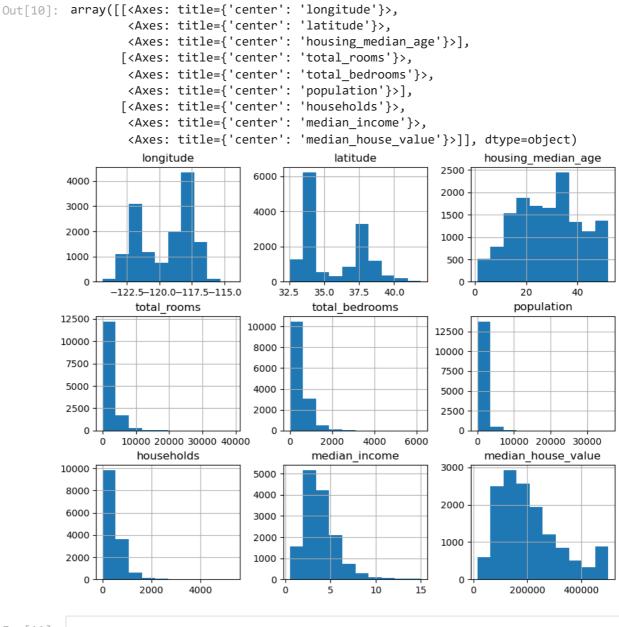
```
In [1]:
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
          import seaborn as sns
In [2]:
         data=pd.read csv("housing.csv")
In [3]:
         data
Out[3]:
                longitude latitude housing_median_age total_rooms total_bedrooms popu
             0
                  -122.23
                             37.88
                                                                880
                                                    41
                                                                              129.0
             1
                  -122.22
                             37.86
                                                    21
                                                               7099
                                                                             1106.0
             2
                  -122.24
                             37.85
                                                    52
                                                               1467
                                                                              190.0
             3
                  -122.25
                             37.85
                                                    52
                                                               1274
                                                                              235.0
             4
                  -122.25
                             37.85
                                                    52
                                                               1627
                                                                              280.0
         20635
                  -121.09
                             39.48
                                                    25
                                                               1665
                                                                              374.0
         20636
                  -121.21
                             39.49
                                                    18
                                                                697
                                                                              150.0
         20637
                  -121.22
                             39.43
                                                    17
                                                               2254
                                                                              485.0
         20638
                  -121.32
                             39.43
                                                    18
                                                               1860
                                                                              409.0
         20639
                  -121.24
                             39.37
                                                    16
                                                               2785
                                                                              616.0
        20640 rows × 10 columns
In [4]:
         data.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 20640 entries, 0 to 20639
       Data columns (total 10 columns):
        #
            Column
                                 Non-Null Count Dtype
       ---
                                 20640 non-null float64
        0
            longitude
            latitude
                                 20640 non-null float64
        1
            housing_median_age 20640 non-null int64
        2
        3
                                 20640 non-null int64
            total_rooms
        4
            total_bedrooms
                                 20433 non-null float64
        5
                                 20640 non-null int64
            population
        6
            households
                                 20640 non-null int64
        7
            median_income
                                 20640 non-null float64
        8
            ocean_proximity
                                 20640 non-null object
            median house value 20640 non-null
                                                 int64
       dtypes: float64(4), int64(5), object(1)
       memory usage: 1.6+ MB
In [5]:
         data.dropna(inplace=True)
```

```
In [6]:
          data.info()
       <class 'pandas.core.frame.DataFrame'>
       Int64Index: 20433 entries, 0 to 20639
       Data columns (total 10 columns):
                                Non-Null Count Dtype
         #
            Column
        ---
                                 -----
         0
            longitude
                                 20433 non-null float64
                                20433 non-null float64
         1
            latitude
            housing_median_age 20433 non-null int64
         2
            total_rooms
         3
                                 20433 non-null int64
                                 20433 non-null float64
         4
         5
            population
                                 20433 non-null int64
         6
            households
                                20433 non-null int64
         7
            median_income
                                 20433 non-null float64
            ocean_proximity
         8
                                 20433 non-null object
            median_house_value 20433 non-null
         9
                                                 int64
       dtypes: float64(4), int64(5), object(1)
       memory usage: 1.7+ MB
 In [7]:
          from sklearn.model_selection import train_test_split
          X=data.drop(['median house value'],axis=1)
          y=data['median_house_value']
 In [8]:
          X train, X test, y train, y test=train test split(X, y, test size=0.3)
 In [9]:
          train_data=X_train.join(y_train)
          train_data
Out[9]:
                 longitude latitude housing_median_age total_rooms total_bedrooms popu
           2130
                   -119.71
                             36.80
                                                    17
                                                              2056
                                                                              366.0
           5868
                   -118.35
                                                                             491.0
                             34.18
                                                    46
                                                              2711
                   -119.05
                                                    35
           2815
                             35.42
                                                              2353
                                                                              483.0
           5704
                   -118.26
                             34.23
                                                    38
                                                              1107
                                                                              194.0
                                                    52
                   -122.26
            486
                             37.86
                                                              3497
                                                                              832.0
                        •••
                                                     •••
           7145
                   -118.13
                             34.02
                                                    43
                                                               396
                                                                              91.0
           4713
                   -118.36
                             34.06
                                                    52
                                                              2130
                                                                             455.0
                                                                             756.0
           5780
                   -118.26
                             34.15
                                                    18
                                                              2481
         15957
                   -122.46
                             37.71
                                                    52
                                                              1580
                                                                              337.0
                                                                              47.0
         12849
                   -121.39
                             38.69
                                                    38
                                                               300
         14303 rows × 10 columns
In [10]:
          train_data.hist(figsize=(10,8))
```



In [11]:
 plt.figure(figsize=(10,6))
 sns.heatmap(train\_data.corr(),annot=True, cmap="YlGnBu")

C:\Users\Sruti Dey\AppData\Local\Temp\ipykernel\_8560\3354081449.py:2: FutureWa rning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

1.00

0.75

0.50

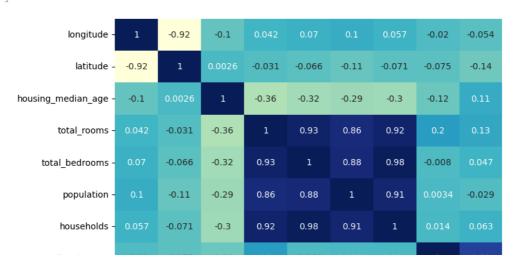
0.25

0.00

-0.25

sns.heatmap(train\_data.corr(),annot=True, cmap="YlGnBu")

Out[11]: <Axes: >



```
In [12]:
          train_data['total_rooms']=np.log(train_data['total_rooms']+1)
          train_data['total_bedrooms']=np.log(train_data['total_bedrooms']+1)
          train_data['population']=np.log(train_data['population']+1)
          train_data['households']=np.log(train_data['households']+1)
          train_data
```

Out[12]:		longitude	latitude	housing_median_age	total_rooms	total_bedrooms	popu
	2130	-119.71	36.80	17	7.629004	5.905362	7.1
	5868	-118.35	34.18	46	7.905442	6.198479	7.1
	2815	-119.05	35.42	35	7.763871	6.182085	7.2
	5704	-118.26	34.23	38	7.010312	5.273000	6.2
	486	-122.26	37.86	52	8.159947	6.725034	7.3
	•••						
	7145	-118.13	34.02	43	5.983936	4.521789	5.5
	4713	-118.36	34.06	52	7.664347	6.122493	6.8
	5780	-118.26	34.15	18	7.816820	6.629363	7.4
	15957	-122.46	37.71	52	7.365813	5.823046	7.2
	12849	-121.39	38.69	38	5.707110	3.871201	5.0

14303 rows × 10 columns

1000

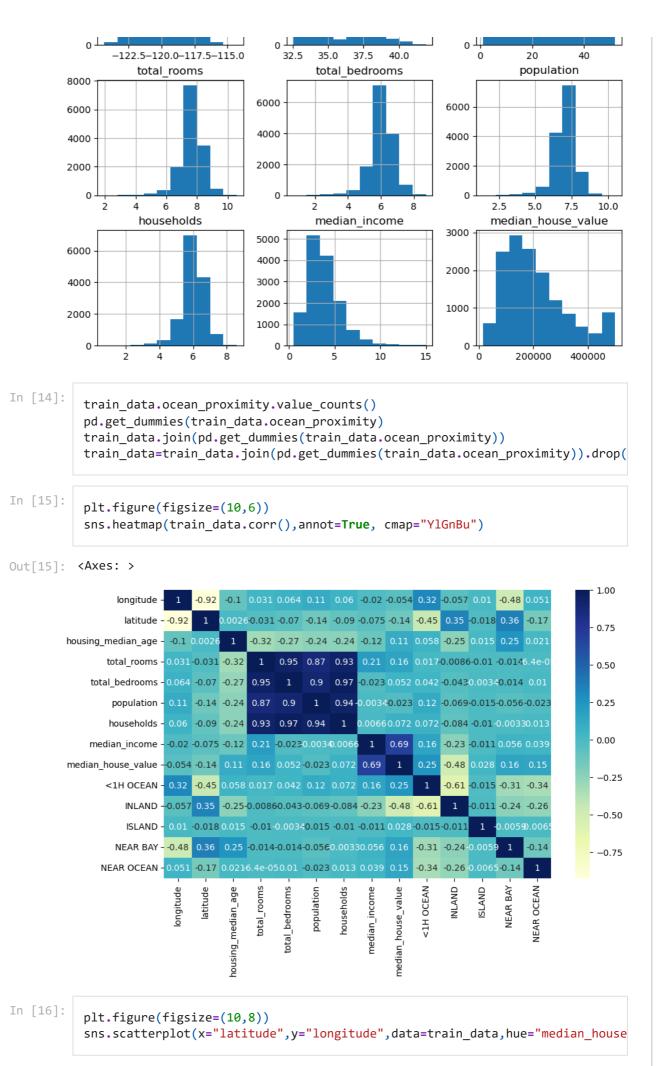
```
In [13]:
          train_data.hist(figsize=(10,8))
```

```
Out[13]: array([[<Axes: title={'center': 'longitude'}>,
                  <Axes: title={'center': 'latitude'}>,
                  <Axes: title={'center': 'housing_median_age'}>],
                 [<Axes: title={'center': 'total_rooms'}>,
                  <Axes: title={'center': 'total_bedrooms'}>,
                  <Axes: title={'center': 'population'}>],
                 [<Axes: title={'center': 'households'}>,
                  <Axes: title={'center': 'median_income'}>,
                  <Axes: title={'center': 'median_house_value'}>]], dtype=object)
                   longitude
                                                latitude
                                                                      housing_median_age
                                                                2500
                                    6000
        4000
                                                                2000
        3000
                                    4000
                                                                1500
        2000
```

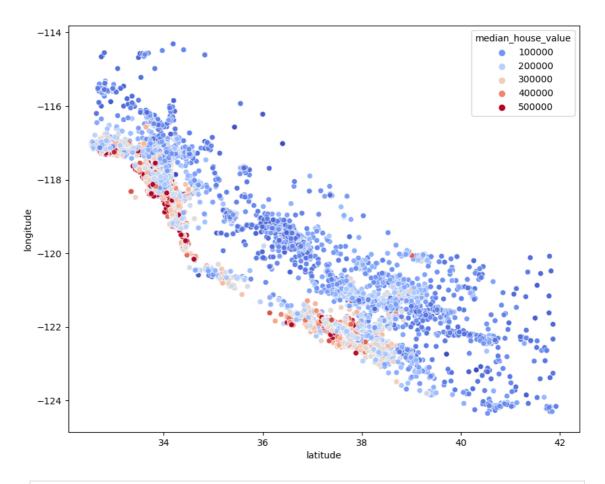
2000

1000

500



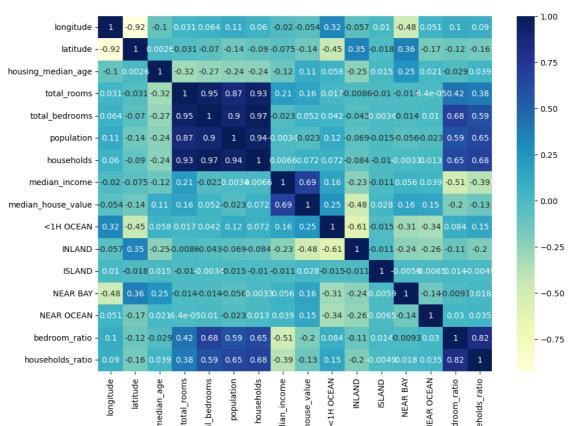
Out[16]: <Axes: xlabel='latitude', ylabel='longitude'>



In [17]:
 train\_data['bedroom\_ratio']=train\_data['total\_bedrooms']/train\_data['total\_r
 train\_data['households\_ratio']=train\_data['households']/train\_data['total\_ro

```
In [18]: plt.figure(figsize=(11,8))
    sns.heatmap(train_data.corr(),annot=True, cmap="YlGnBu")
```

## Out[18]: <Axes: >



## Linear regression

```
from sklearn.linear_model import LinearRegression
    train_data=train_data.drop(['ISLAND'],axis=1)
    X_train, y_train= train_data.drop(['median_house_value'],axis=1), train_data
    reg=LinearRegression()
    reg.fit(X_train,y_train)
```

Out[19]: LinearRegression()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [20]: train_data
```

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( )	IT.	レノ	и	
$\sim$	<i>^</i>	L ~	$\sim$	

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	popu
2130	-119.71	36.80	17	7.629004	5.905362	7.1
5868	-118.35	34.18	46	7.905442	6.198479	7.1
2815	-119.05	35.42	35	7.763871	6.182085	7.2
5704	-118.26	34.23	38	7.010312	5.273000	6.2
486	-122.26	37.86	52	8.159947	6.725034	7.3
•••						
7145	-118.13	34.02	43	5.983936	4.521789	5.5
4713	-118.36	34.06	52	7.664347	6.122493	6.8
5780	-118.26	34.15	18	7.816820	6.629363	7.4
15957	-122.46	37.71	52	7.365813	5.823046	7.2
12849	-121.39	38.69	38	5.707110	3.871201	5.0

14303 rows × 15 columns

In [21]: test\_data=X\_test.join(y\_test)
 test\_data

Out[21]:		longitude	latitude	housing_median_age	total_rooms	total_bedrooms	popu
	7495	-118.25	33.93	38	180	43.0	
	18847	-122.38	41.43	45	2245	448.0	
	13646	-117.31	34.08	43	1697	387.0	

3594	-118.49	34.25	30	2871	470.0
18827	-122.26	41.66	17	1885	350.0
•••					
9655	-120.63	36.98	20	2380	489.0
11367	-117.95	33.74	21	3576	554.0
18752	-122.42	40.63	23	2248	489.0
16622	-120.85	35.37	21	1033	195.0
5135	-118.26	33.97	46	1521	352.0

6130 rows × 10 columns

Out[26]:

```
In [22]:
    test_data['total_rooms']=np.log(test_data['total_rooms']+1)
    test_data['total_bedrooms']=np.log(test_data['total_bedrooms']+1)
    test_data['population']=np.log(test_data['population']+1)
    test_data['households']=np.log(test_data['households']+1)
```

In [23]:
 test\_data=test\_data.join(pd.get\_dummies(test\_data.ocean\_proximity)).drop(['o
 test\_data=test\_data.drop(['ISLAND'],axis=1)

In [24]:
 test\_data['bedroom\_ratio']=test\_data['total\_bedrooms']/test\_data['total\_room
 test\_data['households\_ratio']=test\_data['households']/test\_data['total\_rooms

In [25]: X\_test, y\_test= test\_data.drop(['median\_house\_value'],axis=1), test\_data['me

In [26]: test\_data

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	popu
7495	-118.25	33.93	38	5.198497	3.784190	5.5
18847	-122.38	41.43	45	7.716906	6.107023	7.0
13646	-117.31	34.08	43	7.437206	5.961005	7.0
3594	-118.49	34.25	30	7.962764	6.154858	7.1
18827	-122.26	41.66	17	7.542213	5.860786	6.8
•••						
9655	-120.63	36.98	20	7.775276	6.194405	7.3
11367	-117.95	33.74	21	8.182280	6.318968	7.5
18752	-122.42	40.63	23	7.718241	6.194405	7.0
16622	-120.85	35.37	21	6.941190	5.278115	6.3
5135	-118.26	33.97	46	7.327781	5.866468	7.0

or trust the notebook.

```
In [27]: reg.score(X_test,y_test)

Out[27]: 0.6675915831011885

random-forest (Extra)

In [32]: from sklearn.ensemble import RandomForestRegressor forest=RandomForestRegressor() forest.fit(X_train,y_train)

Out[32]: RandomForestRegressor()
In a Jupyter environment, please rerun this cell to show the HTML representation
```

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [33]: forest.score(X_test,y_test)
Out[33]: 0.8188478263561374
In [ ]:
```