

assn1

May 22, 2023

```
[2]: #Aishwarya kelgandre Roll no.73 batch T3
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
s1 =pd.Series(range(1,10,1))
s1
```

```
[2]: 0    1
     1    2
     2    3
     3    4
     4    5
     5    6
     6    7
     7    8
     8    9
dtype: int64
```

```
[5]: s3 = pd.Series({1:21, 2:13,3:45})
s3
```

```
[5]: 1    21
     2    13
     3    45
dtype: int64
```

```
[6]: s2 = pd.Series([1, 2, 3, 4], index=['p', 'q', 'r', 's'], name='one')
s2
```

```
[6]: p    1
     q    2
     r    3
     s    4
Name: one, dtype: int64
```

```
[7]: df1 = pd.DataFrame(s2)
df1
```

```
[7]:      one
      p      1
      q      2
      r      3
      s      4
```

```
[10]: df2 = pd.read_csv("E:\\TRINITY ACADEMY OF ENGINEERING PUNE\\TE_
      ↪2022-23\\assignment\\dsbda\\csv\\housing.csv\\housing.csv")

      df2.head(10)
```

```
[10]:      longitude  latitude  housing_median_age  total_rooms  total_bedrooms
0      -122.23      37.88           41.0           880.0           129.0  \
1      -122.22      37.86           21.0          7099.0          1106.0
2      -122.24      37.85           52.0          1467.0           190.0
3      -122.25      37.85           52.0          1274.0           235.0
4      -122.25      37.85           52.0          1627.0           280.0
5      -122.25      37.85           52.0           919.0           213.0
6      -122.25      37.84           52.0          2535.0           489.0
7      -122.25      37.84           52.0          3104.0           687.0
8      -122.26      37.84           42.0          2555.0           665.0
9      -122.25      37.84           52.0          3549.0           707.0

      population  households  median_income  median_house_value  ocean_proximity
0           322.0         126.0          8.3252          452600.0          NEAR BAY
1          2401.0        1138.0          8.3014          358500.0          NEAR BAY
2           496.0         177.0          7.2574          352100.0          NEAR BAY
3           558.0         219.0          5.6431          341300.0          NEAR BAY
4           565.0         259.0          3.8462          342200.0          NEAR BAY
5           413.0         193.0          4.0368          269700.0          NEAR BAY
6          1094.0         514.0          3.6591          299200.0          NEAR BAY
7          1157.0         647.0          3.1200          241400.0          NEAR BAY
8          1206.0         595.0          2.0804          226700.0          NEAR BAY
9          1551.0         714.0          3.6912          261100.0          NEAR BAY
```

```
[11]: df2.tail(3)
```

```
[11]:      longitude  latitude  housing_median_age  total_rooms  total_bedrooms
20637      -121.22      39.43           17.0          2254.0           485.0  \
20638      -121.32      39.43           18.0          1860.0           409.0
20639      -121.24      39.37           16.0          2785.0           616.0

      population  households  median_income  median_house_value
20637          1007.0         433.0          1.7000          92300.0  \
20638           741.0         349.0          1.8672          84700.0
20639          1387.0         530.0          2.3886          89400.0
```

```

ocean_proximity
20637      INLAND
20638      INLAND
20639      INLAND

```

```
[14]: df2['median_house_value_new']=df2['median_house_value']+111
df2.tail(3)
```

```
[14]:      longitude  latitude  housing_median_age  total_rooms  total_bedrooms
20637    -121.22    39.43             17.0         2254.0           485.0 \
20638    -121.32    39.43             18.0         1860.0           409.0
20639    -121.24    39.37             16.0         2785.0           616.0
```

```

population  households  median_income  median_house_value
20637      1007.0        433.0         1.7000           92300.0 \
20638       741.0        349.0         1.8672           84700.0
20639      1387.0        530.0         2.3886           89400.0

```

```

ocean_proximity  median_house_value_new
20637      INLAND           92411.0
20638      INLAND           84811.0
20639      INLAND           89511.0

