lab03_partA

September 16, 2024

[25]:
```
import pandas as pd
```

[26]:
```
housing = pd.read_csv('housing.csv')
housing.head()
```

[26]:

	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
0	322.0	126.0	8.3252	452600.0
1	2401.0	1138.0	8.3014	358500.0
2	496.0	177.0	7.2574	352100.0
3	558.0	219.0	5.6431	341300.0
4	565.0	259.0	3.8462	342200.0

[27]:
```
housing.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20640 entries, 0 to 20639
Data columns (total 9 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
dtypes: float64(9)
memory usage: 1.4 MB
```

[28]:
```
housing.describe()
```

[28]:

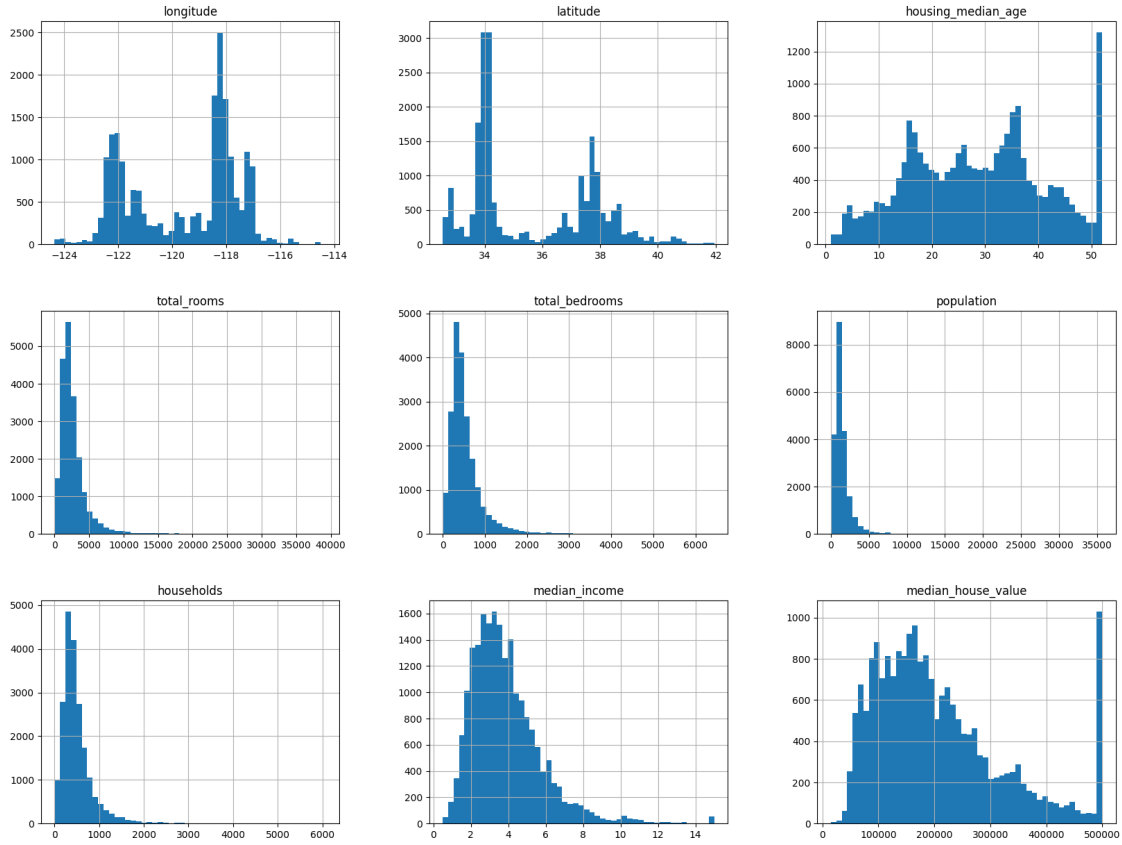
	longitude	latitude	housing_median_age	total_rooms	\
count	20640.000000	20640.000000	20640.000000	20640.000000	
mean	-119.569704	35.631861	28.639486	2635.763081	
std	2.003532	2.135952	12.585558	2181.615252	
min	-124.350000	32.540000	1.000000	2.000000	
25%	-121.800000	33.930000	18.000000	1447.750000	
50%	-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

0.1 Question 3

[29]:
```
%matplotlib inline
import matplotlib.pyplot as plt
housing.hist(bins=50, figsize=(20,15))
plt.show()
```



0.2 Question 4

[30]:
```
# to make this notebook's output identical and reproducible at every run, we need set seed
import numpy as np
np.random.seed(10)
```

[31]:
```
# For illustration only. Sklearn has train_test_split()
def split_train_test(data, test_ratio):
    shuffled_indices = np.random.permutation(len(data))
    test_set_size = int(len(data) * test_ratio)
    test_indices = shuffled_indices[:test_set_size]
    train_indices = shuffled_indices[test_set_size:]
    return data.iloc[train_indices], data.iloc[test_indices]
```

[32]:
```
# run the function to get the train & test set
train_set, test_set = split_train_test(housing, 0.2)
```

[33]:
```
train_samples = train_set.shape[0]
test_samples = test_set.shape[0]
```

```
print(f"Number of samples in train_set: {train_samples}")
print(f"Number of samples in test_set: {test_samples}")
print(f"Total samples: {train_samples + test_samples}")

total_samples = housing.shape[0]
print(f"Total samples in the housing dataset: {total_samples}")

assert train_samples + test_samples == total_samples, "The total samples do not␣
 ↪match!"
```

```
Number of samples in train_set: 16512
Number of samples in test_set: 4128
Total samples: 20640
Total samples in the housing dataset: 20640
```

[34]:
```
%pip install sklearn
```

```
Collecting sklearn
  Using cached sklearn-0.0.post12.tar.gz (2.6 kB)
  Preparing metadata (setup.py): started
  Preparing metadata (setup.py): finished with status 'error'
Note: you may need to restart the kernel to use updated packages.
  error: subprocess-exited-with-error

  × python setup.py egg_info did not run successfully.
    exit code: 1
   > [15 lines of output]
      The 'sklearn' PyPI package is deprecated, use 'scikit-learn'
      rather than 'sklearn' for pip commands.

      Here is how to fix this error in the main use cases:
      - use 'pip install scikit-learn' rather than 'pip install sklearn'
      - replace 'sklearn' by 'scikit-learn' in your pip requirements files
        (requirements.txt, setup.py, setup.cfg, Pipfile, etc …)
      - if the 'sklearn' package is used by one of your dependencies,
        it would be great if you take some time to track which package uses
        'sklearn' instead of 'scikit-learn' and report it to their issue tracker
      - as a last resort, set the environment variable
        SKLEARN_ALLOW_DEPRECATED_SKLEARN_PACKAGE_INSTALL=True to avoid this
error

      More information is available at
      https://github.com/scikit-learn/sklearn-pypi-package
      [end of output]

  note: This error originates from a subprocess, and is likely not a problem
with pip.
```

```
error: metadata-generation-failed

× Encountered error while generating package metadata.
 > See above for output.

note: This is an issue with the package mentioned above, not pip.
hint: See above for details.

[notice] A new release of pip is available: 24.1.2 -> 24.2
[notice] To update, run: python.exe -m pip install --upgrade pip
```

[35]:
```
from sklearn.model_selection import train_test_split
train_set, test_set = train_test_split(housing, test_size=0.2, random_state=10)
```

[36]:
```
train_set[train_set.isna().any(axis=1)]
```

[36]:

	longitude	latitude	housing_median_age	total_rooms	total_bedrooms	\
6814	-118.07	34.10	32.0	4275.0	NaN	
4738	-118.38	34.05	49.0	702.0	NaN	
290	-122.16	37.77	47.0	1256.0	NaN	
19833	-119.38	36.53	38.0	1281.0	NaN	
4852	-118.31	34.03	47.0	1315.0	NaN	
…	…	…	…	…	…	
11512	-118.10	33.74	32.0	2035.0	NaN	
13656	-117.30	34.05	6.0	2155.0	NaN	
13015	-121.19	38.71	11.0	4415.0	NaN	
10236	-117.92	33.87	33.0	1597.0	NaN	
14521	-117.14	32.90	16.0	3217.0	NaN	

	population	households	median_income	median_house_value
6814	2812.0	1012.0	3.3512	214100.0
4738	458.0	187.0	4.8958	333600.0
290	570.0	218.0	4.3750	161900.0
19833	1423.0	293.0	1.9602	51400.0
4852	785.0	245.0	1.2300	138400.0
…	…	…	…	…
11512	934.0	512.0	4.2287	500001.0
13656	1039.0	391.0	1.6675	95800.0
13015	1520.0	627.0	3.2321	390800.0
10236	1888.0	423.0	3.0550	157800.0
14521	2054.0	687.0	4.2234	162100.0

[166 rows x 9 columns]

