In [1]:

2

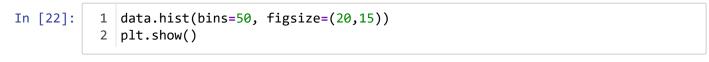
import pandas as pd
import numpy as np

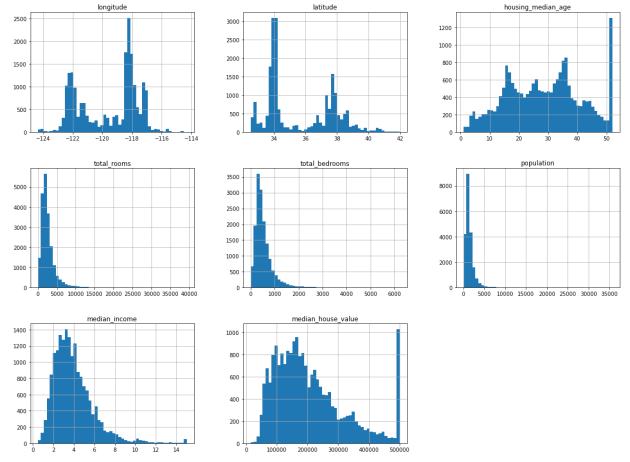
import matplotlib.pyplot as plt

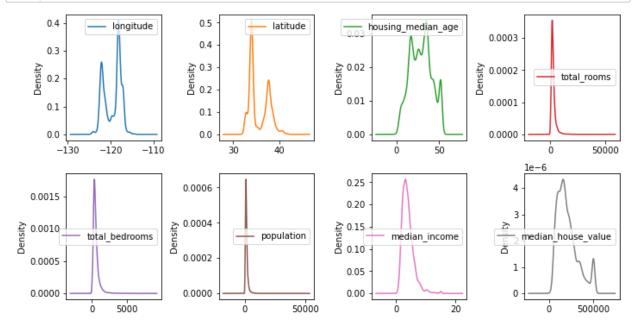
```
import seaborn as sns
In [16]:
               data = pd.read csv("housing2.csv", sep=',',encoding="utf-8")
In [18]:
               data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 20640 entries, 0 to 20639
          Data columns (total 11 columns):
                Column
           #
                                      Non-Null Count
                                                       Dtype
                                      -----
                                                       _ _ _ _ _
                                                       float64
           0
               longitude
                                      20640 non-null
           1
               latitude
                                      20640 non-null
                                                       float64
           2
               housing median age
                                     20382 non-null
                                                       float64
           3
               total rooms
                                      20640 non-null int64
           4
               total_bedrooms
                                      15758 non-null
                                                       float64
           5
               population
                                      20596 non-null
                                                       float64
           6
               households
                                      19335 non-null
                                                       object
           7
               median income
                                      17873 non-null
                                                       float64
           8
               median_house_value
                                     20640 non-null
                                                       int64
           9
               ocean proximity
                                      20640 non-null
                                                       object
           10
               gender
                                                       object
                                      16620 non-null
          dtypes: float64(6), int64(2), object(3)
          memory usage: 1.7+ MB
               data.describe()
 In [3]:
 Out[3]:
                     longitude
                                   latitude
                                           housing_median_age
                                                               total_rooms total_bedrooms
                                                                                            population
           count 20640.000000
                              20640.000000
                                                  20382.000000
                                                              20640.000000
                                                                             15758.000000
                                                                                          20596.0000
                   -119.569704
                                 35.631861
                                                     28.676283
                                                               2635.763081
                                                                               539.920104
                                                                                           1424.9287
           mean
             std
                     2.003532
                                  2.135952
                                                     12.589284
                                                               2181.615252
                                                                               419.834171
                                                                                           1132.2377
            min
                   -124.350000
                                 32.540000
                                                     1.000000
                                                                  2.000000
                                                                                 1.000000
                                                                                              3.0000
            25%
                   -121.800000
                                 33.930000
                                                     18.000000
                                                                                            787.0000
                                                               1447.750000
                                                                               296.000000
            50%
                   -118.490000
                                 34.260000
                                                     29.000000
                                                               2127.000000
                                                                               435.000000
                                                                                           1166.0000
            75%
                   -118.010000
                                 37.710000
                                                     37.000000
                                                               3148.000000
                                                                               652.000000
                                                                                           1725.0000
            max
                   -114.310000
                                 41.950000
                                                     52.000000
                                                              39320.000000
                                                                              6210.000000
                                                                                          35682.0000
In [19]:
            1 data.columns
Out[19]: Index(['longitude', 'latitude', 'housing median age', 'total rooms',
                  'total_bedrooms', 'population', 'households', 'median_income',
                  'median_house_value', 'ocean_proximity', 'gender'],
                 dtype='object')
```

