float64
 6   street_name      963 non-null    object 
 7   street_suffix    963 non-null    object 
 8   city              921 non-null    object 
 9   state             920 non-null    object 
 10  postcode          1000 non-null    int32  
 11  country           1000 non-null    object 
 12  membership_status 1000 non-null    object 
 13  Membership_Status 1000 non-null    object 
 14  membershipStatus 1000 non-null    object 
dtypes: float64(2), int32(1), object(12)
memory usage: 153.4+ KB
```

```
# Calculated Columns , Add Column [Column From Examples]
```

```
customers['customer_name'] = customers['first_name'] + ' ' + customers['last_name']
customers
```

ffix	city	state	postcode	country	membership_status	Membership_Status	membershipStatus	customer_name
oad	San Diego	California	92127	United States	Diamond	Diamond	Gold	Romain Southcott
aN	El Paso	Texas	0	United States	Platinum	Platinum	Gold	Cosimo Molyneaux
rcle	San Antonio	Texas	78220	United States	Gold	Gold	Gold	Bambi Westrip
oint	Memphis	Tennessee	38150	United States	Silver	Silver	Silver	Roarke Pankettman
rive	NaN	NaN	83732	United States	Bronze	Bronze	Silver	Mikaela Althorpe
...

```
products[['company', 'product', 'cost']]
```

	company	product	cost
0	Skipfire	Liners - Baking Cups	\$6.36
1	Dynazzy	Nori Sea Weed - Gold Label	\$85.74
2	Ntag	Bar Bran Honey Nut	\$65.40
3	Photojam	Soup - Campbells Beef Stew	\$68.16
4	Eare	Wine - Shiraz Wolf Blass Premium	\$87.39
5	Livepath	Wine - White, Riesling, Semi - Dry	\$99.22
6	Oloo	Brandy - Bar	\$13.83
7	Oozz	Onions - White	\$42.19

```
customers.columns
```

```
Index(['id', 'first_name', 'last_name', 'email', 'gender', 'street_num',
       'street_name', 'street_suffix', 'city', 'state', 'postcode', 'country',
       'membership_status', 'Membership_Status', 'membershipStatus',
       'customer_name'],
      dtype='object')
```

```
customer_col_drop = customers.iloc[:, -5: ].columns
customer_col_drop
```

```
Index(['country', 'membership_status', 'Membership_Status', 'membershipStatus',
       'customer_name'],
      dtype='object')
```

```
# customers.drop(columns = ['country', 'membership_status', 'Membership_Status',
#                   'membershipStatus', 'customer_name'] , axis = 1 , inplace = True)
customers.drop(columns = customer_col_drop , axis = 1 , inplace = True)
customers
```

ast_name	email	gender	street_num	street_name	street_suffix	city	state	postcode
Southcott	rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road	San Diego	California	92127
Molyneaux	cmolyneaux1@wiley.com	Male	NaN	NaN	NaN	El Paso	Texas	0
Westrip	bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle	San Antonio	Texas	78220
Pankettman	rpankettman3@wiley.com	Male	74.0	Debs	Point	Memphis	Tennessee	38150
Althorpe	malthorpe4@51.la	NaN	NaN	2nd	Drive	NaN	NaN	83732
...
Alman	malmanrn@cornell.edu	Female	0.0	Thompson	Way	Reading	Pennsylvania	19610
Heckle	whecklero@fc2.com	Female	701.0	Rowland	Hill	Indianapolis	Indiana	46278
Rirt	NaN	Male	2.0	Rasil	Road	Waterbury	Connecticut	6721