[37]:
```
train_set_clean = train_set.dropna(subset=["total_bedrooms"])
train_set_clean
```

[37]:
```
       longitude  latitude  housing_median_age  total_rooms  total_bedrooms  \
12346    -116.52     33.82                21.0      10227.0          2315.0
19326    -122.94     38.50                46.0       2280.0           492.0
16824    -122.49     37.63                31.0       3109.0           621.0
6869     -118.11     34.06                16.0       2416.0           565.0
16677    -120.66     35.13                41.0       2666.0           751.0
…           …        …                    …          …               …
9372     -122.53     37.93                37.0       1722.0           352.0
7291     -118.22     33.98                18.0       1781.0           765.0
17728    -121.79     37.32                 6.0       2850.0           561.0
7293     -118.23     33.98                35.0       1366.0           496.0
17673    -121.88     37.30                16.0       2692.0           749.0

       population  households  median_income  median_house_value
12346      3623.0      1734.0         2.5212            145200.0
19326       807.0       366.0         2.6316            117000.0
16824      1472.0       618.0         5.1550            263900.0
6869       1750.0       514.0         2.8229            163700.0
16677       940.0       507.0         1.9653            236100.0
…             …          …              …                 …
9372        648.0       337.0         4.1250            310300.0
7291       1913.0       702.0         1.2059            255000.0
17728      2160.0       581.0         5.5336            241900.0
7293       2160.0       497.0         2.2059            150000.0
17673      1674.0       681.0         2.6763            191100.0

[16346 rows x 9 columns]
```

[38]:
```
train_labels = train_set_clean["median_house_value"].copy() # get labels for␣
 ↪output label Y
train_features = train_set_clean.drop("median_house_value", axis=1) # drop␣
 ↪labels to get features X for training set
```

0.3 Question 8

Task 1

[39]:
```
train_features.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 16346 entries, 12346 to 17673
Data columns (total 8 columns):
 #   Column              Non-Null Count  Dtype
---  ------              --------------  -----
 0   longitude           16346 non-null  float64
 1   latitude            16346 non-null  float64
 2   housing_median_age  16346 non-null  float64
 3   total_rooms         16346 non-null  float64
 4   total_bedrooms      16346 non-null  float64
```

```
 5   population          16346 non-null  float64
 6   households          16346 non-null  float64
 7   median_income       16346 non-null  float64
dtypes: float64(8)
memory usage: 1.1 MB
```

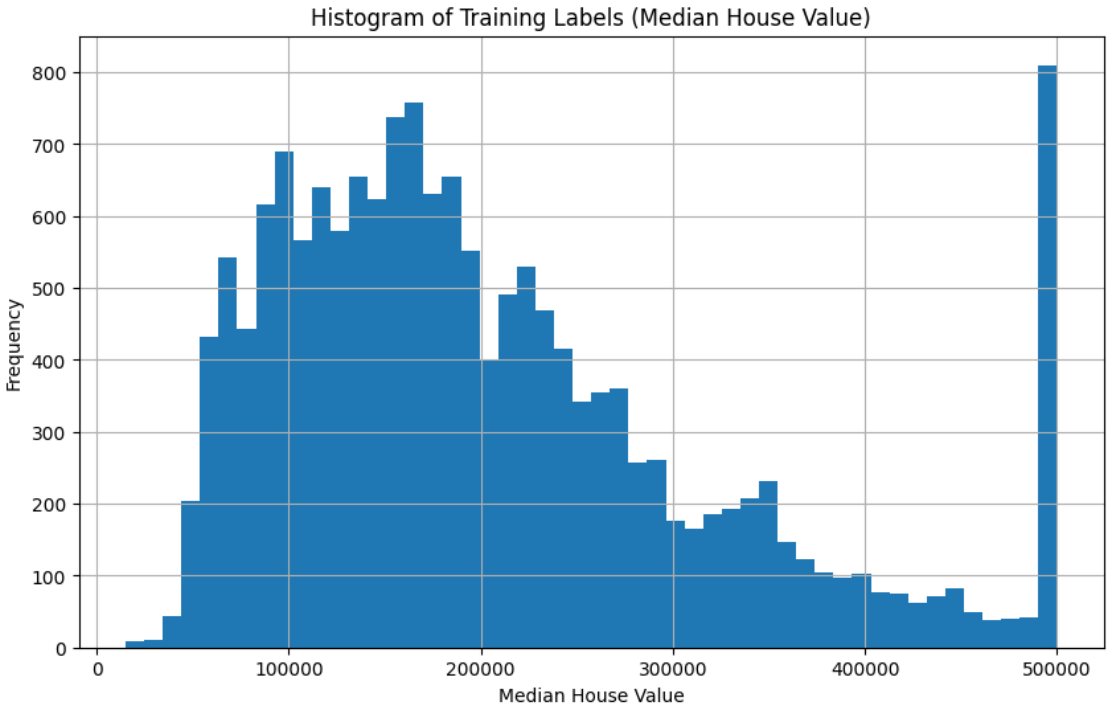
Task 2

[40]:
```
print(f"Shape of train_features: {train_features.shape}")

print("Missing values in train_features:")
print(train_features.isnull().sum())

print("Statistics summary of train_features:")
print(train_features.describe())
```

```
Shape of train_features: (16346, 8)
Missing values in train_features:
longitude             0
latitude              0
housing_median_age    0
total_rooms           0
total_bedrooms        0
population            0
households            0
median_income         0
dtype: int64
Statistics summary of train_features:
          longitude      latitude  housing_median_age   total_rooms  \
count  16346.000000  16346.000000        16346.000000  16346.000000
mean    -119.576731     35.639888           28.606265   2630.196745
std        2.006389      2.139648           12.632219   2158.731496
min     -124.350000     32.540000            1.000000      2.000000
25%     -121.800000     33.930000           18.000000   1451.000000
50%     -118.500000     34.260000           29.000000   2125.500000
75%     -118.010000     37.720000           37.000000   3137.000000
max     -114.310000     41.950000           52.000000  39320.000000

       total_bedrooms    population    households  median_income
count    16346.000000  16346.000000  16346.000000   16346.000000
mean       537.625413   1425.149456    498.992659       3.864135
std        418.511077   1136.460782    378.999712       1.892231
min          1.000000      3.000000      1.000000       0.499900
25%        296.000000    785.000000    279.000000       2.559275
50%        435.000000   1165.500000    409.000000       3.533800
75%        647.000000   1728.000000    605.000000       4.739900
max       6445.000000  35682.000000   6082.000000      15.000100
```

[41]:
```
print(train_labels.head())

import matplotlib.pyplot as plt

train_labels.hist(bins=50, figsize=(10, 6))
plt.xlabel("Median House Value")
plt.ylabel("Frequency")
plt.title("Histogram of Training Labels (Median House Value)")
plt.show()
```

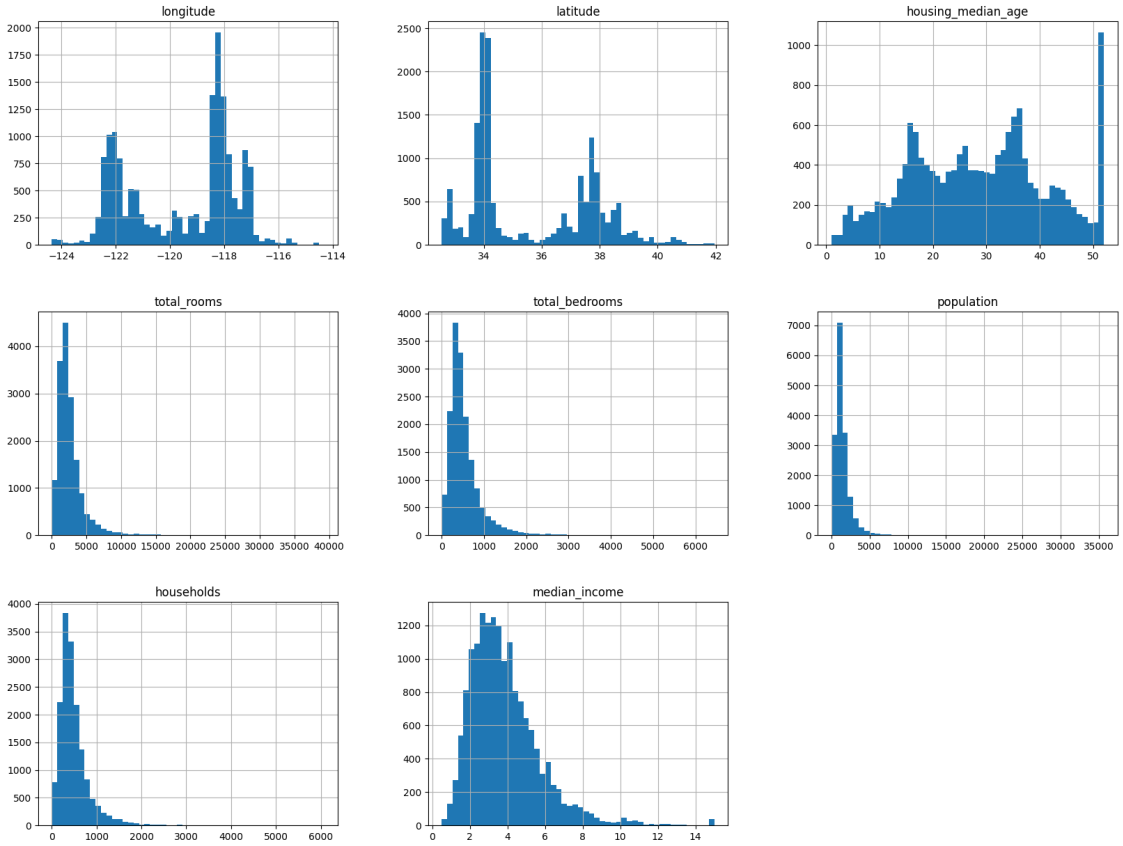
```
12346    145200.0
19326    117000.0
16824    263900.0
6869     163700.0
16677    236100.0
Name: median_house_value, dtype: float64
```



[42]:
```
train_features.hist(bins=50, figsize=(20,15))
```

[42]:
```
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: >]],
dtype=object)
```



[43]:
```
train_features.describe()
```

[43]:

	longitude	latitude	housing_median_age	total_rooms
count	16346.000000	16346.000000	16346.000000	16346.000000
mean	-119.576731	35.639888	28.606265	2630.196745
std	2.006389	2.139648	12.632219	2158.731496
min	-124.350000	32.540000	1.000000	2.000000
25%	-121.800000	33.930000	18.000000	1451.000000
50%	-118.500000	34.260000	29.000000	2125.500000
75%	-118.010000	37.720000	37.000000	3137.000000
max	-114.310000	41.950000	52.000000	39320.000000

	total_bedrooms	population	households	median_income
count	16346.000000	16346.000000	16346.000000	16346.000000
mean	537.625413	1425.149456	498.992659	3.864135

```
std        418.511077   1136.460782    378.999712       1.892231
min          1.000000      3.000000      1.000000       0.499900
25%        296.000000    785.000000    279.000000       2.559275
50%        435.000000   1165.500000    409.000000       3.533800
75%        647.000000   1728.000000    605.000000       4.739900
max       6445.000000  35682.000000   6082.000000      15.000100
```

[44]:
```
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler() ## define the transformer
scaler.fit(train_features) ## call .fit() method to calculate the min and max␣
 ↪value for each column in datas
```

[44]:
```
MinMaxScaler()
```

[45]:
```
print("Min of each column: ",scaler.data_min_)
print("Max of each column: ",scaler.data_max_)
```

```
Min of each column:  [-124.35     32.54      1.         2.         1.         3.
1.
    0.4999]
Max of each column:  [-1.14310e+02  4.19500e+01  5.20000e+01  3.93200e+04
6.44500e+03
  3.56820e+04  6.08200e+03  1.50001e+01]
```

[46]:
```
train_features_normalized = scaler.transform(train_features)
train_features_normalized
```

[46]:
```
array([[0.77988048, 0.1360255 , 0.39215686, …, 0.10146024, 0.28498602,
        0.13939808],
       [0.14043825, 0.63336876, 0.88235294, …, 0.02253426, 0.06002302,
        0.14701177],
       [0.18525896, 0.54091392, 0.58823529, …, 0.04117268, 0.10146358,
        0.32103695],
       …,
       [0.25498008, 0.50797024, 0.09803922, …, 0.06045573, 0.09537905,
        0.34714694],
       [0.60956175, 0.15302869, 0.66666667, …, 0.06045573, 0.08156553,
        0.11765355],
       [0.24601594, 0.50584485, 0.29411765, …, 0.04683427, 0.11182371,
        0.15009448]])
```

0.4 Question 9

Task 1

[48]:
```
# Assuming train_features_normalized is your normalized dataset

import matplotlib.pyplot as plt
import pandas as pd
```

```
# Convert the normalized NumPy array back to a DataFrame for plotting
normalized_df = pd.DataFrame(train_features_normalized, columns=train_features.
  ↪columns)

# Plot histograms for each normalized feature
normalized_df.hist(bins=50, figsize=(20, 15))
plt.show()
```