```

```
[15]: df2.to_json('data1.json')
```

```
[16]: len(df2['total_rooms'])
```

```
[16]: 20640
```

```
[17]: df2['total_rooms'].count()
```

```
[17]: 20640
```

```
[18]: df2['total_rooms'].mean()
```

```
[18]: 2635.7630813953488
```

```
[19]: df2['total_rooms'].describe()
```

```
[19]: count      20640.000000
mean         2635.763081
std          2181.615252
min           2.000000
25%          1447.750000
50%          2127.000000
75%          3148.000000
max          39320.000000
Name: total_rooms, dtype: float64
```

```
[20]: df2['total_rooms'].cumsum()
```

```
[20]: 0          880.0
      1       7979.0
      2       9446.0
      3      10720.0
      4      12347.0
      ...
      20635  54394554.0
      20636  54395251.0
      20637  54397505.0
      20638  54399365.0
      20639  54402150.0
      Name: total_rooms, Length: 20640, dtype: float64
```

```
[23]: df2.describe()
```

```
[23]:
```

	longitude	latitude	housing_median_age	total_rooms	
count	20640.000000	20640.000000	20640.000000	20640.000000	\
mean	-119.569704	35.631861	28.639486	2635.763081	
std	2.003532	2.135952	12.585558	2181.615252	
min	-124.350000	32.540000	1.000000	2.000000	
25%	-121.800000	33.930000	18.000000	1447.750000	
50%	-118.490000	34.260000	29.000000	2127.000000	
75%	-118.010000	37.710000	37.000000	3148.000000	
max	-114.310000	41.950000	52.000000	39320.000000	

	total_bedrooms	population	households	median_income	
count	20433.000000	20640.000000	20640.000000	20640.000000	\
mean	537.870553	1425.476744	499.539680	3.870671	
std	421.385070	1132.462122	382.329753	1.899822	
min	1.000000	3.000000	1.000000	0.499900	
25%	296.000000	787.000000	280.000000	2.563400	
50%	435.000000	1166.000000	409.000000	3.534800	
75%	647.000000	1725.000000	605.000000	4.743250	
max	6445.000000	35682.000000	6082.000000	15.000100	

	median_house_value	median_house_value_new
count	20640.000000	20640.000000
mean	206855.816909	206966.816909
std	115395.615874	115395.615874
min	14999.000000	15110.000000
25%	119600.000000	119711.000000
50%	179700.000000	179811.000000
75%	264725.000000	264836.000000
max	500001.000000	500112.000000

```
[26]: df = pd.read_csv("E:\\TRINITY ACADEMY OF ENGINEERING PUNE\\TE_
↪2022-23\\assignment\\dsbda\\csv\\housing.csv\\housing.csv")
df.columns
```

```
[26]: Index(['longitude', 'latitude', 'housing_median_age', 'total_rooms',
            'total_bedrooms', 'population', 'households', 'median_income',
            'median_house_value', 'ocean_proximity'],
            dtype='object')
```

```
[27]: df.longitude
```

```
[27]: 0      -122.23
      1      -122.22
      2      -122.24
      3      -122.25
      4      -122.25
      ...
20635  -121.09
20636  -121.21
20637  -121.22
20638  -121.32
20639  -121.24
Name: longitude, Length: 20640, dtype: float64
```

```
[28]: df.iloc[:,1:3]
```

```
[28]:      latitude  housing_median_age
0         37.88             41.0
1         37.86             21.0
2         37.85             52.0
3         37.85             52.0
4         37.85             52.0
...         ...                 ...
20635     39.48             25.0
20636     39.49             18.0
20637     39.43             17.0
20638     39.43             18.0
20639     39.37             16.0
```

```
[20640 rows x 2 columns]
```

```
[33]: import pandas as pd

data = pd.read_csv("Downloads\\attendance_19-11-22.csv")
data.head(10)
```

```
[33]:
```