```

# Renaming the row index
index_rename = {
    0 : 'FirstCustomers',
    1 : 'SecondCustomers',
    2 : 'ThirdCustomers',
    3 : 'FourthCustomers',
    4 : 'FifthCustomers',
}
# customers.rename(index = index_rename , inplace = True)
customers = customers.rename(index = index_rename)
customers

```

	id	first_name	last_name		email	gender	street_num	street_name	street_sufi
FirstCustomers	1.0	Romain	Southcott		rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road
SecondCustomers	2.0	Cosimo	Molyneaux		cmolyneaux1@wiley.com	Male	NaN	NaN	NaN
ThirdCustomers	3.0	Bambi	Westrip		bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle
FourthCustomers	4.0	Roarke	Pankettman		rpankettman3@wiley.com	Male	74.0	Debs	Point
FifthCustomers	5.0	Mikaela	Althorpe		malthorpe4@51.la	NaN	NaN	2nd	Drive
...
995	996.0	Merrili	Alman		malmanrn@cornell.edu	Female	0.0	Thompson	Way
996	997.0	Winonah	Heckle		whecklero@fc2.com	Female	701.0	Rowland	Hill
997	998.0	Tobit	Birt		NaN	Male	2.0	Basil	Road
998	999.0	Issiah	Standbrooke		istandbrookerq@yellowpages.com	Male	NaN	Kenwood	Savannah
999	1000.0	Elmore	Malpas		NaN	Male	205.0	Farwell	Park

1000 rows × 11 columns

```

# Renaming the column headers
col_rename = {
    'street_num' : 'street_number',
    'street_name' : 'street_address',
}
# customers.rename(columns = col_rename , inplace = True)
customers = customers.rename(columns = col_rename)
customers

```

	id	first_name	last_name		email	gender	street_number	street_address	street_suffix
0	1	Romain	Southcott		rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road
1	2	Cosimo	Molyneaux		cmolyneaux1@wiley.com	Male	NaN	NaN	NaN
2	3	Bambi	Westrip		bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle
3	4	Roarke	Pankettman		rpankettman3@wiley.com	Male	74.0	Debs	Point
4	5	Mikaela	Althorpe		malthorpe4@51.la	NaN	NaN	2nd	Drive
...
995	996	Merrili	Alman		malmanrn@cornell.edu	Female	0.0	Thompson	Way
996	997	Winonah	Heckle		whecklero@fc2.com	Female	701.0	Rowland	Hill
997	998	Tobit	Birt		NaN	Male	2.0	Basil	Road
998	999	Issiah	Standbrooke		istandbrookerq@yellowpages.com	Male	NaN	Kenwood	Drive
999	1000	Elmore	Malpas		NaN	Male	205.0	Farwell	Park

1000 rows × 11 columns

```
# SELECT DISTINCT FROM Tables -> Unique Values
customers['state'].shape

(1000,)

customers['state'].ndim # Series

1

customers['state'].unique()

array(['California', 'Texas', 'Tennessee', nan, 'Maryland', 'Illinois',
       'Missouri', 'Alaska', 'Louisiana', 'Florida', 'New York',
       'District of Columbia', 'South Carolina', 'Washington', 'Nevada',
       'Indiana', 'Pennsylvania', 'Georgia', 'Virginia', 'Michigan',
       'Iowa', 'Kentucky', 'Oregon', 'North Carolina', 'Ohio', 'Kansas',
       'West Virginia', 'Oklahoma', 'North Dakota', 'Arizona', 'Utah',
       'Minnesota', 'Wisconsin', 'New Jersey', 'Montana', 'New Hampshire',
       'Colorado', 'Massachusetts', 'Alabama', 'Delaware', 'Mississippi',
       'Idaho', 'Arkansas', 'Hawaii', 'Nebraska', 'Connecticut',
       'Rhode Island', 'New Mexico', 'Maine'], dtype=object)

len(customers['state'].unique())

49

customers['state'].nunique() # Count of Unique States

48
```

```
customers['gender'].unique()

array(['Male', 'Female', nan], dtype=object)

customers['gender'].nunique() # 2 [Exclude 'NaN' values]

2

# Value_Counts() [Freq]
products['company'].value_counts().head(5)

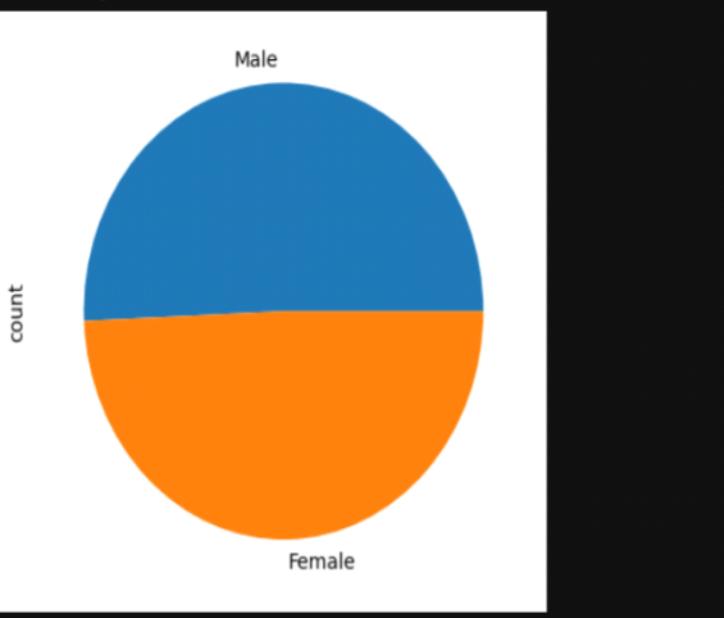
company
Vitz      2
Ntag      2
Livepath  2
Cogibox   2
Browsezoom 2
Name: count, dtype: int64

customers['gender'].value_counts()

gender
Male     485
Female   472
Name: count, dtype: int64
```

```
customers['gender'].value_counts().plot(kind = 'pie')
```

```
<Axes: ylabel='count'>
```



```
purchases.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6000 entries, 0 to 5999
Data columns (total 7 columns):
 #   Column      Non-Null Count  Dtype  
 ---  --          -----          ----  
 0   Unnamed: 0    6000 non-null   int64  
 1   id           6000 non-null   int64  
 2   purch_date   6000 non-null   object 
 3   customer_num 6000 non-null   int64  
 4   product_num  6000 non-null   int64  
 5   amount        6000 non-null   int64  
 6   paid          6000 non-null   object 
dtypes: int64(5), object(2)
memory usage: 328.3+ KB
```

```
purchases['paid'] = purchases['paid'].astype('str').str.replace('[,$,]', '' , regex = True).astype('float')
purchases.head()
```

	Unnamed: 0	id	purch_date	customer_num	product_num	amount	paid
0	0	1	2019-01-03 00:00:00	823	27	12	568.92
1	1	2	2019-01-03 00:00:00	606	28	14	395.36
2	2	3	2019-01-03 00:00:00	955	9	17	510.17
3	3	4	2019-01-03 00:00:00	577	19	3	68.49
4	4	5	2019-01-03 00:00:00	429	8	18	759.42

```

purchases.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6000 entries, 0 to 5999
Data columns (total 7 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   Unnamed: 0    6000 non-null   int64  
 1   id           6000 non-null   int64  
 2   purch_date   6000 non-null   object  
 3   customer_num 6000 non-null   int64  
 4   product_num  6000 non-null   int64  
 5   amount        6000 non-null   int64  
 6   paid          6000 non-null   float64 
dtypes: float64(1), int64(5), object(1)
memory usage: 328.3+ KB

# Trend Axis [Time Intelligence] [Purchase_data] [Object - DateTime]
pd.to_datetime(purchases['purch_date']) # YYYY-MM-DD

```

0	2019-01-03
1	2019-01-03
2	2019-01-03
3	2019-01-03
4	2019-01-03
...	
5995	2019-06-20
5996	2019-06-20
5997	2019-06-20
5998	2019-06-20
5999	2019-06-20

```

purchases['purch_date'] = pd.to_datetime(purchases['purch_date'])
purchases

```

	Unnamed: 0	id	purch_date	customer_num	product_num	amount	paid
0	0	1	2019-01-03	823	27	12	568.92
1	1	2	2019-01-03	606	28	14	395.36
2	2	3	2019-01-03	955	9	17	510.17
3	3	4	2019-01-03	577	19	3	68.49
4	4	5	2019-01-03	429	8	18	759.42
...
5995	5995	5996	2019-06-20	893	33	5	411.10
5996	5996	5997	2019-06-20	566	23	11	178.97
5997	5997	5998	2019-06-20	114	19	9	205.47
5998	5998	5999	2019-06-20	404	11	20	429.40
5999	5999	6000	2019-06-20	88	57	4	274.52

6000 rows × 7 columns

