

1011_heatmap

February 7, 2023

1 Heat Map

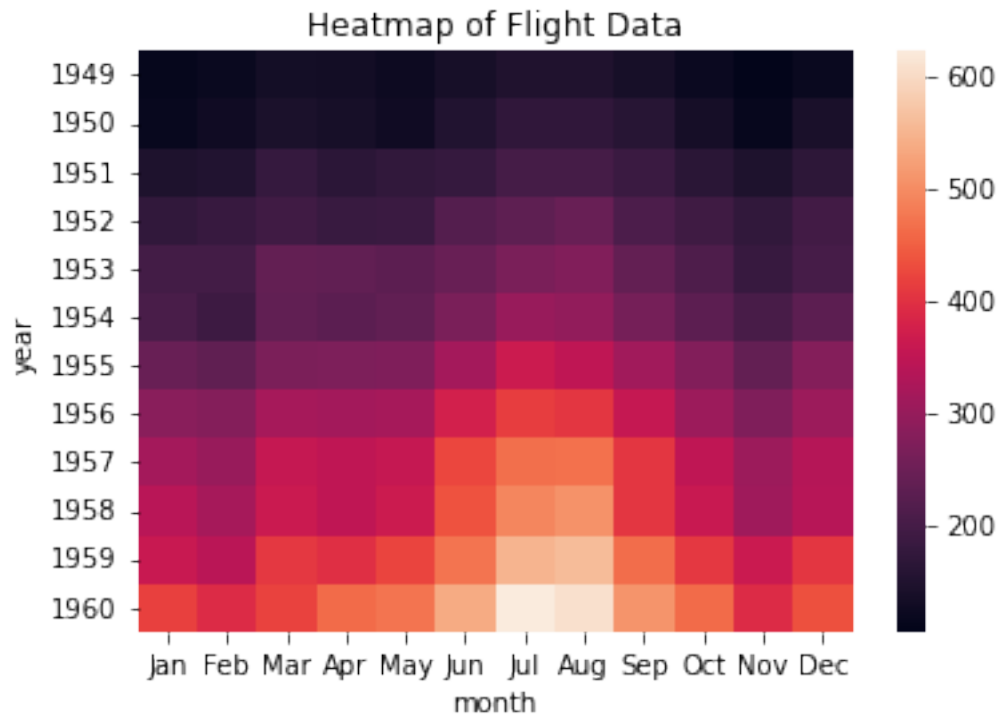
- 1011
- github: <https://github.com/Maruf2309>
- Linkedin: <https://www.linkedin.com/in/maruf-hossain-682213150/>
- Kaggle : <https://www.kaggle.com/marufnthewindows>

```
[15]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
import sklearn
```

```
[2]: flight_data = sns.load_dataset('flights') # Built in Dataset loaded
flight_data.head()
```

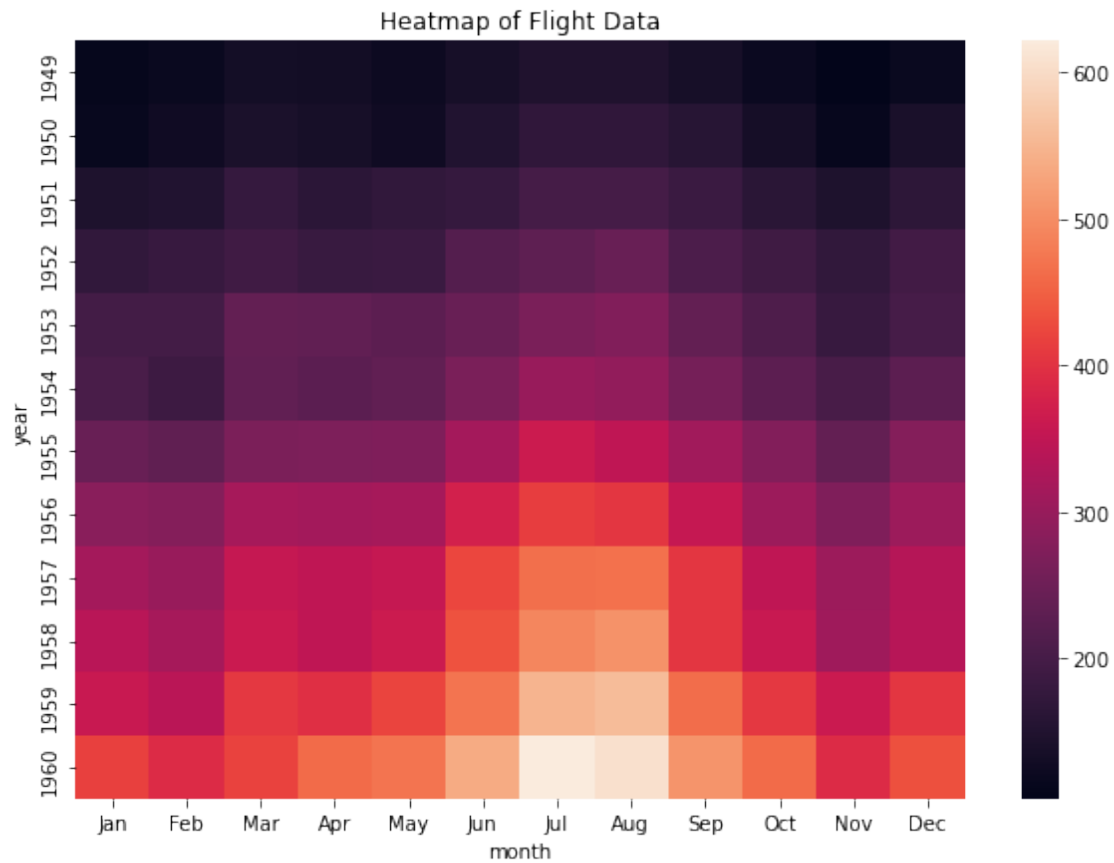
```
[2]:   year month  passengers
0  1949   Jan         112
1  1949   Feb         118
2  1949   Mar         132
3  1949   Apr         129
4  1949   May         121
```

```
[3]: flights = flight_data.pivot("year", "month", "passengers") # Left: Year, x =
    ↪Month, right: Passengers
ax = sns.heatmap(flights)
plt.title("Heatmap of Flight Data")
plt.show()
```