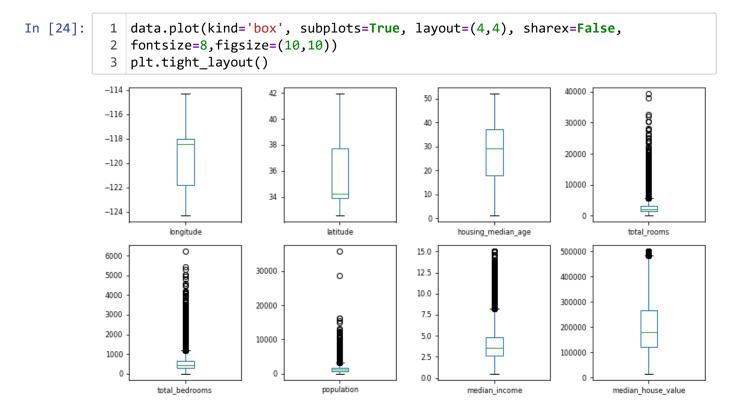
```
1 data['ocean_proximity'].value_counts()
In [20]:
Out[20]: <1H OCEAN
                        9136
         INLAND
                        6551
         NEAR OCEAN
                        2658
         NEAR BAY
                        2290
         ISLAND
         Name: ocean_proximity, dtype: int64
In [47]:
              data['households'].value_counts()
Out[47]: no
                  3080
         282
                    47
         375
                    46
         306
                    45
          380
                    45
         2392
                     1
         1577
                     1
         2838
                     1
         1747
                     1
         2159
                     1
         Name: households, Length: 1703, dtype: int64
```

```
In [21]:
           1 data['housing_median_age'].value_counts()
Out[21]: 52.0
                  1261
          36.0
                    856
                   820
          35.0
          16.0
                    764
          17.0
                   686
          34.0
                   683
          33.0
                    609
          26.0
                   604
          18.0
                    562
                    558
          25.0
          32.0
                    556
          37.0
                   532
          15.0
                   508
          19.0
                   497
                   476
          27.0
          24.0
                   471
          30.0
                   468
          28.0
                   465
          20.0
                   455
          29.0
                   455
                   453
          31.0
                   441
          21.0
          23.0
                   437
          14.0
                   412
                    391
          22.0
          38.0
                    389
          42.0
                    366
          39.0
                    365
          44.0
                    356
          43.0
                    349
                    301
          40.0
          13.0
                   297
          41.0
                    295
          45.0
                    291
          10.0
                    257
          11.0
                    246
          46.0
                   245
          5.0
                    239
                    235
          12.0
          9.0
                    204
          8.0
                    204
          47.0
                   195
          4.0
                   188
          48.0
                   176
          7.0
                    173
          6.0
                   153
          50.0
                   134
          49.0
                   133
          3.0
                    61
                     58
          2.0
                     48
          51.0
          1.0
          Name: housing_median_age, dtype: int64
```

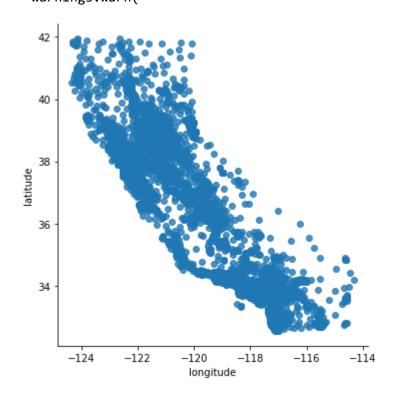






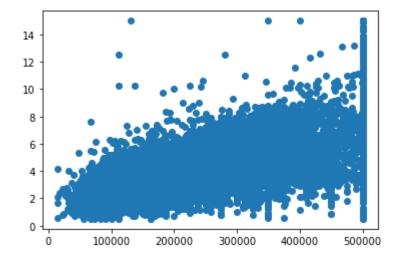


```
In [29]: 1 sns.lmplot('longitude', 'latitude', data=data, fit_reg=False, height=5)
2 plt.show()
```



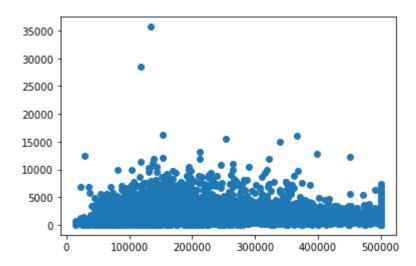
In [38]: 1 plt.scatter(data['median_house_value'], data['median_income'])

Out[38]: <matplotlib.collections.PathCollection at 0x19e397462e0>



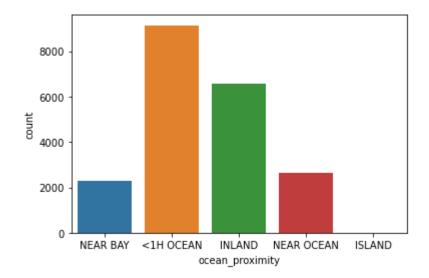
In [41]: 1 plt.scatter(data['median_house_value'], data['population'])

Out[41]: <matplotlib.collections.PathCollection at 0x19e354a03d0>

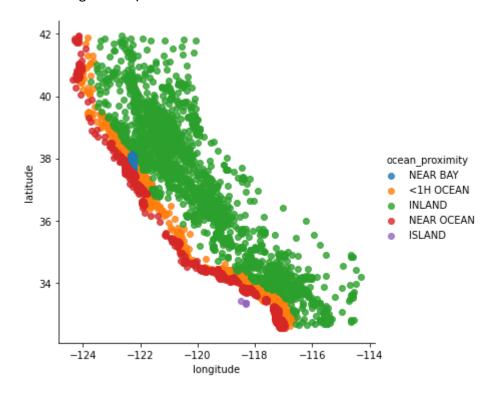


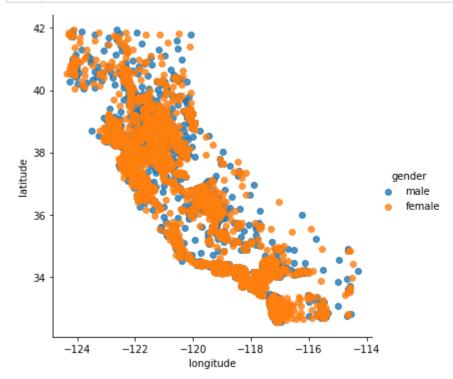
In [25]: 1 sns.countplot(data['ocean_proximity'])

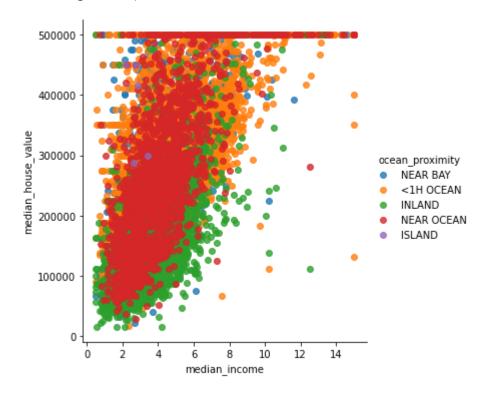
Out[25]: <AxesSubplot:xlabel='ocean_proximity', ylabel='count'>