```

purchases.drop(columns = 'Unnamed: 0', axis = 1 , inplace = True)
purchases.head()

  id  purch_date  customer_num  product_num  amount  paid
0   1  2019-01-03        823           27      12  568.92
1   2  2019-01-03        606           28      14  395.36
2   3  2019-01-03        955            9      17  510.17
3   4  2019-01-03        577           19       3   68.49
4   5  2019-01-03        429            8      18  759.42

purchases.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6000 entries, 0 to 5999
Data columns (total 6 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   id          6000 non-null    int64  
 1   purch_date  6000 non-null    datetime64[ns]
 2   customer_num 6000 non-null    int64  
 3   product_num  6000 non-null    int64  
 4   amount       6000 non-null    int64  
 5   paid         6000 non-null    float64 
dtypes: datetime64[ns](1), float64(1), int64(4)
memory usage: 281.4 KB

```

```

products.info()

<class 'pandas.core.frame.DataFrame'>
Index: 60 entries, 0 to 59
Data columns (total 5 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   Unnamed: 0   60 non-null    float64 
 1   id          60 non-null    float64 
 2   product     60 non-null    object  
 3   cost         60 non-null    object  
 4   company     55 non-null    object  
dtypes: float64(2), object(3)
memory usage: 2.8+ KB

products['cost'] = products['cost'].astype('str').str.replace('[,$,]', '' , regex = True).astype('float')
products.head()

  Unnamed: 0  id          product  cost  company
0           0  1  Liners - Baking Cups  6.36  Skipfire
1           1  2  Nori Sea Weed - Gold Label  85.74  Dynazzy
2           2  3  Bar Bran Honey Nut  65.40   Ntag
3           3  4  Soup - Campbells Beef Stew  68.16  Photojam
4           4  5  Wine - Shiraz Wolf Blass Premium  87.39   Eare

```

```
products.drop(columns = 'Unnamed: 0', axis = 1 , inplace = True)
products.head()
```

	id	product	cost	company
0	1	Liners - Baking Cups	6.36	Skipfire
1	2	Nori Sea Weed - Gold Label	85.74	Dynazzy
2	3	Bar Bran Honey Nut	65.40	Ntag
3	4	Soup - Campbells Beef Stew	68.16	Photojam
4	5	Wine - Shiraz Wolf Blass Premium	87.39	Eare

```
products.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 60 entries, 0 to 59
Data columns (total 4 columns):
 #   Column   Non-Null Count  Dtype  
 ---  --       --           --    
 0   id       60 non-null    float64 
 1   product  60 non-null    object  
 2   cost     60 non-null    float64 
 3   company  55 non-null    object  
dtypes: float64(2), object(2)
memory usage: 2.3+ KB
```

```
products['company'].isna().sum()
```

```
5
```

```
products[products['company'].isna()]
```

	id	product	cost	company
15	16.0	Sausage - Chorizo	55.45	NaN
21	22.0	Scotch - Queen Anne	60.26	NaN
26	27.0	Spaghetti Squash	47.41	NaN
27	28.0	Wine - Niagara,vqa Reisling	28.24	NaN
45	46.0	Aromat Spice / Seasoning	61.34	NaN

```
# Replace NaN value with Unknown , Fill Down [.ffill()] or Fill Up[.bfill()]
# products.fillna({'company' : 'Unknown'} , inplace = True)
products['company'] = products['company'].ffill()
# products['company'] = products['company'].bfill()
products
```

	id	product	cost	company
0	1	Liners - Baking Cups	6.36	Skipfire
1	2	Nori Sea Weed - Gold Label	85.74	Dynazzy

0	1	Liners - Baking Cups	6.36	Skipfire
1	2	Nori Sea Weed - Gold Label	85.74	Dynazzy
2	3	Bar Bran Honey Nut	65.40	Ntag
3	4	Soup - Campbells Beef Stew	68.16	Photojam
4	5	Wine - Shiraz Wolf Blass Premium	87.39	Eare
5	6	Wine - White, Riesling, Semi - Dry	99.22	Livepath
6	7	Brandy - Bar	13.83	Oloo
7	8	Onions - White	42.19	Oozz
8	9	Lettuce - Baby Salad Greens	30.01	Meevee

