

# Pawan Chaudhary

## 1. Working with pandas

```
In [1]: import pandas as pd
print(f"Panda's library version: {pd.__version__}")
```

Panda's library version: 2.2.3

## 2. Using Series Data Structures

```
In [8]: # creating a list with 6 population numbers representing different city districts
# to be precise, i have used the data from 6 metropolitan cities with highest popul
# reference: https://en.wikipedia.org/wiki/List_of_cities_in_Nepal

population = [862400, 513504, 369268, 294098, 272382, 243927]
print(f"Population numbers: {population}")
```

Population numbers: [862400, 513504, 369268, 294098, 272382, 243927]

```
In [5]: # using the same reference as above, i have a list with respective district names

district = ["Kathmandu", "Pokhara", "Bharatpur", "Lalitpur", "Birgunj", "Biratnagar"]
print(f"District names: {district}")
```

District names: ['Kathmandu', 'Pokhara', 'Bharatpur', 'Lalitpur', 'Birgunj', 'Biratnagar']

```
In [ ]: # creating a pandas series and assigning the index

district_series = pd.Series(data = population, index = district, name = "Population")
print(f"{district_series}")
```

Kathmandu	862400
Pokhara	513504
Bharatpur	369268
Lalitpur	294098
Birgunj	272382
Biratnagar	243927

Name: Population Series, dtype: int64

```
In [16]: # calculating mean population across all districts
mean_population = district_series.mean()
print(f"Mean population across all districts: {mean_population}")
```

Mean population across all districts: 425929.8333333333

```
In [18]: # identifying maximum populations
max_population = district_series.idxmax()
print(f"District with maximum population: {max_population}")
```

District with maximum population: Kathmandu

```
In [20]: # identifying minimum populations
min_population = district_series.idxmin()
print(f"District with minimum population: {min_population}")
```

District with minimum population: Biratnagar

### 3. Using DataFrame Data Structures

```
In [35]: # creating a dictionary with five nepali city parks, including their area in acres
parks_data = {
    "park_name": ["Garden of Dreams", "Rani Pokhari", "Ratna Park", "Godavari Botan",
    "area_acres": [3.0, 1.5, 5.0, 82.0, 2.0],
    "annual_visitors": [250000, 200000, 500000, 150000, 100000]
}

print(f"City Parks: {parks_data}")
```

City Parks: {'park\_name': ['Garden of Dreams', 'Rani Pokhari', 'Ratna Park', 'Godavari Botanical Garden', 'Sahid Gate Park'], 'area\_acres': [3.0, 1.5, 5.0, 82.0, 2.0], 'annual\_visitors': [250000, 200000, 500000, 150000, 100000]}

```
In [43]: # converting dictionary to pandas dataframe
parks_df = pd.DataFrame(parks_data)
```

```
In [44]: # set park name as the index
parks_df = parks_df.set_index("park_name")
```

```
In [48]: print(f"Park DataFrame: \n{parks_df}")
```

Park DataFrame:

	area_acres	annual_visitors
park_name		
Garden of Dreams	3.0	250000
Rani Pokhari	1.5	200000
Ratna Park	5.0	500000
Godavari Botanical Garden	82.0	150000
Sahid Gate Park	2.0	100000

```
In [51]: # the first three entries of the dataframe
print(f"First three parks: \n {parks_df.head(3)}")
```

First three parks:

	area_acres	annual_visitors
park_name		
Garden of Dreams	3.0	250000
Rani Pokhari	1.5	200000
Ratna Park	5.0	500000

```
In [ ]: # total number of visitors of the park
total_visitors = parks_df["annual_visitors"].sum()
print(f"Total visitors of the park: {total_visitors}")
```

Total visitors of the park: 1200000

```
In [56]: # average number of visitors of the park
average_visitors = parks_df["annual_visitors"].mean()
```

```
print(f"Mean visitors of the park: {average_visitors}")
```

Mean visitors of the park: 240000.0

```
In [58]: # Largest area and its details
largest_park = parks_df["area_acres"].idxmax()
print(f"Largest park in terms of area: {largest_park}")
print(parks_df.loc[largest_park])
```

Largest park in terms of area: Godavari Botanical Garden  
 area\_acres 82.0  
 annual\_visitors 150000.0  
 Name: Godavari Botanical Garden, dtype: float64

```
In [59]: # parks with more 1,000,000 visitors
popular_parks = parks_df[parks_df["annual_visitors"] > 1_000_000]
print("\nParks with more than 1,000,000 visitors:\n", popular_parks)
```

Parks with more than 1,000,000 visitors:  
 Empty DataFrame  
 Columns: [area\_acres, annual\_visitors]  
 Index: []

```
In [ ]: # increasing the size of one of the parks by 10 areas and updating dataframe

parks_df.loc["Garden of Dreams", "area_acres"] += 10
print(f"Updated 'Garden of Dreams' area: {parks_df.loc["Garden of Dreams", "area_acres"]}")
```

Updated 'Garden of Dreams' area: 23.0

#### 4. Combining Datasets

```
In [63]: facilities_data = {
    "park_name": ["Garden of Dreams", "Rani Pokhari", "Ratna Park", "Godavari Botanical Garden", "Sahid Gate Park"],
    "playgrounds": [2, 5, 1, 3, 8],
    "sports_facilities": [4, 10, 2, 6, 12]
}

print(f"Updated Dataframe with no. of school playground and sports facilities \n {facilities_data}")
```

Updated Dataframe with no. of school playground and sports facilities  
 {'park\_name': ['Garden of Dreams', 'Rani Pokhari', 'Ratna Park', 'Godavari Botanical Garden', 'Sahid Gate Park'], 'playgrounds': [2, 5, 1, 3, 8], 'sports\_facilities': [4, 10, 2, 6, 12]}

```
In [65]: facilities_df = pd.DataFrame(facilities_data).set_index("park_name")
```

```
In [71]: # merging using index (join)

parks_merged = parks_df.join(facilities_df)
print(f"Merged parks DataFrame with facilities:\n {parks_merged}")
```

Merged parks DataFrame with facilities:

	area_acres	annual_visitors	playgrounds \
park_name			
Garden of Dreams	23.0	250000	2
Rani Pokhari	1.5	200000	5
Ratna Park	5.0	500000	1
Godavari Botanical Garden	82.0	150000	3
Sahid Gate Park	2.0	100000	8

	sports_facilities
park_name	
Garden of Dreams	4
Rani Pokhari	10
Ratna Park	2
Godavari Botanical Garden	6
Sahid Gate Park	12

## 5. Data Retrieval

```
In [74]: column_using_indexing = parks_merged["area_acres"]
print(f"Area column (using indexing operator): {column_using_indexing}")
```

Area column (using indexing operator): park\_name

Garden of Dreams	23.0
Rani Pokhari	1.5
Ratna Park	5.0
Godavari Botanical Garden	82.0
Sahid Gate Park	2.0

Name: area\_acres, dtype: float64

```
In [77]: column_using_iloc = parks_merged["annual_visitors"]
print(f"Annual visitors column (using iloc): {column_using_iloc}")
```

Annual visitors column (using iloc): park\_name

Garden of Dreams	250000
Rani Pokhari	200000
Ratna Park	500000
Godavari Botanical Garden	150000
Sahid Gate Park	100000

Name: annual\_visitors, dtype: int64

```
In [78]: column_using_loc = parks_merged["sports_facilities"]
print(f"Sports facilities column (using loc): {column_using_loc}")
```

Sports facilities column (using loc): park\_name

Garden of Dreams	4
Rani Pokhari	10
Ratna Park	2
Godavari Botanical Garden	6
Sahid Gate Park	12

Name: sports\_facilities, dtype: int64

```
In [80]: # iloc uses integer positions

rows_iloc = parks_merged.iloc[[2, 4]]
print(f"Rows 2 and 4 via iloc:\n {rows_iloc}")
```

Rows 2 and 4 via iloc:

	area_acres	annual_visitors	playgrounds	sports_facilities
park_name				
Ratna Park	5.0	500000	1	2
Sahid Gate Park	2.0	100000	8	12

In [81]: *# loc uses index labels*

```
rows_loc = parks_merged.loc[[parks_merged.index[0], parks_merged.index[1]]]
print(f"Rows 0 and 1 via loc:\n {rows_loc}")
```

Rows 0 and 1 via loc:

	area_acres	annual_visitors	playgrounds	sports_facilities
park_name				
Garden of Dreams	23.0	250000	2	4
Rani Pokhari	1.5	200000	5	10

## 6. Understanding your Data (Bonus)

In [83]: *# random seed for reproducibility*

```
import numpy as np
np.random.seed(60)
```

In [85]: `customer_ID = np.arange(1001, 1011)`  
`print(f"Customer ID: {customer_ID}")`

Customer ID: [1001 1002 1003 1004 1005 1006 1007 1008 1009 1010]

In [86]: `names = ["Aarav", "Priya", "Rahul", "Anika", "Kabir", "Sanya", "Rohan", "Isha", "Sa`  
`print(f"Customer Names: {names}")`

Customer Names: ['Aarav', 'Priya', 'Rahul', 'Anika', 'Kabir', 'Sanya', 'Rohan', 'Isha', 'Sameer', 'Naina']

In [87]: `ages = np.random.randint(18, 66, size=10)`  
`print(f"Customer ages: {ages}")`

Customer ages: [31 19 24 28 33 53 26 35 52 51]

In [88]: `total_spending = np.round(np.random.uniform(100, 5000, size=10), 2)`  
`print(f"Total spending: {total_spending}")`

Total spending: [2698.95 4662.01 3551.47 4484.18 4328.43 997.96 2691.32 4876.66 2076.42 2263.59]

In [90]: *# creating a dataa frame for these fictional customers*

```
customers_df = pd.DataFrame({
    "customer_ID" : customer_ID,
    "name" : names,
    "age" : ages,
    "total_spending" : total_spending
})

print(f"Customer DataFrame: \n{customers_df}")
```

Customer DataFrame:

	customer_ID	name	age	total_spending
0	1001	Aarav	31	2698.95
1	1002	Priya	19	4662.01
2	1003	Rahul	24	3551.47
3	1004	Anika	28	4484.18
4	1005	Kabir	33	4328.43
5	1006	Sanya	53	997.96
6	1007	Rohan	26	2691.32
7	1008	Isha	35	4876.66
8	1009	Sameer	52	2076.42
9	1010	Naina	51	2263.59

In [91]: *# calculating correlation in the dataframe*

```
customers_corr = customers_df[["age", "total_spending"]].corr()
print(f"Correlation (Age vs Total Spending):\n", customers_corr)
```

Correlation (Age vs Total Spending):

	age	total_spending
age	1.000000	-0.720228
total_spending	-0.720228	1.000000

The correlation analysis between age and total spending shows a strong negative relationship, meaning that as age increases, customers tend to spend less overall. This suggests younger customers are likely to spend more compared to older ones.