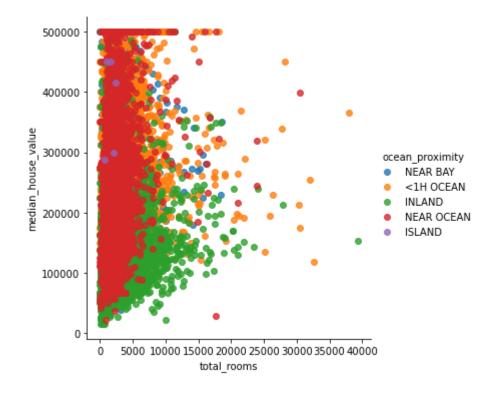


```
In [4]: 1 sns.lmplot('longitude', 'latitude', data=data, hue='ocean_proximity', fit_re
    plt.show()
```



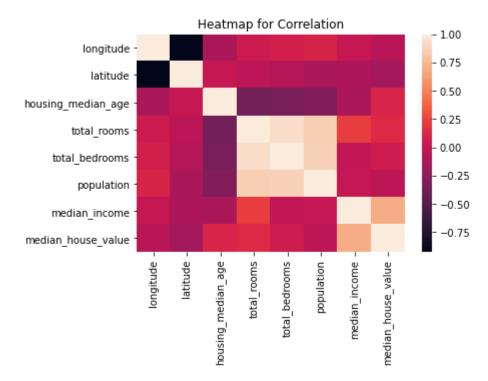






In [30]: 1 sns.heatmap(data.corr()).set_title('Heatmap for Correlation')

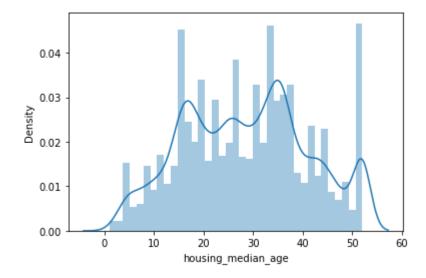
Out[30]: Text(0.5, 1.0, 'Heatmap for Correlation')



In [31]: | 1 | sns.distplot(data['housing_median_age'])

C:\Users\Qebaa\anaconda3\lib\site-packages\seaborn\distributions.py:2551: Futur eWarning: `distplot` is a deprecated function and will be removed in a future v ersion. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histogram s).

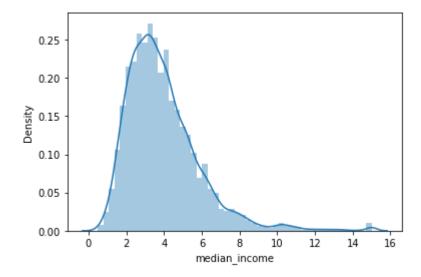
Out[31]: <AxesSubplot:xlabel='housing_median_age', ylabel='Density'>



In [33]: 1 | sns.distplot(data['median_income'])

C:\Users\Qebaa\anaconda3\lib\site-packages\seaborn\distributions.py:2551: Futur eWarning: `distplot` is a deprecated function and will be removed in a future v ersion. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histogram s).

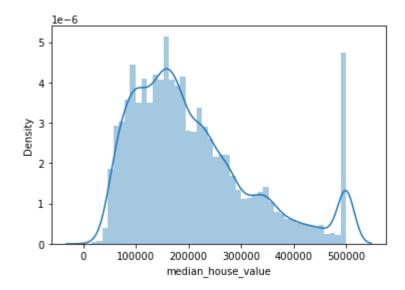
Out[33]: <AxesSubplot:xlabel='median_income', ylabel='Density'>



In [34]: 1 sns.distplot(data['median_house_value'])

C:\Users\Qebaa\anaconda3\lib\site-packages\seaborn\distributions.py:2551: Futur eWarning: `distplot` is a deprecated function and will be removed in a future v ersion. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histogram s).

Out[34]: <AxesSubplot:xlabel='median_house_value', ylabel='Density'>



In [43]: 1 data.sort_values(['median_income', 'median_house_value'], ascending=True)

Out[43]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	househol
19800	-123.32	40.43	15.0	661	NaN	131.0	N
73	-122.29	37.81	46.0	12	4.0	18.0	
3258	-122.89	39.42	16.0	411	114.0	26.0	
19523	-121.01	37.65	52.0	178	53.0	152.0	
5213	-118.28	33.93	52.0	117	33.0	74.0	
14545	-117.26	32.95	15.0	1036	149.0	395.0	1
14546	-117.26	32.95	15.0	1882	233.0	704.0	2
14806	-117.18	32.68	29.0	1539	344.0	556.0	2
14807	-117.18	32.69	52.0	1837	313.0	668.0	3
14810	-117.17	32.69	40.0	2236	331.0	767.0	3

20640 rows × 11 columns

In [44]: 1 data.sort_values(['housing_median_age','median_house_value'],ascending=True)

Out[44]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	househol				
12286	-116.95	33.86	1.0	6	2.0	8.0					
3130	-117.95	35.08	1.0	83	15.0	32.0					
19536	-120.93	37.65	1.0	2254	328.0	402.0	1				
18972	-122.00	38.23	1.0	2062	343.0	872.0	2				
2774	-115.80	33.26	2.0	96	18.0	30.0					
20422	-118.90	34.14	NaN	1503	NaN	576.0	N				
20426	-118.69	34.18	NaN	1177	NaN	415.0	N				
20427	-118.80	34.19	NaN	15572	NaN	5495.0	N;				
20436	-118.69	34.21	NaN	3663	NaN	1179.0	Ni				
20443	-118.85	34.27	NaN	187	NaN	130.0	N				
20640	20640 rows × 11 columns										

In [45]: 1 data.sort_values(['housing_median_age','population'],ascending=True)

Out[45]:

longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	households	me
-116.95	33.86	1.0	6	2.0	8.0	2	
-117.95	35.08	1.0	83	15.0	32.0	15	
-120.93	37.65	1.0	2254	328.0	402.0	112	
-122.00	38.23	1.0	2062	343.0	872.0	268	
-121.96	37.74	2.0	200	20.0	25.0	9	
-118.90	34.26	NaN	25187	NaN	11956.0	NaN	
-122.25	37.85	NaN	1627	280.0	NaN	259	
-122.25	37.85	NaN	919	213.0	NaN	193	
-122.25	37.84	NaN	2535	NaN	NaN	514	
-122.25	37.84	NaN	3104	NaN	NaN	NaN	

rows × 11 columns

In [46]: 1 data.sort_values(['median_house_value','ocean_proximity'],ascending=True)

Out[46]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	househol
2521	-122.74	39.71	16.0	255	73.0	85.0	
2799	-117.02	36.40	19.0	619	239.0	490.0	1
9188	-117.86	34.24	52.0	803	NaN	628.0	
19802	-123.17	40.31	36.0	98	NaN	18.0	Ni
5887	-118.33	34.15	39.0	493	168.0	259.0	1
20233	-119.29	34.24	27.0	4742	NaN	1682.0	Ni
20272	-119.23	34.19	16.0	5297	NaN	1489.0	Ni
20273	-119.23	34.17	18.0	6171	NaN	2164.0	Ni
20322	-119.14	34.23	8.0	243	NaN	102.0	Ni
20380	-118.83	34.14	16.0	1316	NaN	450.0	Ni

20640 rows × 11 columns

```
In [50]: 1 new_val.head(5)
```

Out[50]:

_		<1H OCEAN	INLAND	ISLAND	NEAR BAY	NEAR OCEAN
-	0	0	0	0	1	0
	1	0	0	0	1	0
	2	0	0	0	1	0
	3	0	0	0	1	0
	4	0	0	0	1	0

```
In [51]: 1 data[new_val.columns] = new_val
In [52]: 1 data.describe()
2
```

Out[52]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	populati
cou	nt 20640.000000	20640.000000	20382.000000	20640.000000	15758.000000	20596.0000
mea	n -119.569704	35.631861	28.676283	2635.763081	539.920104	1424.9287
st	d 2.003532	2.135952	12.589284	2181.615252	419.834171	1132.2377
mi	n -124.350000	32.540000	1.000000	2.000000	1.000000	3.0000
25	-121.800000	33.930000	18.000000	1447.750000	296.000000	787.0000
50	-118.490000	34.260000	29.000000	2127.000000	435.000000	1166.0000
75	-118.010000	37.710000	37.000000	3148.000000	652.000000	1725.0000
ma	x -114.310000	41.950000	52.000000	39320.000000	6210.000000	35682.0000
4						>

In [53]: 1 data.columns

In [58]: 1 data.corr()

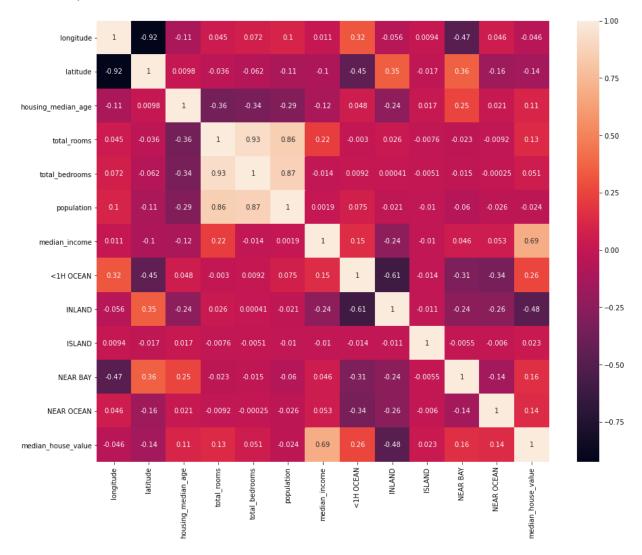
Out[58]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	рорі
longitude	1.000000	-0.924664	-0.107447	0.044568	0.071610	0.1
latitude	-0.924664	1.000000	0.009773	-0.036100	-0.062301	-0.1
housing_median_age	-0.107447	0.009773	1.000000	-0.360441	-0.336824	-0.2
total_rooms	0.044568	-0.036100	-0.360441	1.000000	0.926709	3.0
total_bedrooms	0.071610	-0.062301	-0.336824	0.926709	1.000000	3.0
population	0.100390	-0.109222	-0.294245	0.856914	0.871130	1.0
median_income	0.011440	-0.103496	-0.117771	0.220391	-0.013803	0.0
<1H OCEAN	0.321121	-0.446969	0.048219	-0.003031	0.009239	0.0
INLAND	-0.055575	0.351166	-0.238743	0.025624	0.000414	-0.0
ISLAND	0.009446	-0.016572	0.017076	-0.007572	-0.005072	-0.0
NEAR BAY	-0.474489	0.358771	0.252832	-0.023022	-0.015300	-0.0
NEAR OCEAN	0.045509	-0.160818	0.021194	-0.009175	-0.000252	-0.0
median_house_value	-0.045967	-0.144160	0.107378	0.134153	0.050963	-0.0

In [61]: 1 plt.figure(figsize=(15,12))

sns.heatmap(data.corr(), annot=True)

Out[61]: <AxesSubplot:>



```
In [62]: 1 data['gender'].replace("male",0,inplace=True)
    data['gender'].replace("female",1,inplace=True)
    data.head()
4
```

C:\Users\Qebaa\anaconda3\lib\site-packages\pandas\core\series.py:4563: SettingW
ithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

return super().replace(

Out[62]:

ng_median_age	total_rooms	total_bedrooms	population	households	median_income	gender	OCE,
41.0	880	129.0	322.0	126	8.3252	0.0	
21.0	7099	1106.0	2401.0	1138	8.3014	1.0	
52.0	1467	190.0	496.0	177	7.2574	0.0	
52.0	1274	235.0	558.0	219	5.6431	1.0	
NaN	1627	280.0	NaN	259	3.8462	0.0	

```
In [63]: 1 data.isna().sum()
```

Out[63]: longitude 0 latitude 0 housing_median_age 258 total rooms 0 total bedrooms 4882 population 44 households 1305 2767 median_income gender 4020 <1H OCEAN 0 **INLAND** 0 **ISLAND** 0 **NEAR BAY** 0 **NEAR OCEAN** 0 median_house_value

```
In [70]: 1 data = data.fillna(data.mean())
```

dtype: int64

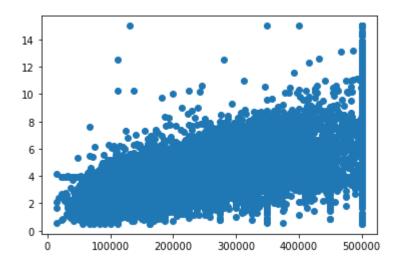
In [73]: 1 data.head(20)

Out[73]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	population	househo
0	-122.23	37.88	41.000000	880	129.000000	322.000000	
1	-122.22	37.86	21.000000	7099	1106.000000	2401.000000	1
2	-122.24	37.85	52.000000	1467	190.000000	496.000000	
3	-122.25	37.85	52.000000	1274	235.000000	558.000000	
4	-122.25	37.85	28.676283	1627	280.000000	1424.928724	
5	-122.25	37.85	28.676283	919	213.000000	1424.928724	
6	-122.25	37.84	28.676283	2535	539.920104	1424.928724	
7	-122.25	37.84	28.676283	3104	539.920104	1424.928724	1
8	-122.26	37.84	42.000000	2555	539.920104	1424.928724	1
9	-122.25	37.84	52.000000	3549	539.920104	1424.928724	1
10	-122.26	37.85	52.000000	2202	539.920104	1424.928724	1
11	-122.26	37.85	52.000000	3503	539.920104	1424.928724	1
12	-122.26	37.85	52.000000	2491	539.920104	1424.928724	L
13	-122.26	37.84	52.000000	696	191.000000	345.000000	1
14	-122.26	37.85	52.000000	2643	626.000000	1212.000000	1
15	-122.26	37.85	50.000000	1120	283.000000	697.000000	1
16	-122.27	37.85	52.000000	1966	347.000000	793.000000	1
17	-122.27	37.85	52.000000	1228	293.000000	648.000000	
18	-122.26	37.84	50.000000	2239	455.000000	990.000000	
19	-122.27	37.84	52.000000	1503	298.000000	690.000000	
4							+

In [72]: 1 plt.scatter(data['median_house_value'], data['median_income'])

Out[72]: <matplotlib.collections.PathCollection at 0x19e3577f8e0>

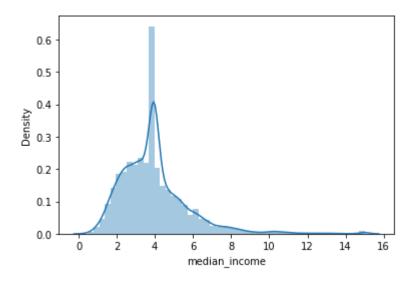


In [74]: 1 | sns.distplot(data['median_income'])

C:\Users\Qebaa\anaconda3\lib\site-packages\seaborn\distributions.py:2551: Futur eWarning: `distplot` is a deprecated function and will be removed in a future v ersion. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histogram s).

warnings.warn(msg, FutureWarning)

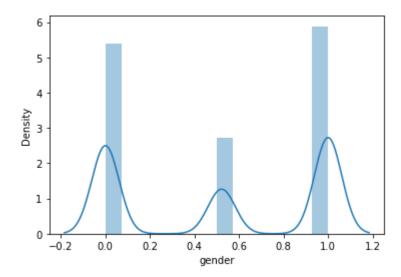
Out[74]: <AxesSubplot:xlabel='median_income', ylabel='Density'>



In [76]: 1 sns.distplot(data['gender'])

C:\Users\Qebaa\anaconda3\lib\site-packages\seaborn\distributions.py:2551: Futur eWarning: `distplot` is a deprecated function and will be removed in a future v ersion. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histogram s).

Out[76]: <AxesSubplot:xlabel='gender', ylabel='Density'>

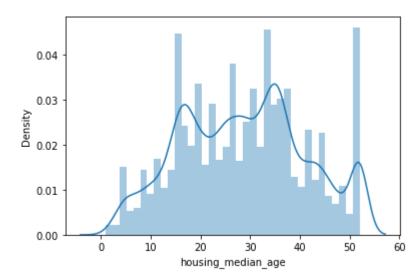


In [77]: 1 sns.distplot(data['housing_median_age'])

C:\Users\Qebaa\anaconda3\lib\site-packages\seaborn\distributions.py:2551: Futur eWarning: `distplot` is a deprecated function and will be removed in a future v ersion. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histogram s).

warnings.warn(msg, FutureWarning)

Out[77]: <AxesSubplot:xlabel='housing_median_age', ylabel='Density'>



In []: 1