```
products.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 60 entries, 0 to 59
Data columns (total 4 columns):
 #   Column   Non-Null Count  Dtype  
--- 
 0   id       60 non-null    float64
 1   product  60 non-null    object  
 2   cost     60 non-null    float64
 3   company  60 non-null    object  
dtypes: float64(2), object(2)
memory usage: 2.3+ KB
```

```
products.groupby('company')['cost'].sum().head()
```

```
company
Aibox      56.33
Babbleopia 63.98
Brainsphere 22.83
Browsezoom  82.10
Cogibox    102.27
Name: cost, dtype: float64
```

```
company_revenue = products.groupby('company')['cost'].sum().reset_index()
company_revenue
```

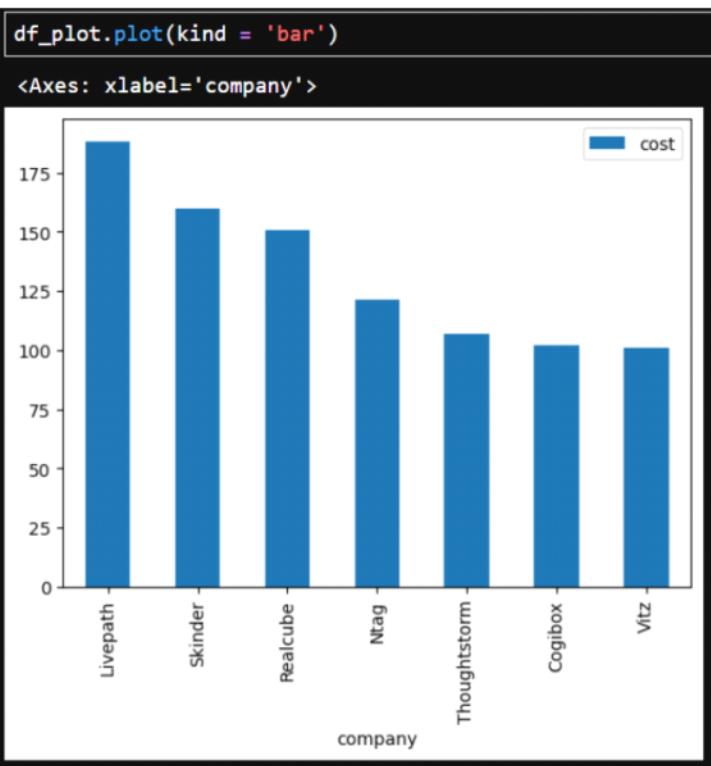
	company	cost
0	Aibox	56.33
1	Babbleopia	63.98
2	Brainsphere	22.83
3	Browsezoom	82.10
4	Cogibox	102.27
5	Digitube	8.55
6	Dynabox	99.40
7	Dynazzy	85.74
8	Eare	87.39

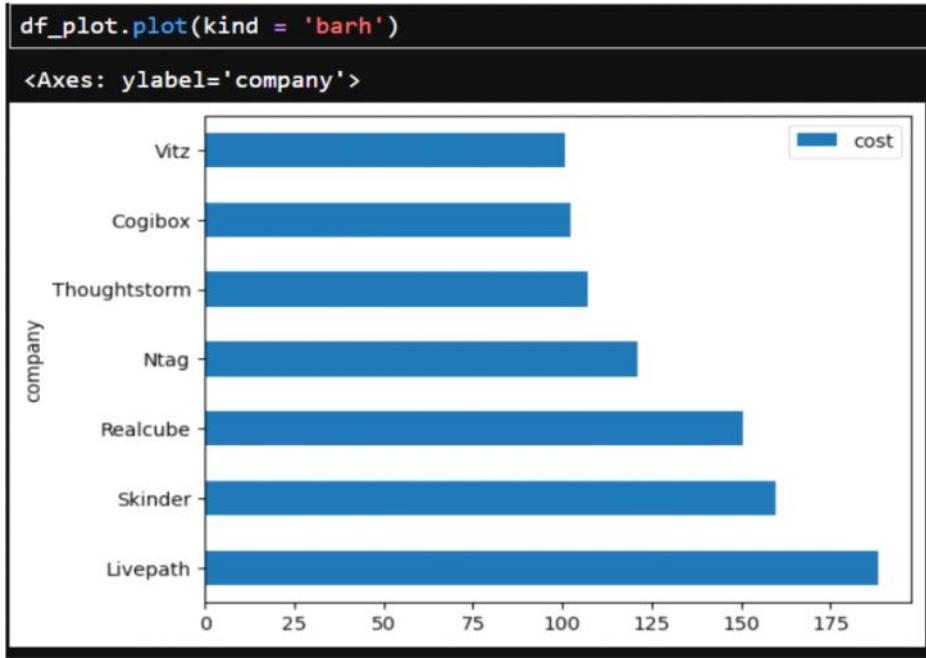
```
# Top 7 Company based on Revenue
company_revenue.sort_values(by = 'cost' , ascending = False).head(7)
```

	company	cost
19	Livepath	188.21
37	Skinder	159.85
34	Realcube	150.74
25	Ntag	121.17
40	Thoughtstorm	107.11
4	Cogibox	102.27
44	Vitz	100.84

```
df_plot = company_revenue.sort_values(by = 'cost' , ascending = False).head(7).set_index('company') # Series
df_plot
```

company	cost
Livepath	188.21
Skinder	159.85
Realcube	150.74
Ntag	121.17
Thoughtstorm	107.11
Cogibox	102.27
Vitz	100.84





```
customers.info()
```

<class 'pandas.core.frame.DataFrame'>
Index: 1000 entries, FirstCustomers to 999
Data columns (total 11 columns):
 # Column Non-Null Count Dtype
--- -- -- --
 0 id 1000 non-null float64
 1 first_name 1000 non-null object
 2 last_name 1000 non-null object
 3 email 878 non-null object
 4 gender 957 non-null object
 5 street_number 738 non-null float64
 6 street_address 963 non-null object
 7 street_suffix 963 non-null object
 8 city 921 non-null object
 9 state 920 non-null object
 10 postcode 1000 non-null int32
dtypes: float64(2), int32(1), object(8)
memory usage: 122.1+ KB

```
# 'email' -> Unique  

customers.fillna({'email' : 'Unknown'} , inplace = True)  

customers.head()
```

	id	first_name	last_name	email	gender	street_number	street_address	street_suffix	city	state
0	1	Romain	Southcott	rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road	San Diego	California
1	2	Cosimo	Molyneaux	cmolyneaux1@wiley.com	Male	NaN	NaN	NaN	El Paso	Texas
2	3	Bambi	Westrip	bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle	San Antonio	Texas
3	4	Roarke	Pankettman	rpankettman3@wiley.com	Male	74.0	Debs	Point	Memphis	Tennessee
4	5	Mikaela	Althorpe	malthorpe4@51.la	NaN	NaN	2nd	Drive	NaN	NaN

```
customers['gender'].isnull().sum()
```

43

```
customers[customers['gender'].isnull()].head()
```

	id	first_name	last_name		email	gender	street_number	street_address	street_suffix
FifthCustomers	5.0	Mikaela	Althorpe		malthorpe4@51.la	NaN	NaN	2nd	Drive
7	8.0	Tine	McSperrin	tmcsperrin7@statcounter.com		NaN	530.0	Erie	Plaza
55	56.0	Shandy	Ditzel		sditzel1j@google.com	NaN	6051.0	Nova	Circle
64	65.0	Gamaliel	Cottel		gcottel1s@merriam-webster.com	NaN	28.0	Laurel	Place
87	88.0	Baxie	Bridger		bbridger2f@1und1.de	NaN	79016.0	Havey	Circle

```
customers['gender'].mode()
```

```
0    Male  
Name: gender, dtype: object
```

```
customers['gender'].mode()[0]
```

```
'Male'
```

```
type(customers['gender'].mode()[0])
```

```
str
```

```
customers['gender'].value_counts()
```

```
gender  
Male     485  
Female   472  
Name: count, dtype: int64
```

```
# customers.fillna({'gender' : 'Male'} , inplace = True)  
customers.fillna({'gender' : customers['gender'].mode()[0]} , inplace = True)  
# customers['gender'] = customers['gender'].ffill()  
# customers['gender'] = customers['gender'].bfill()  
customers.head()
```

	id	first_name	last_name		email	gender	street_number	street_address	street_suffix	city	state
0	1	Romain	Southcott	rsouthcott0@clickbank.net	Male		1.0	Trailsway	Road	San Diego	California
1	2	Cosimo	Molyneaux	cmolyneaux1@wiley.com	Male		NaN	NaN	NaN	El Paso	Texas
2	3	Bambi	Westrip	bwestrip2@symantec.com	Female		4057.0	Arkansas	Circle	San Antonio	Texas
3	4	Roarke	Pankettman	rpankettman3@wiley.com	Male		74.0	Debs	Point	Memphis	Tennessee
4	5	Mikaela	Althorpe		malthorpe4@51.la	Male	NaN	2nd	Drive	NaN	NaN

```

customers['street_number'] = customers['street_number'].fillna(0)
customers['street_address'] = customers['street_address'].ffill()
customers['street_suffix'] = customers['street_suffix'].ffill()
customers['city'] = customers['city'].bfill()
customers['state'] = customers['state'].bfill()
customers.head()

```

name	last_name	email	gender	street_number	street_address	street_suffix	city	state	postcode
omain	Southcott	rsouthcott0@clickbank.net	Male	1.0	Trailsway	Road	San Diego	California	92127
osimo	Molyneaux	cmolyneaux1@wiley.com	Male	0.0	Trailsway	Road	El Paso	Texas	0
Bambi	Westrip	bwestrip2@symantec.com	Female	4057.0	Arkansas	Circle	San Antonio	Texas	78220
Roarke	Pankettman	rpankettman3@wiley.com	Male	74.0	Debs	Point	Memphis	Tennessee	38150
likaela	Althorpe	malthorpe4@51.la	Male	0.0	2nd	Drive	Baltimore	Maryland	83732

customers.info()

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 11 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   id          1000 non-null    int64  
 1   first_name  1000 non-null    object  
 2   last_name   1000 non-null    object  
 3   email        1000 non-null    object  
 4   gender       1000 non-null    object  
 5   street_number 1000 non-null    float64
 6   street_address 1000 non-null    object  
 7   street_suffix 1000 non-null    object  
 8   city         1000 non-null    object  
 9   state        1000 non-null    object  
 10  postcode     1000 non-null    int32  
dtypes: float64(1), int32(1), int64(1), object(8)
memory usage: 82.2+ KB

```

products.info()

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 60 entries, 0 to 59
Data columns (total 4 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   id          60 non-null    int64  
 1   product     60 non-null    object  
 2   cost         60 non-null    float64
 3   company     60 non-null    object  
dtypes: float64(1), int64(1), object(2)
memory usage: 2.0+ KB

```

```
purchases.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6000 entries, 0 to 5999
Data columns (total 6 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   id          6000 non-null    int64  
 1   purch_date  6000 non-null    datetime64[ns]
 2   customer_num 6000 non-null    int64  
 3   product_num  6000 non-null    int64  
 4   amount       6000 non-null    int64  
 5   paid         6000 non-null    float64 
dtypes: datetime64[ns](1), float64(1), int64(4)
memory usage: 281.4 KB
```

-> Filling the Missing Values

-> Handling the Missing Values() -> 'NaN', 'NaT', 'None', '<NA>'

1. Identify them , isnull(), isna(), notnull() [Boolean Return] -> Filter

```
customers[customers['gender'].isna()]
customers[customers['gender'].isnull()]
```

2. Strategies to handle the missing values

- fillna() - Object / Categorical Columns ['Unknown']
- fillna() - Numerical Columns / Continuous [0 or statistical_analysis]
- .bfill() - Backward Filling [Fill Up] [Bottom Up]
- .ffill() - Forward Filling [Fill Down] [Top To Bottom]
- .dropna() - Removing the NaN Values [axis =0(rows) | axis = 1(cols)] [Impact Analysis]
- .interpolate() - [Linear , Polynomial , degree 2] = [0 2 4 _ 8 10 _ 14 _ 18]