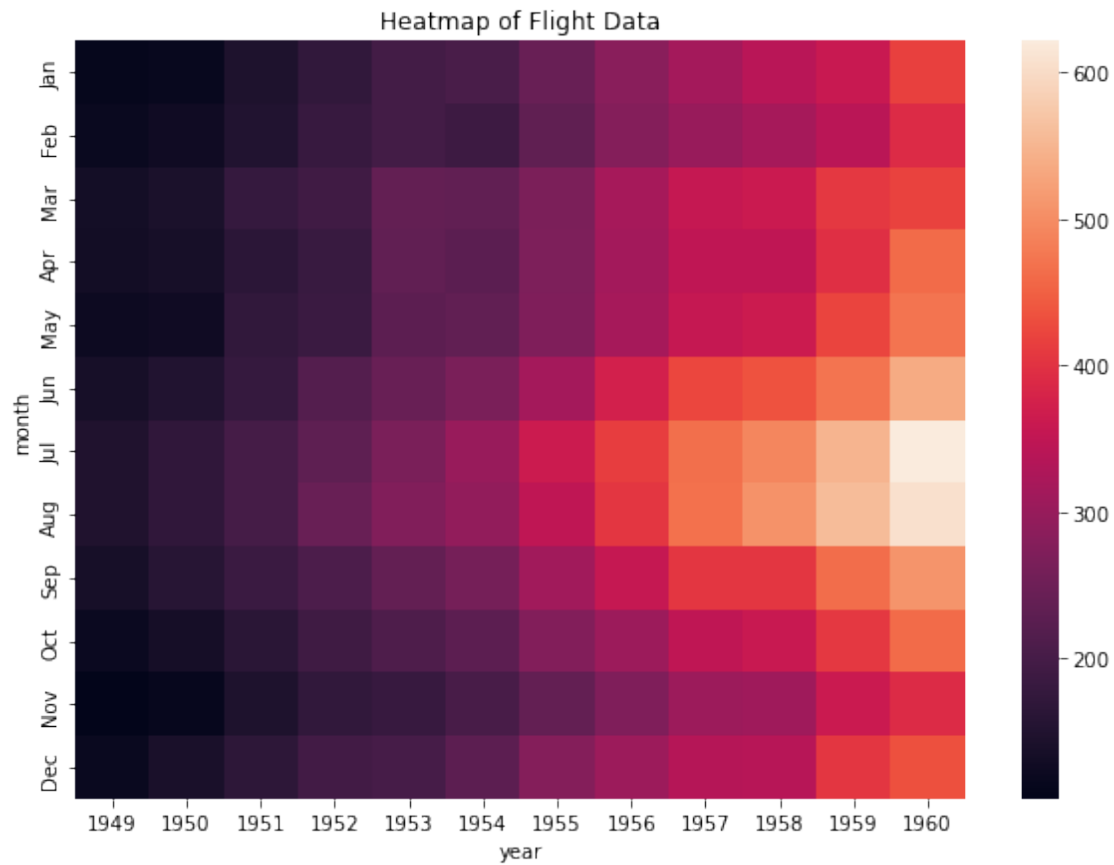


- Lighter shade indicate higher value & lower shade indicate lower value

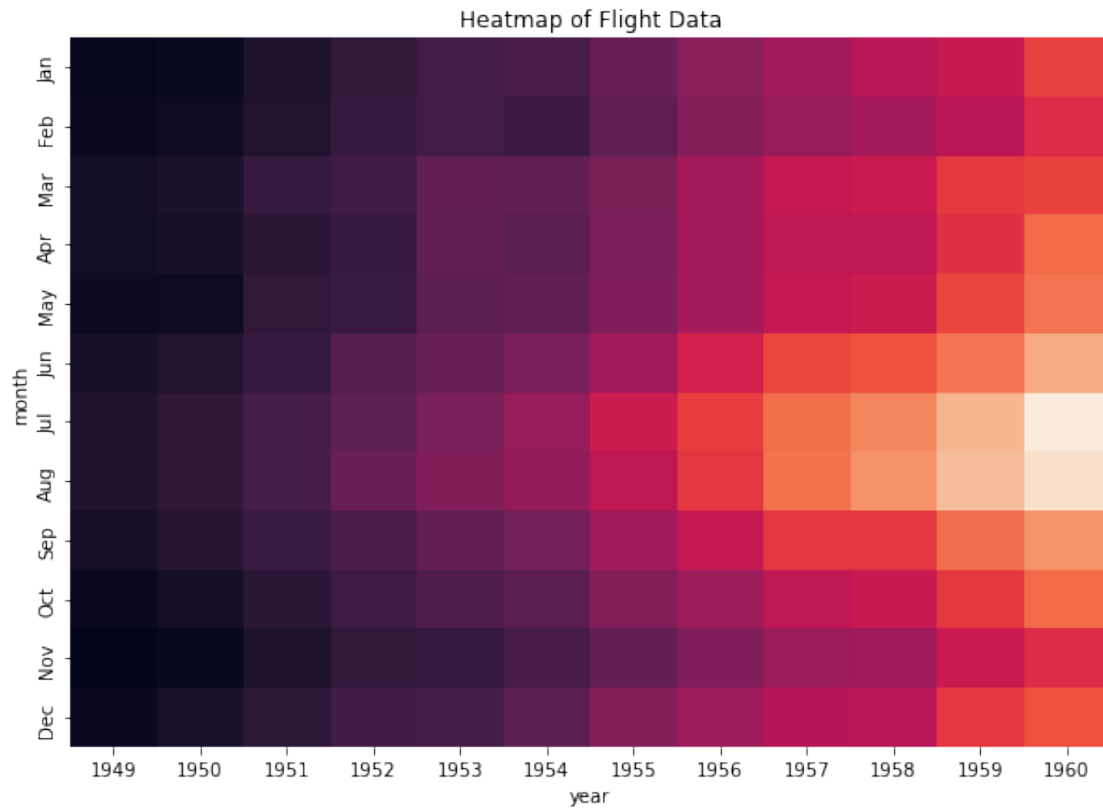
```
[4]: # Size Customization
flights = flight_data.pivot("year", "month", "passengers") # Left: Year, x = Month, right: Passengers
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights)
plt.title("Heatmap of Flight Data")
plt.show()
```



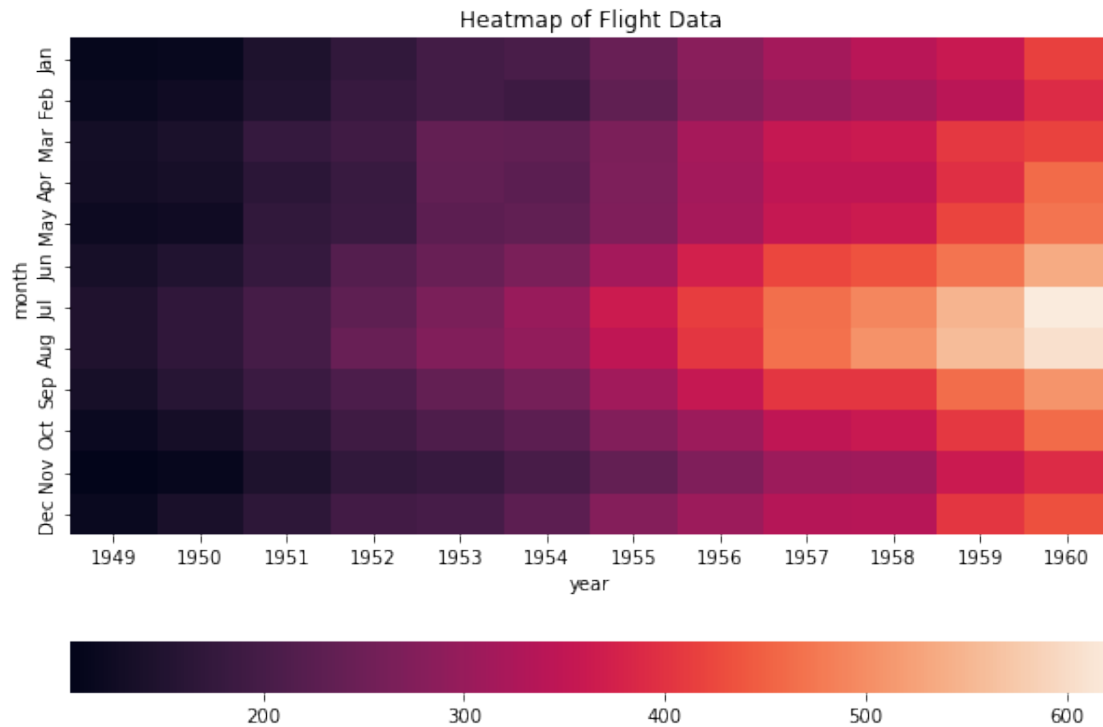
```
[5]: # Altering Position
flights = flight_data.pivot( "month", "year", "passengers") # Altering Position
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights)
plt.title("Heatmap of Flight Data")
plt.show()
```



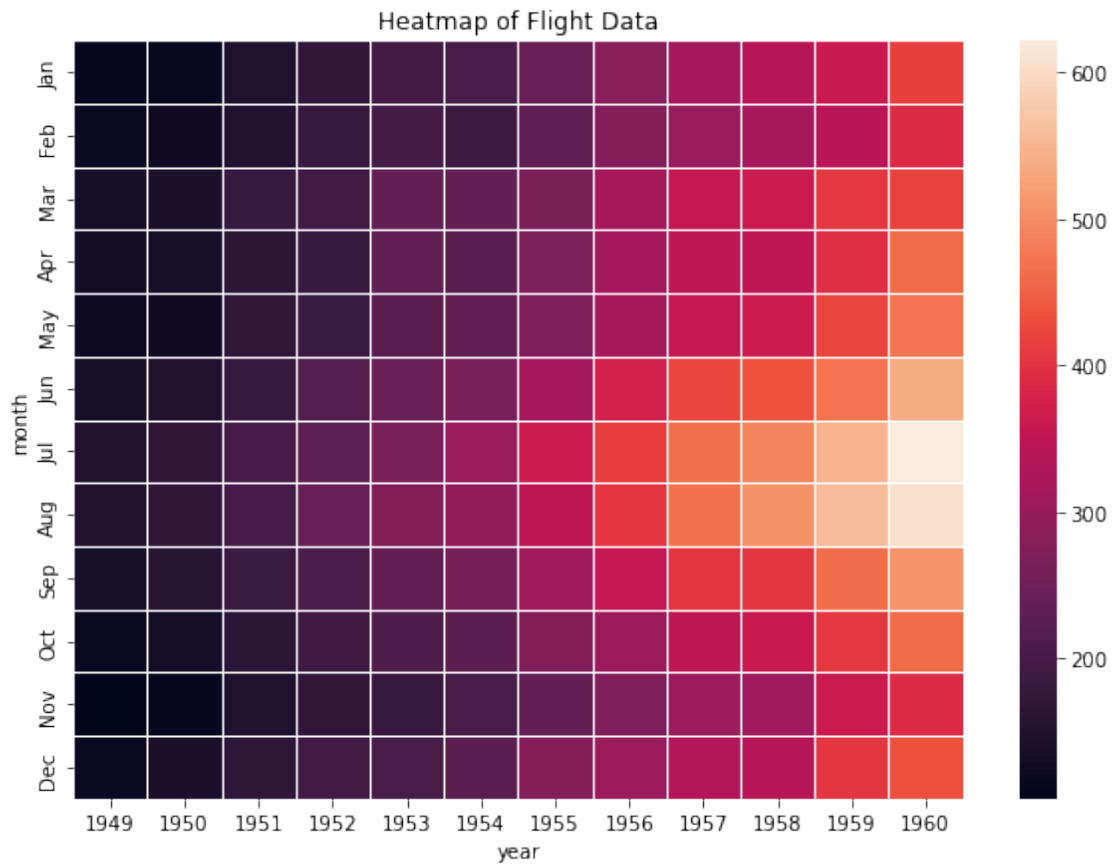
```
[6]: flights = flight_data.pivot( "month", "year", "passengers")
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights, cbar=False)           # Remove Color Bar
plt.title("Heatmap of Flight Data")
plt.show()
```



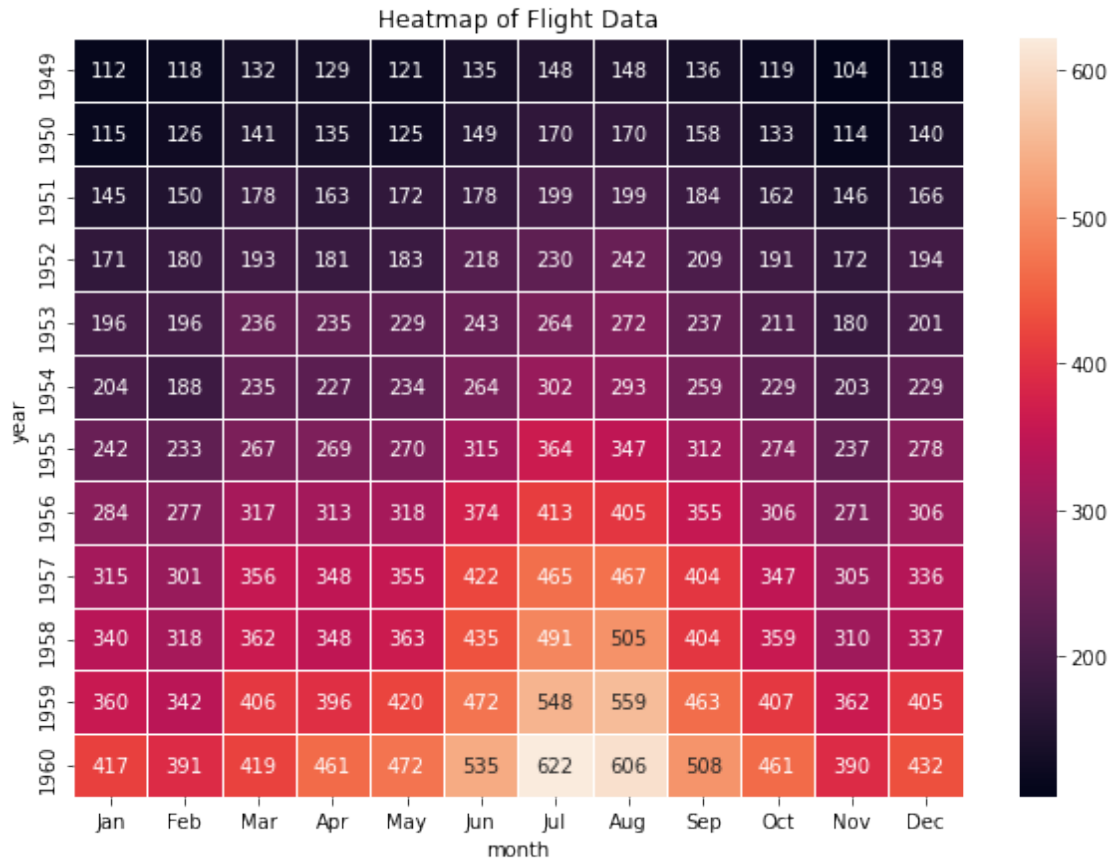
```
[7]: flights = flight_data.pivot( "month", "year", "passengers")
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights, cbar_kws={"orientation":"horizontal"})    # Change color bar position
plt.title("Heatmap of Flight Data")
plt.show()
```



```
[8]: flights = flight_data.pivot( "month", "year", "passengers")
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights, linewidths=0.5)    # Linewidth
plt.title("Heatmap of Flight Data")
plt.show()
```



```
[9]: flights = flight_data.pivot( "year", "month", "passengers")
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights, linewidths=0.5, annot=True, fmt="d")    # format
plt.title("Heatmap of Flight Data")
plt.show()
```

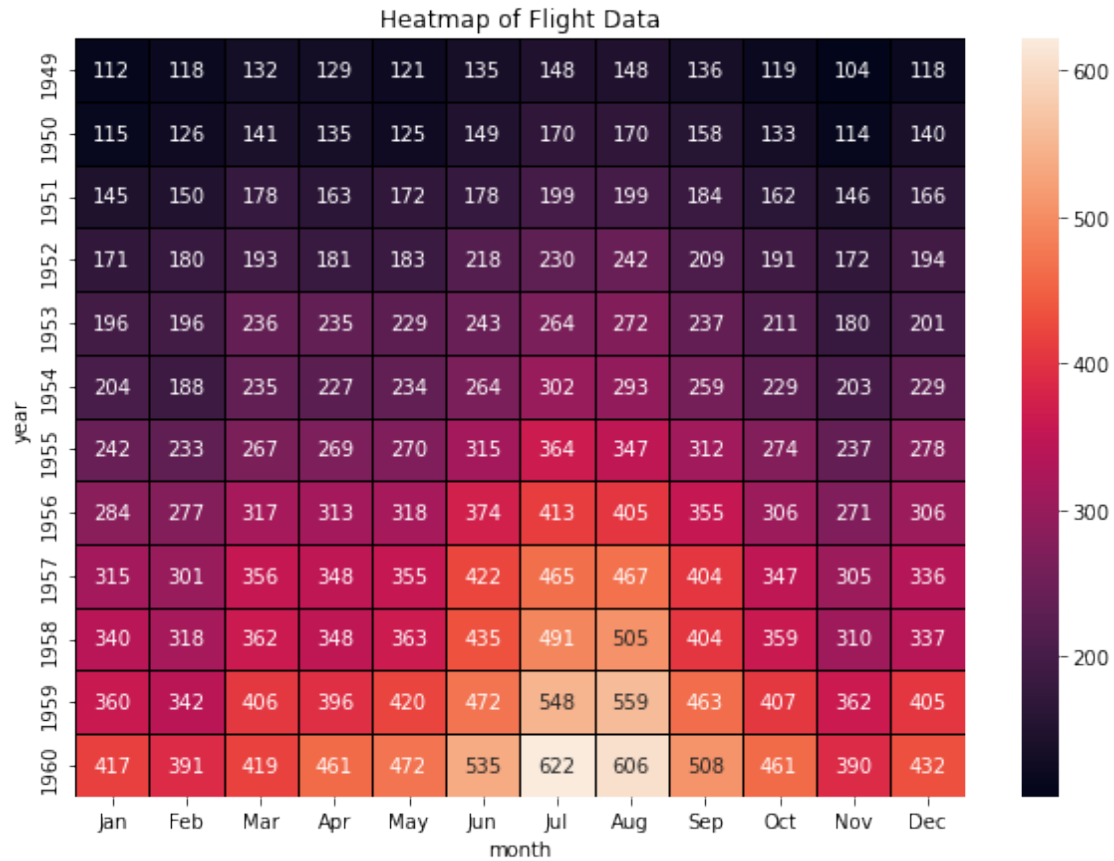


```
[10]: flight_data.describe()
```

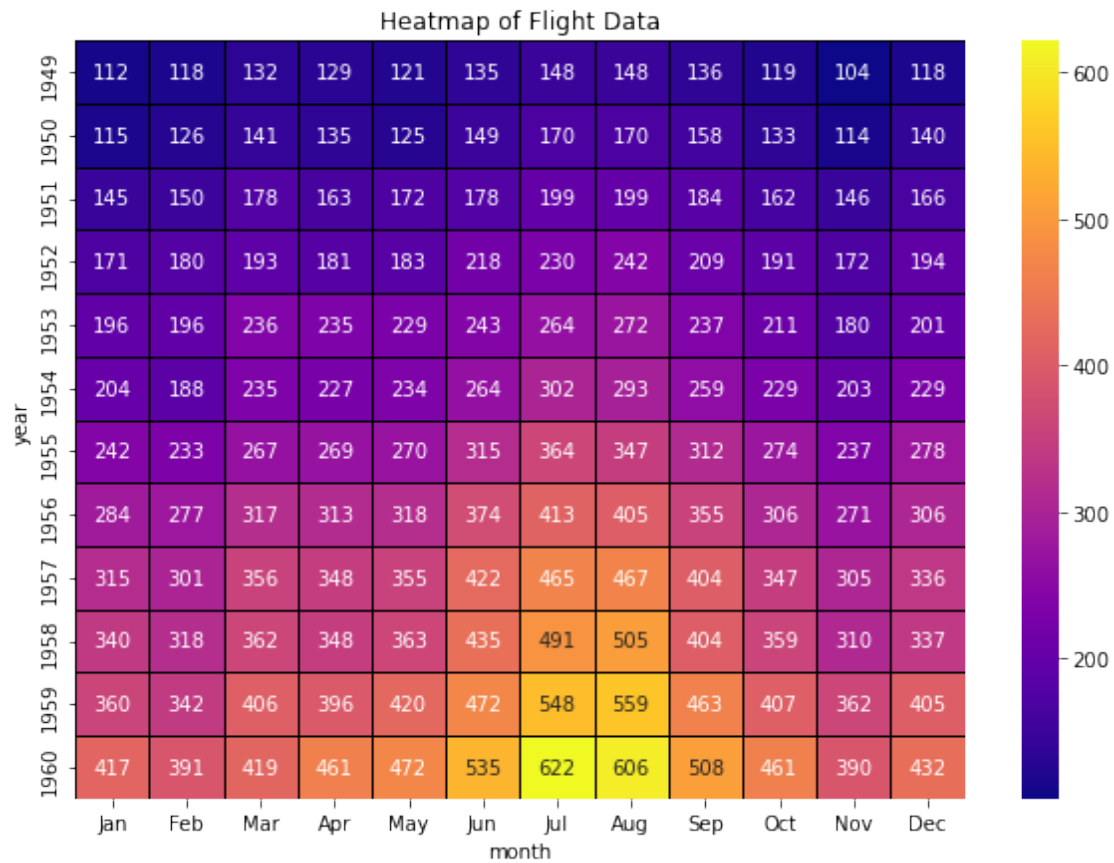
```
[10]:
```

	year	passengers
count	144.000000	144.000000
mean	1954.500000	280.298611
std	3.464102	119.966317
min	1949.000000	104.000000
25%	1951.750000	180.000000
50%	1954.500000	265.500000
75%	1957.250000	360.500000
max	1960.000000	622.000000

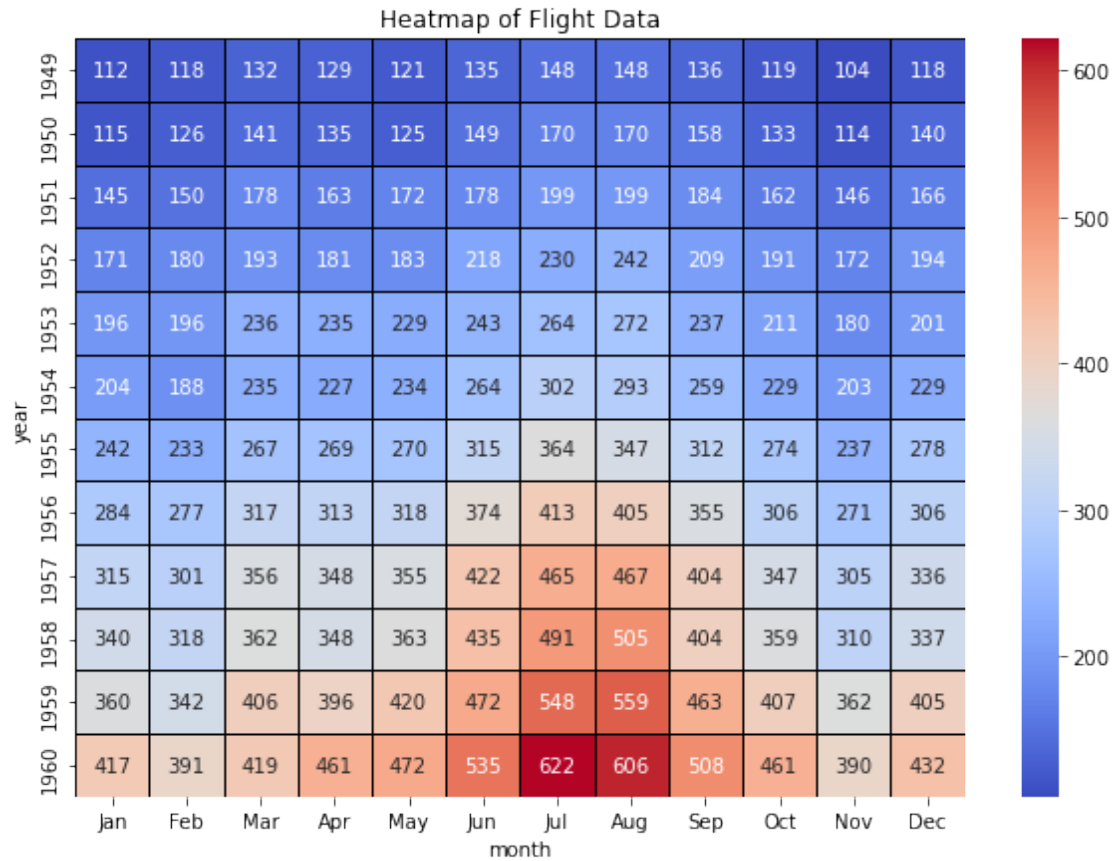
```
[11]: flights = flight_data.pivot( "year", "month", "passengers")
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights, linewidths=0.5, linecolor='black', annot=True,
    fmt="d") # format
plt.title("Heatmap of Flight Data")
plt.show()
```

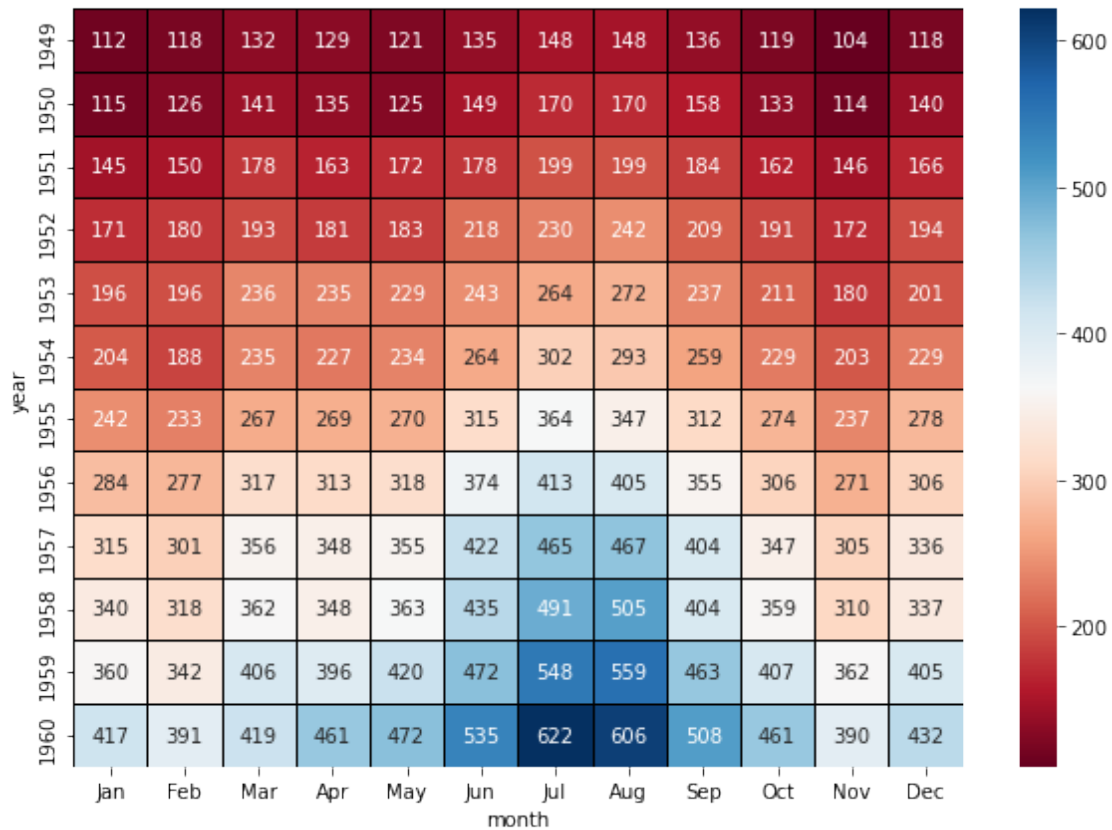
```
[12]: flights = flight_data.pivot( "year", "month", "passengers")
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights, linewidths=0.5, linecolor='black', annot=True,
    fmt="d", cmap='plasma') # Change heat map color by color map/ cmap
plt.title("Heatmap of Flight Data")
plt.show()
```



```
[13]: flights = flight_data.pivot( "year", "month", "passengers")
plt.figure(figsize=(10,7))
ax = sns.heatmap(flights, linewidths=0.5, linecolor='black', annot=True,
    ↪fmt="d", cmap='coolwarm') # Change heat map color by color map/ cmap
plt.title("Heatmap of Flight Data")
plt.show()
```



```
[14]: flight_data = flight_data.pivot('year', 'month', 'passengers')
plt.figure(figsize=(10,7))
ax = sns.heatmap(flight_data, linewidths=0.5, linecolor='black', annot=True,
    ↪fmt="d", cmap='RdBu') # more map >> cmap='magma'
```



2 Correlation Matrix

```
[17]: from sklearn.datasets import fetch_california_housing
housing_data = fetch_california_housing()
```

```
[18]: df = pd.DataFrame(housing_data.data)
df.columns = housing_data.feature_names
df['price'] = housing_data.target
df.head()
```

```
[18]: MedInc  HouseAge  AveRooms  AveBedrms  Population  AveOccup  Latitude  \
0  8.3252    41.0    6.984127  1.023810    322.0    2.555556    37.88
1  8.3014    21.0    6.238137  0.971880    2401.0    2.109842    37.86
2  7.2574    52.0    8.288136  1.073446    496.0    2.802260    37.85
3  5.6431    52.0    5.817352  1.073059    558.0    2.547945    37.85
4  3.8462    52.0    6.281853  1.081081    565.0    2.181467    37.85

Longitude  price
0  -122.23  4.526
1  -122.22  3.585
```

```
2    -122.24  3.521
3    -122.25  3.413
4    -122.25  3.422
```

```
[21]: print(housing_data.DESCR) # Know About the Data or Data description
```

```
.. _california_housing_dataset:
```

```
California Housing dataset
```

```
-----
```

```
**Data Set Characteristics:**
```

```
:Number of Instances: 20640
```

```
:Number of Attributes: 8 numeric, predictive attributes and the target
```

```
:Attribute Information:
```

```
- MedInc           median income in block group
- HouseAge         median house age in block group
- AveRooms         average number of rooms per household
- AveBedrms        average number of bedrooms per household
- Population       block group population
- AveOccup         average number of household members
- Latitude         block group latitude
- Longitude        block group longitude
```

```
:Missing Attribute Values: None
```

This dataset was obtained from the StatLib repository.

https://www.dcc.fc.up.pt/~ltorgo/Regression/cal_housing.html

The target variable is the median house value for California districts, expressed in hundreds of thousands of dollars (\$100,000).

This dataset was derived from the 1990 U.S. census, using one row per census block group. A block group is the smallest geographical unit for which the U.S. Census Bureau publishes sample data (a block group typically has a population of 600 to 3,000 people).

An household is a group of people residing within a home. Since the average number of rooms and bedrooms in this dataset are provided per household, these columns may take surprisingly large values for block groups with few households and many empty houses, such as vacation resorts.

It can be downloaded/loaded using the

:func:`sklearn.datasets.fetch_california_housing` function.

.. topic:: References

- Pace, R. Kelley and Ronald Barry, Sparse Spatial Autoregressions, Statistics and Probability Letters, 33 (1997) 291-297

```
[24]: df.corr() # Data Correlation
```

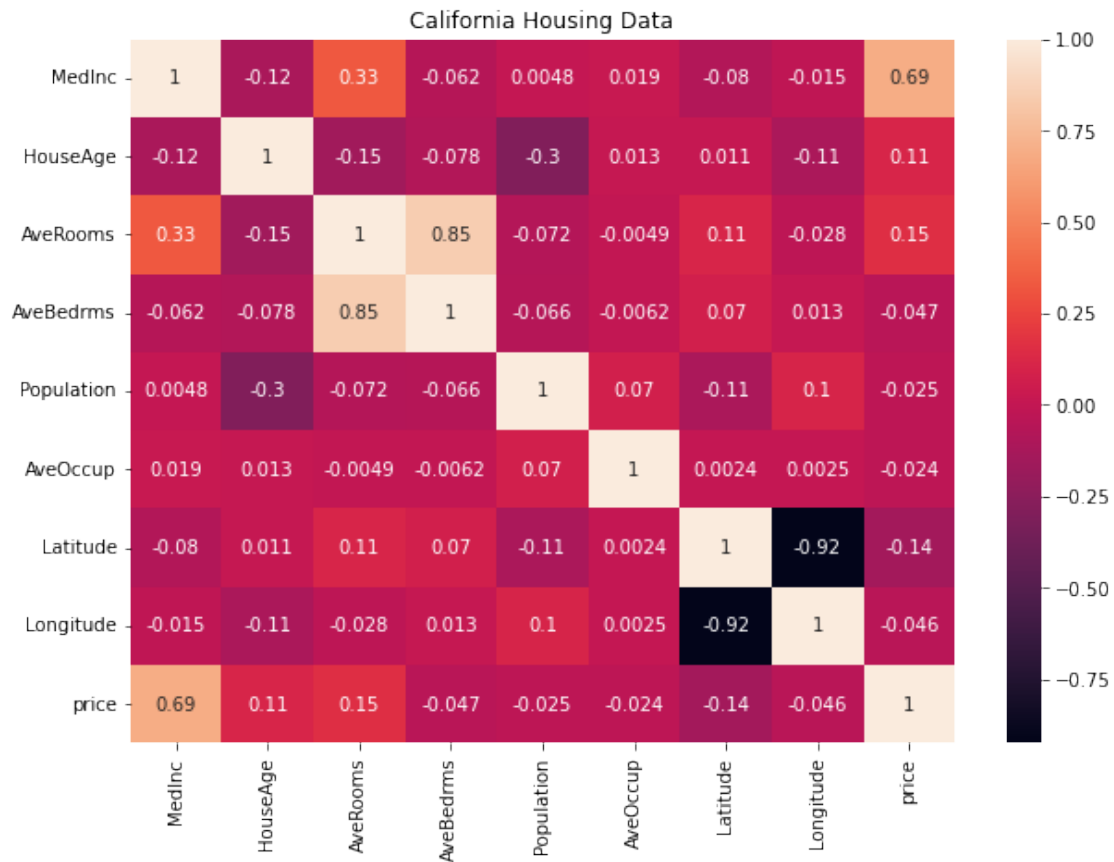
```
[24]:
```

	MedInc	HouseAge	AveRooms	AveBedrms	Population	AveOccup	\
MedInc	1.000000	-0.119034	0.326895	-0.062040	0.004834	0.018766	
HouseAge	-0.119034	1.000000	-0.153277	-0.077747	-0.296244	0.013191	
AveRooms	0.326895	-0.153277	1.000000	0.847621	-0.072213	-0.004852	
AveBedrms	-0.062040	-0.077747	0.847621	1.000000	-0.066197	-0.006181	
Population	0.004834	-0.296244	-0.072213	-0.066197	1.000000	0.069863	
AveOccup	0.018766	0.013191	-0.004852	-0.006181	0.069863	1.000000	
Latitude	-0.079809	0.011173	0.106389	0.069721	-0.108785	0.002366	
Longitude	-0.015176	-0.108197	-0.027540	0.013344	0.099773	0.002476	
price	0.688075	0.105623	0.151948	-0.046701	-0.024650	-0.023737	

	Latitude	Longitude	price
MedInc	-0.079809	-0.015176	0.688075
HouseAge	0.011173	-0.108197	0.105623
AveRooms	0.106389	-0.027540	0.151948
AveBedrms	0.069721	0.013344	-0.046701
Population	-0.108785	0.099773	-0.024650
AveOccup	0.002366	0.002476	-0.023737
Latitude	1.000000	-0.924664	-0.144160
Longitude	-0.924664	1.000000	-0.045967
price	-0.144160	-0.045967	1.000000

```
[28]: plt.figure(figsize=(10,7))
ax = sns.heatmap(df.corr(), annot=True) # 0.85 is most correlated values;
↳whereas 0-92 is leaset corraled values
plt.title('California Housing Data')
```

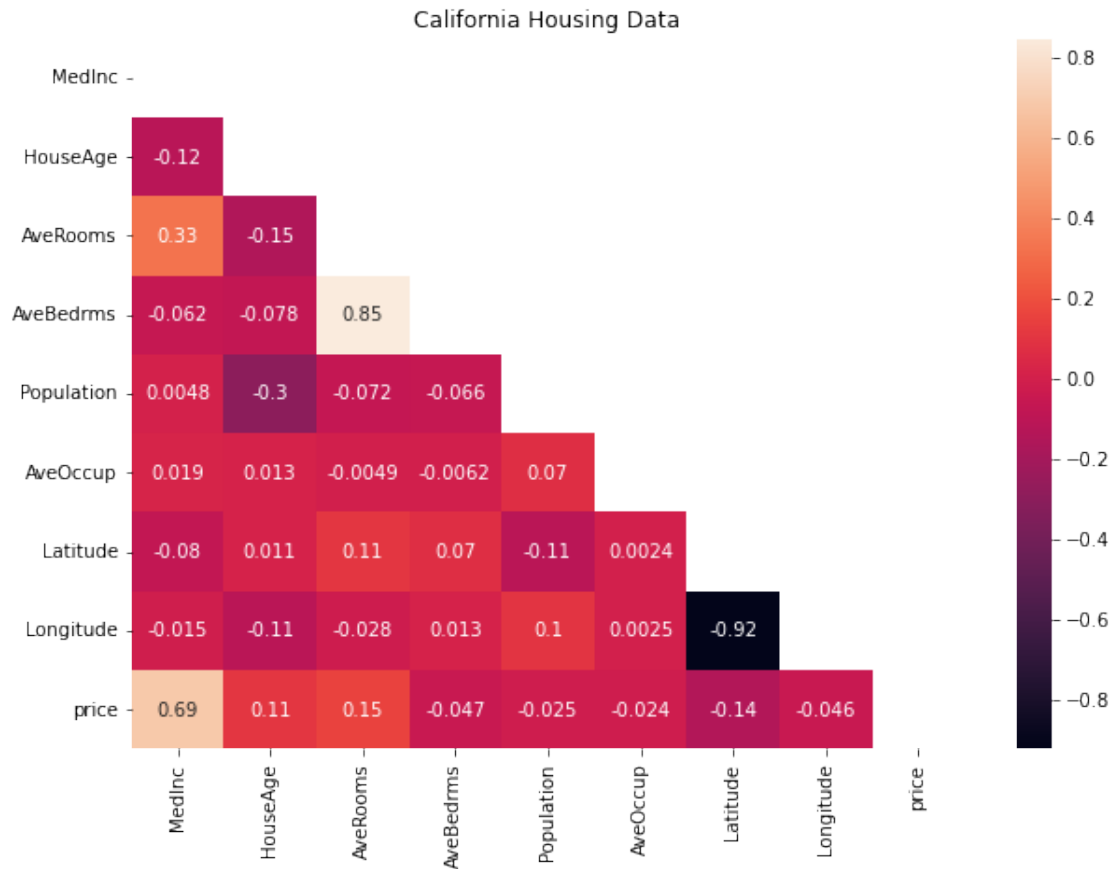
```
[28]: Text(0.5, 1.0, 'California Housing Data')
```



2.0.1 Lower half Correlation

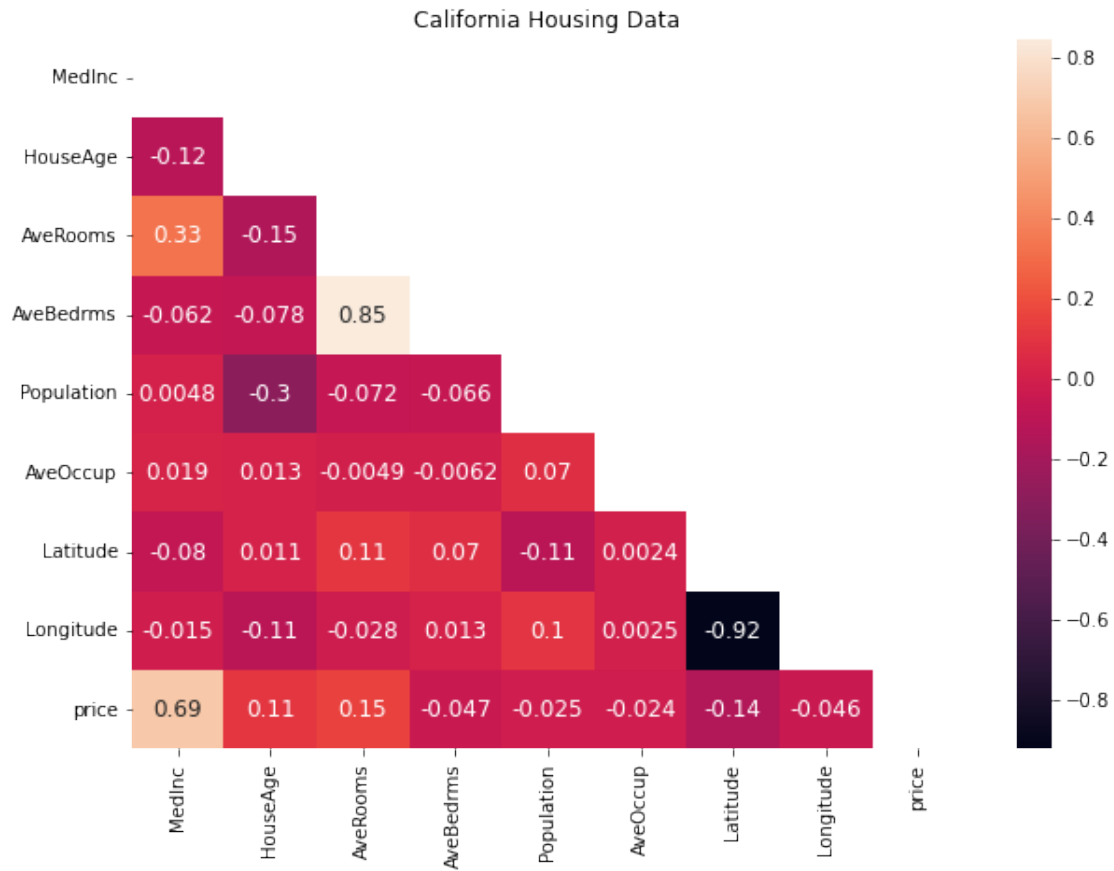
```
[29]: plt.figure(figsize=(10,7))
      mask = np.triu(np.ones_like(df.corr())) # lower half by mask
      ax = sns.heatmap(df.corr(), annot=True, mask= mask)
      plt.title('California Housing Data')
```

```
[29]: Text(0.5, 1.0, 'California Housing Data')
```



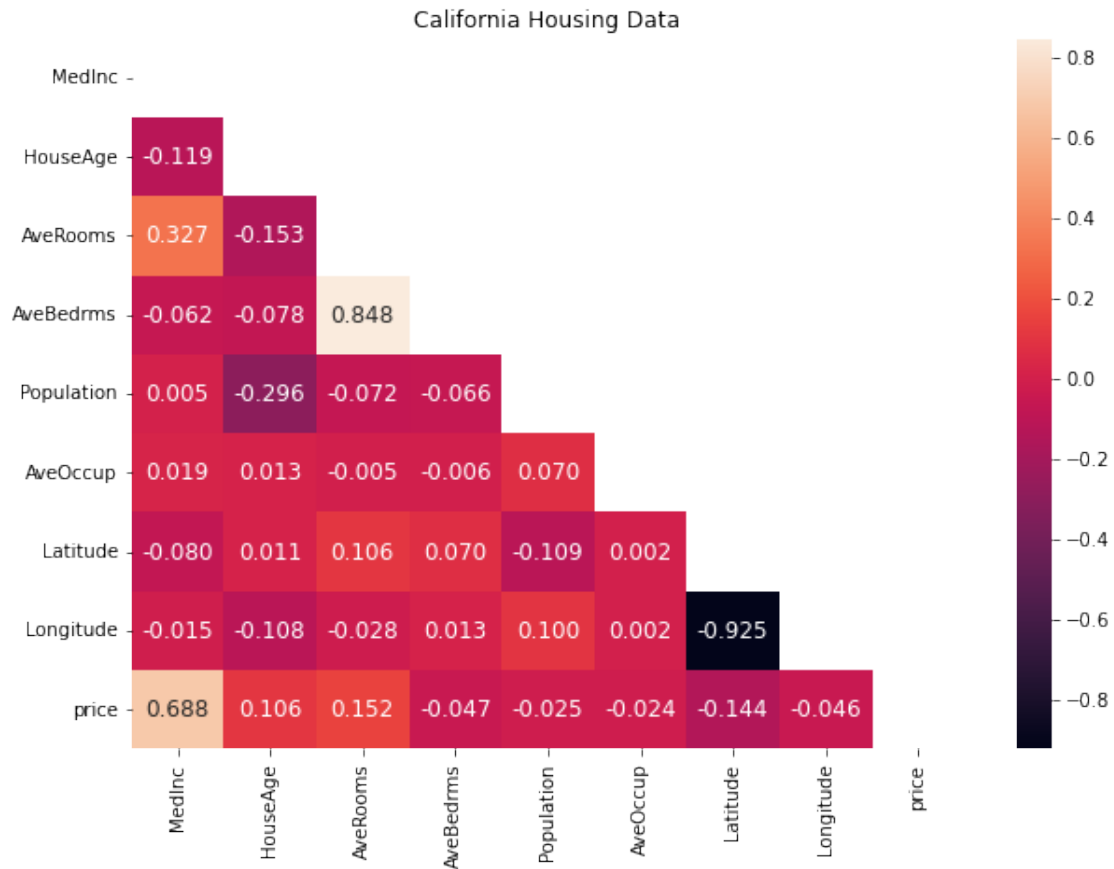
```
[31]: plt.figure(figsize=(10,7))
mask = np.triu(np.ones_like(df.corr())) # lower half by mask
ax = sns.heatmap(df.corr(), annot=True, mask= mask, annot_kws={'fontsize':12})
    ↪ # Increase font size
plt.title('California Housing Data')
```

```
[31]: Text(0.5, 1.0, 'California Housing Data')
```

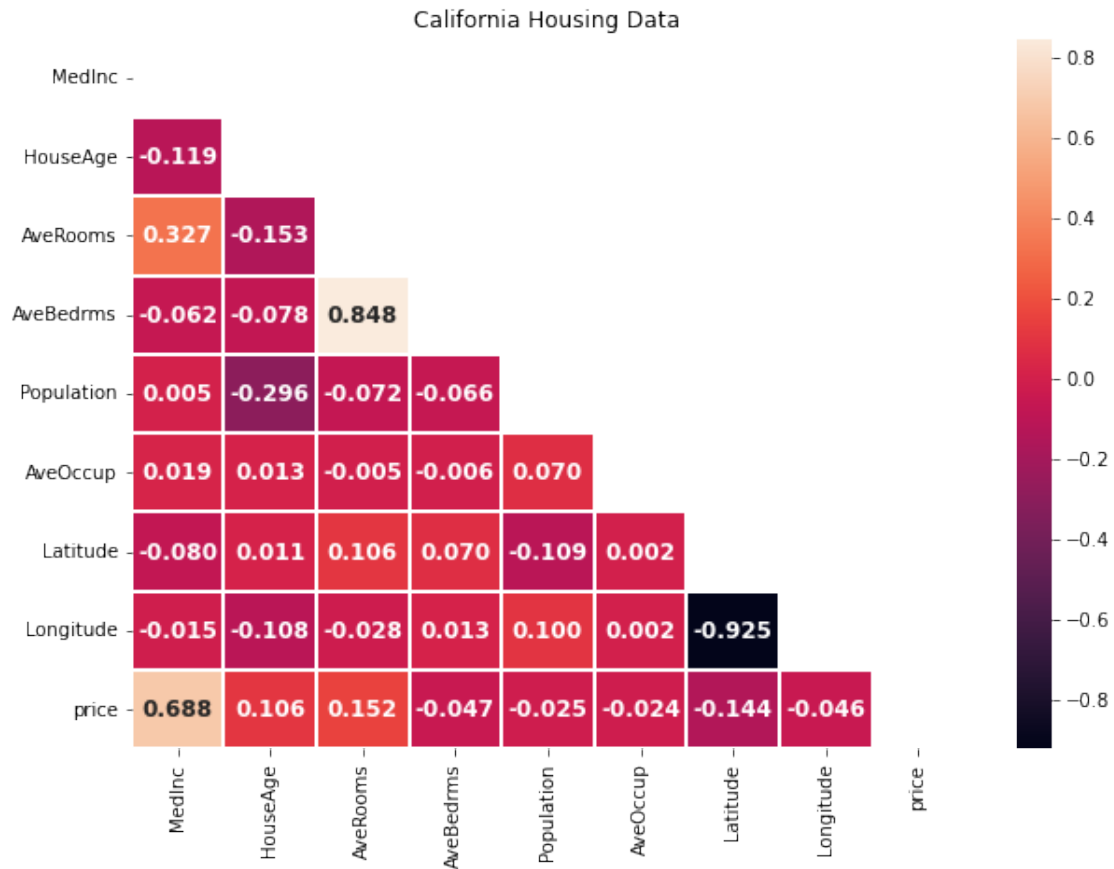
```
[33]: plt.figure(figsize=(10,7))
mask = np.triu(np.ones_like(df.corr())) # lower half by mask
ax = sns.heatmap(df.corr(), annot=True, mask= mask, annot_kws={'fontsize':12},
    ↪fmt="0.3f") # all value at 3 decimal point
plt.title('California Housing Data')
```

```
[33]: Text(0.5, 1.0, 'California Housing Data')
```



```
[36]: plt.figure(figsize=(10,7))
mask = np.triu(np.ones_like(df.corr())) # lower half by mask
ax = sns.heatmap(df.corr(), annot=True, mask= mask, annot_kws={'fontsize':12,
↳ 'fontweight':'bold'}, linewidth=0.3, fmt="0.3f") # bold
plt.title('California Housing Data')
```

```
[36]: Text(0.5, 1.0, 'California Housing Data')
```



```
[38]: plt.figure(figsize=(10,7))
mask = np.triu(np.ones_like(df.corr())) # lower half by mask
ax = sns.heatmap(df.corr(), annot=True, mask= mask, annot_kws={'fontsize':12,
↪ 'fontweight':'bold'}, linewidth=0.3, linecolor='purple', fmt="0.3f")
plt.title('California Housing Data')
```

```
[38]: Text(0.5, 1.0, 'California Housing Data')
```