	Roll no	PRN no	name	attended lectures	lectures
0	1	72157089F	AJAY ANIL GAIKWAD	5	19 \
1	2	72157090K	APRE OMKAR DILIP	9	19
2	3	72157091H	AWACHAR SNEHAL SUHAS	9	19
3	4	72157092F	BADE SHRIRAM ANIL	10	19
4	5	72157094B	BHOJANE APOORVA VASANT	10	19
5	6	72157095L	BHUTADA UTKARSH HIMANSHU	10	19
6	7	72157096J	BIJJARGI MAHESH PANDIT	10	19
7	8	72157097G	BOBADE ROHAN RAJESH	10	19
8	9	72157098E	BORATE SHWETA AMBADAS	10	19
9	10	72157099C	CHAVAN ATHARVA SANJAY	10	19

```
percentage
```

0	26.3158
1	47.3684
2	47.3684
3	52.6316
4	52.6316
5	52.6316
6	52.6316
7	52.6316
8	52.6316
9	52.6316

```
[34]: data.describe()
```

```
[34]:
```

	Roll no	attended lectures	lectures	percentage
count	79.000000	79.000000	79.0	79.000000
mean	40.000000	9.898734	19.0	52.098620
std	22.949219	0.590414	0.0	3.107443
min	1.000000	5.000000	19.0	26.315800
25%	20.500000	10.000000	19.0	52.631600
50%	40.000000	10.000000	19.0	52.631600
75%	59.500000	10.000000	19.0	52.631600
max	79.000000	10.000000	19.0	52.631600

```
[35]: data.isnull()
```

```
[35]:
```

	Roll no	PRN no	name	attended lectures	lectures	percentage
0	False	False	False	False	False	False
1	False	False	False	False	False	False
2	False	False	False	False	False	False
3	False	False	False	False	False	False
4	False	False	False	False	False	False
..
74	False	False	False	False	False	False
75	False	False	False	False	False	False

76	False	False	False	False	False	False
77	False	False	False	False	False	False
78	False	False	False	False	False	False

[79 rows x 6 columns]

```
[36]: data.isnull().sum()
```

```
[36]: Roll no          0
      PRN no          0
      name            0
      attended lectures 0
      lectures        0
      percentage      0
      dtype: int64
```

```
[53]: data["Roll no"].fillna("Roll no", inplace = False)
      data.isnull().sum()
```

```
[53]: Roll no          0
      PRN no          0
      name            0
      attended lectures 0
      lectures        0
      percentage      0
      dtype: int64
```

```
[46]: import numpy as np
      data.replace(to_replace = np.nan, value = -99)
```

```
[46]:
```

	Roll no	PRN no	name	attended lectures	lectures
0	1	72157089F	AJAY ANIL GAIKWAD	5	19 \
1	2	72157090K	APRE OMKAR DILIP	9	19
2	3	72157091H	AWACHAR SNEHAL SUHAS	9	19
3	4	72157092F	BADE SHRIRAM ANIL	10	19
4	5	72157094B	BHOJANE APOORVA VASANT	10	19
..
74	75	72214538B	KURHADE PRITI RAJENDRA	10	19
75	76	72214544G	PATIL MAMTA DINKAR	10	19
76	77	72214545E	SATALKAR GAURI NAVNATH	10	19
77	78	72214546C	SHIRASKAR ISHA RAHUL	10	19
78	79	72214539L	ALEENA HANIF BAGWAN	9	19

	percentage
0	26.3158
1	47.3684
2	47.3684

```

3      52.6316
4      52.6316
..      ...
74     52.6316
75     52.6316
76     52.6316
77     52.6316
78     47.3684

```

[79 rows x 6 columns]

```
[47]: data.fillna(method='pad')
```

```

[47]:   Roll no    PRN no          name  attended lectures  lectures
0         1  72157089F    AJAY ANIL GAIKWAD             5         19 \
1         2  72157090K    APRE OMKAR DILIP             9         19
2         3  72157091H    AWACHAR SNEHAL SUHAS           9         19
3         4  72157092F    BADE SHRIRAM ANIL            10         19
4         5  72157094B    BHOJANE APOORVA VASANT        10         19
..      ...      ...      ...      ...      ...
74        75  72214538B    KURHADE PRITI RAJENDRA        10         19
75        76  72214544G    PATIL MAMTA DINKAR           10         19
76        77  72214545E    SATALKAR GAURI NAVNATH        10         19
77        78  72214546C    SHIRASKAR ISHA RAHUL         10         19
78        79  72214539L    ALEENA HANIF BAGWAN          9         19

```

```

percentage
0      26.3158
1      47.3684
2      47.3684
3      52.6316
4      52.6316
..      ...
74     52.6316
75     52.6316
76     52.6316
77     52.6316
78     47.3684

```

[79 rows x 6 columns]

```
[54]: data['Roll no'].fillna(int(data['Roll no'].mean()), inplace=False)
```

```

[54]: 0      1
      1      2
      2      3
      3      4

```



```

4      5
      ..
74     75
75     76
76     77
77     78
78     79
Name: Roll no, Length: 79, dtype: int64

```

```
[51]: data.dropna(axis=1)
```

```
[51]:
```

	Roll no	PRN no	name	attended lectures	lectures
0	1	72157089F	AJAY ANIL GAIKWAD	5	19 \
1	2	72157090K	APRE OMKAR DILIP	9	19
2	3	72157091H	AWACHAR SNEHAL SUHAS	9	19
3	4	72157092F	BADE SHRIRAM ANIL	10	19
4	5	72157094B	BHOJANE APOORVA VASANT	10	19
..
74	75	72214538B	KURHADE PRITI RAJENDRA	10	19
75	76	72214544G	PATIL MAMTA DINKAR	10	19
76	77	72214545E	SATALKAR GAURI NAVNATH	10	19
77	78	72214546C	SHIRASKAR ISHA RAHUL	10	19
78	79	72214539L	ALEENA HANIF BAGWAN	9	19

```

percentage
0      26.3158
1      47.3684
2      47.3684
3      52.6316
4      52.6316
..      ...
74     52.6316
75     52.6316
76     52.6316
77     52.6316
78     47.3684

```

```
[79 rows x 6 columns]
```

```
[55]: import pandas as pd

df = pd.DataFrame({"A": [12, 4, 5, None, 1],
                    "B": [None, 2, 54, 3, None],
                    "C": [20, 16, None, 3, 8],
                    "D": [14, 3, None, None, 6]})

df
```

```
[55]:
```

	A	B	C	D
0	12.0	NaN	20.0	14.0
1	4.0	2.0	16.0	3.0
2	5.0	54.0	NaN	NaN
3	NaN	3.0	3.0	NaN
4	1.0	NaN	8.0	6.0

```
[56]: df.interpolate(method='linear', limit_direction='forward')
```

```
[56]:
```

	A	B	C	D
0	12.0	NaN	20.0	14.0
1	4.0	2.0	16.0	3.0
2	5.0	54.0	9.5	4.0
3	3.0	3.0	3.0	5.0
4	1.0	3.0	8.0	6.0

```
[58]: text="today is Monday"
text.replace(' ', '')
```

```
[58]: 'todayisMonday'
```

```
[65]: text=' Today'
text.lstrip()
```

```
[65]: 'Today'
```

```
[60]: text='Today '
text.rstrip()
```

```
[60]: 'Today'
```

```
[66]: text=' Today '
text.strip()
```

```
[66]: 'Today'
```

```
[70]: import pandas
import scipy
import numpy
from sklearn.preprocessing import MinMaxScaler

X = [ [110, 200], [120, 800], [310, 400], [140, 900], [510, 200], [653, 400]
↪ , [310, 880]
] ]

scaler = MinMaxScaler(feature_range=(0,5))
rescaledX = scaler.fit_transform(X)
```

x

[70]: [[110, 200],
[120, 800],
[310, 400],
[140, 900],
[510, 200],
[653, 400],
[310, 880]]

[]: