

# housing\_visualization

March 22, 2023

## 0.1 BACHELOR OF COMPUTER SCIENCE GROUP:

California Housing Visualization & PDF Conversation + GitHub Link

```
[1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

mydata="./data_source/housing.csv"
data=pd.read_csv(mydata)
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20640 entries, 0 to 20639
Data columns (total 10 columns):
#   Column                Non-Null Count  Dtype
---  -
0   longitude              20640 non-null  float64
1   latitude               20640 non-null  float64
2   housing_median_age     20640 non-null  float64
3   total_rooms            20640 non-null  float64
4   total_bedrooms         20433 non-null  float64
5   population             20640 non-null  float64
6   households             20640 non-null  float64
7   median_income          20640 non-null  float64
8   median_house_value     20640 non-null  float64
9   ocean_proximity        20640 non-null  object
dtypes: float64(9), object(1)
memory usage: 1.6+ MB
```

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

```
[2]:
```

	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
count	20640.000000
mean	206855.816909
std	115395.615874
min	14999.000000
25%	119600.000000
50%	179700.000000
75%	264725.000000
max	500001.000000

```
[3]: data.head()
```

[3]:	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	

	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

```
[4]: data.tail()
```

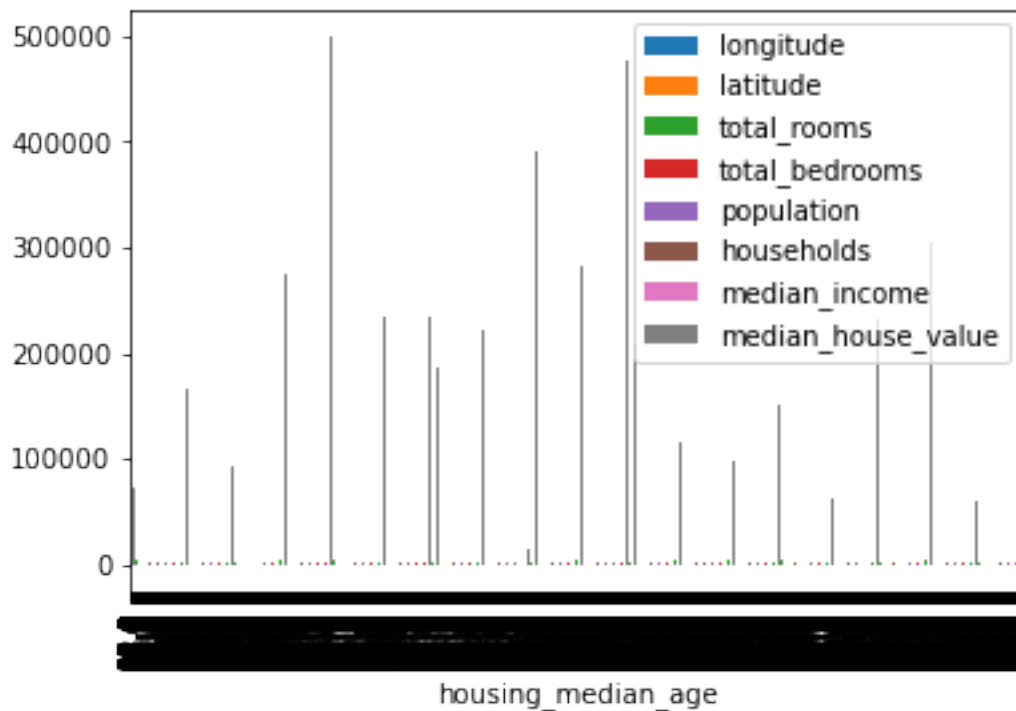
[4]:	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	\
20635	-121.09	39.48	25.0	1665.0	374.0	
20636	-121.21	39.49	18.0	697.0	150.0	
20637	-121.22	39.43	17.0	2254.0	485.0	
20638	-121.32	39.43	18.0	1860.0	409.0	
20639	-121.24	39.37	16.0	2785.0	616.0	

	population	households	median_income	median_house_value \
20635	845.0	330.0	1.5603	78100.0
20636	356.0	114.0	2.5568	77100.0
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
20635	INLAND
20636	INLAND
20637	INLAND
20638	INLAND
20639	INLAND

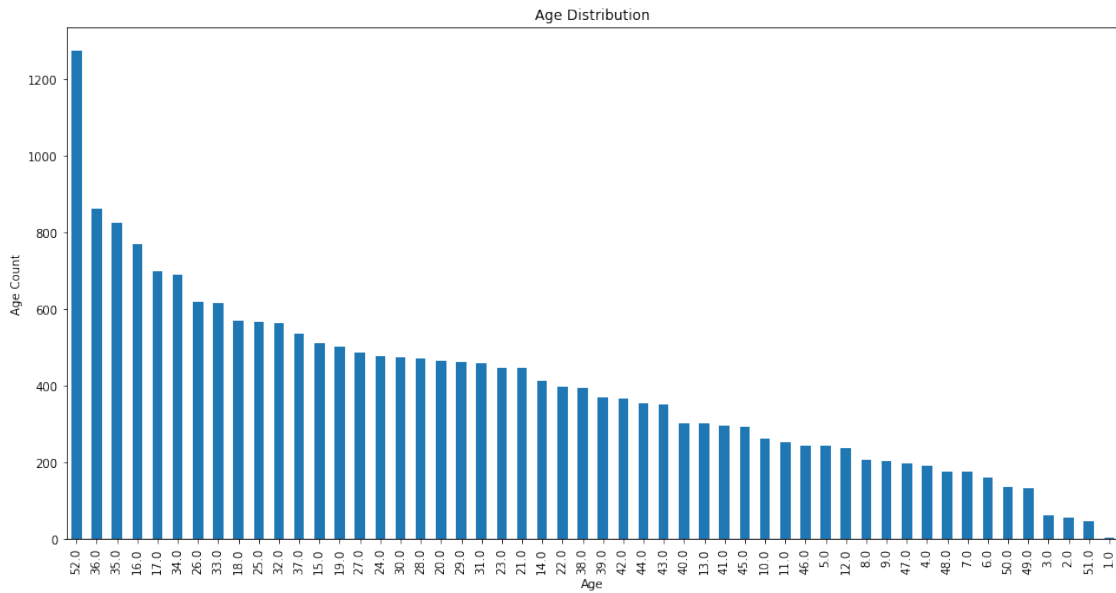
```
[5]: data.set_index('housing_median_age').plot(kind='bar')
```

```
[5]: <AxesSubplot:xlabel='housing_median_age'>
```

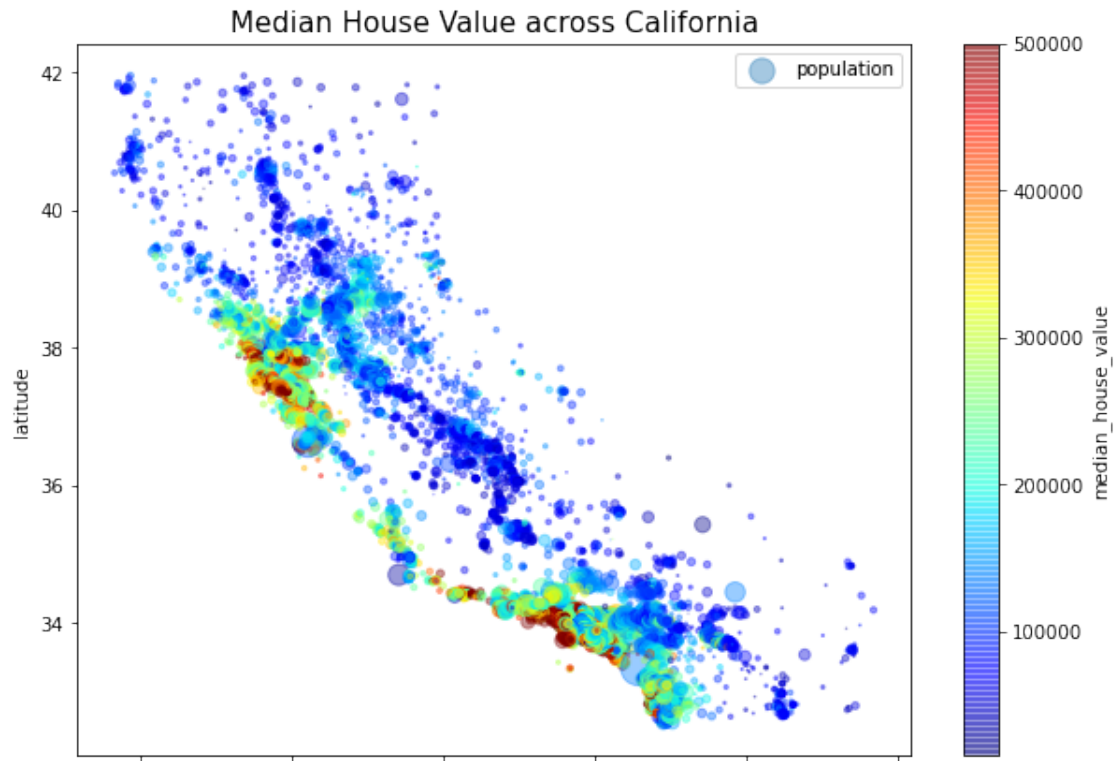


```
[6]: plt.figure(figsize=(16,8))
data['housing_median_age'].value_counts().plot(kind='bar')
plt.title("Age Distribution")
plt.xlabel("Age")
```

```
plt.ylabel("Age Count")
plt.show()
```



```
[7]: data.plot(kind='scatter', x='longitude', y='latitude', alpha=0.4,
              s=data['population']/100, label='population', figsize=(10,7),
              c='median_house_value', cmap=plt.get_cmap('jet'), colorbar=True)
plt.title('Median House Value across California', fontsize=15)
plt.legend()
plt.show()
```



[ ]: