```
In [1]: import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

In [2]: #librairies de data viz
import matplotlib.pyplot as plt
import seaborn as sns
librairies des modèle linéaire
from sklearn import linear_model
#librairie des modèle d'ensemble
from sklearn.ensemble import RandomForestRegressor, AdaBoostRegressor
librairies des métriques
from sklearn.metrics import mean_absolute_error
from sklearn.metrics import mean_squared_error, r2_score
librairies pour le k-fold cross validation
from sklearn.model_selection import train_test_split

In [4]: df = pd.read_csv("C:/Users/99210/Downloads/dataset.csv")

In [5]: df.head()

Out[5]:

	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	sqft_above	sqft_basement	yr_built
	02- 05- 2014 00:00	313000.0	3	1.50	1340	7912	1.5	0	0	3	1340	0	1955
	02- 05- 2014 00:00	2384000.0	5	2.50	3650	9050	2.0	0	4	5	3370	280	1921
	02- 05- 2014 00:00	342000.0	3	2.00	1930	11947	1.0	0	0	4	1930	0	1966
	02- 05- 2014 00:00	420000.0	3	2.25	2000	8030	1.0	0	0	4	1000	1000	1963
,	02- 05- 2014 00:00	550000.0	4	2.50	1940	10500	1.0	0	0	4	1140	800	1976
4													•

```
print('Nbr de lignes et nbr de colonnes : ',df.shape)
In [6]:
        print(df.dtypes)
        Nbr de lignes et nbr de colonnes : (4600, 18)
                          object
        date
        price
                          float64
        bedrooms
                            int64
        bathrooms
                          float64
        sqft_living
                            int64
        sqft_lot
                            int64
        floors
                          float64
        waterfront
                            int64
        view
                            int64
        condition
                            int64
        sqft_above
                            int64
        sqft_basement
                            int64
        yr_built
                            int64
        yr_renovated
                           int64
        street
                           object
                           object
        city
        statezip
                           object
        country
                           object
        dtype: object
In [7]: df[['floors', 'bathrooms', 'bedrooms', 'price']] = df[['floors', 'bathrooms', 'bedrooms', 'price']].astype('int
        df.dtypes
Out[7]: date
                          object
                           int32
        price
        bedrooms
                           int32
                           int32
        bathrooms
        sqft_living
                           int64
        sqft_lot
                           int64
        floors
                           int32
        waterfront
                           int64
        view
                           int64
        condition
                           int64
        sqft above
                           int64
        sqft basement
                           int64
                           int64
        yr_built
                           int64
        yr_renovated
                          object
        street
        city
                          object
                          object
        statezip
        country
                          object
        dtype: object
In [8]: df.bathrooms.dtype
Out[8]: dtype('int32')
```

```
In [9]: df.isnull().sum()
Out[9]: date
                          0
         price
                          0
         bedrooms
                          0
         bathrooms
                          0
         sqft_living
                          0
                          0
         sqft_lot
                          0
         floors
         waterfront
                          0
         view
                          0
         condition
                          0
         sqft_above
                          0
         sqft_basement
                          0
         yr_built
                          0
         yr_renovated
                          0
         street
                          0
         city
                          0
         statezip
                          0
                          0
         country
         dtype: int64
In [10]: df.describe()
```

Out[10]:

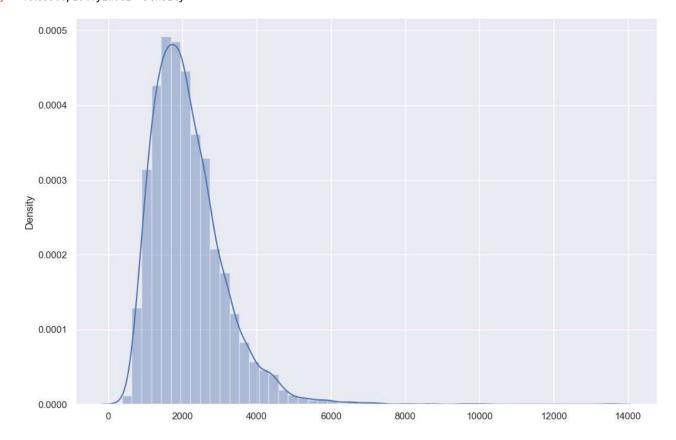
	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	sqft_
count	4.600000e+03	4600.000000	4600.000000	4600.000000	4.600000e+03	4600.000000	4600.000000	4600.000000	4600.000000	4600.0
mean	5.519630e+05	3.400870	1.788913	2139.346957	1.485252e+04	1.459130	0.007174	0.240652	3.451739	1827.2
std	5.638347e+05	0.908848	0.752185	963.206916	3.588444e+04	0.552194	0.084404	0.778405	0.677230	862
min	0.000000e+00	0.000000	0.000000	370.000000	6.380000e+02	1.000000	0.000000	0.000000	1.000000	370.0
25%	3.228750e+05	3.000000	1.000000	1460.000000	5.000750e+03	1.000000	0.000000	0.000000	3.000000	1190.0
50%	4.609430e+05	3.000000	2.000000	1980.000000	7.683000e+03	1.000000	0.000000	0.000000	3.000000	1590.0
75%	6.549625e+05	4.000000	2.000000	2620.000000	1.100125e+04	2.000000	0.000000	0.000000	4.000000	2300.0
max	2.659000e+07	9.000000	8.000000	13540.000000	1.074218e+06	3.000000	1.000000	4.000000	5.000000	9410.0
4										•

```
sns.set(rc={'figure.figsize':(12,8)})
In [11]:
           sns.heatmap(df.corr(), annot=True,cmap='RdBu')
           plt.show()
                                                                                                                                          1.0
                                                    0.43
                                                           0.05
                                                                                  0.23
                                                                                         0.035
                                                                                                         0.21
                                                                                                                0.022
                     price
                                     0.2
                                            0.32
                                                                   0.14
                                                                           0.14
                                                                                                 0.37
                bedrooms
                              0.2
                                             0.5
                                                    0.59
                                                           0.069
                                                                   0.16
                                                                         -0.0035
                                                                                  0.11
                                                                                         0.025
                                                                                                 0.48
                                                                                                         0.33
                                                                                                                0.14
                                                                                                                                         - 0.8
                bathrooms
                             0.32
                                     0.5
                                                    0.71
                                                           0.11
                                                                   0.46
                                                                          0.072
                                                                                   0.2
                                                                                          -0.12
                                                                                                         0.27
                                                                                                                0.39
                                                                                                                        -0.19
                                     0.59
                                                           0.21
                                                                                                 0.88
                                                                                                                0.29
                 sqft_living
                             0.43
                                                                   0.34
                                                                                  0.31
                                                                                                         0.45
                                                                                                                        -0.12
                                                                           0.12
                                                                                                                                          0.6
                   sqft_lot
                             0.05
                                    0.069
                                            0.11
                                                    0.21
                                                                  -0.0051 0.017 0.074 0.00056
                                                                                                 0.22
                                                                                                        0.035
                                                                                                                0.051
                     floors
                             0.14
                                     0.16
                                            0.46
                                                    0.34
                                                          -0.0051
                                                                          0.018 0.021
                                                                                          -0.31
                                                                                                 0.52
                                                                                                         -0.25
                                                                                                                0.56
                                                                                                                        -0.25
                                                                                                                                        -0.4
                 waterfront
                             0.14
                                   -0.0035 0.072
                                                           0.017 0.018
                                                                                  0.36 0.00035 0.079
                                                                                                        0.098
                                                                                                                0.024 0.0086
                                                    0.12
                                                                          0.36
                                                                                                                                        -0.2
                             0.23
                                     0.11
                                             0.2
                                                    0.31
                                                           0.074
                                                                 0.021
                                                                                         0.063
                                                                                                 0.17
                                                                                                         0.32
                                                                                                                       0.023
                      view
                  condition
                             0.035
                                    0.025
                                                   -0.063 0.00056
                                                                   -0.31
                                                                         0.00035 0.063
                                                                                                 -0.18
                                                                                                                 -0.4
                                                                                                                        -0.19
                                                                                                                                        - 0.0
                                                    0.88
                sqft_above
                             0.37
                                     0.48
                                                           0.22
                                                                   0.52
                                                                          0.079
                                                                                  0.17
                                                                                          -0.18
                                                                                                                0.41
                                                                                                                        -0.16
                                                                                                                -0.16
            sqft_basement
                             0.21
                                    0.33
                                            0.27
                                                    0.45
                                                           0.035
                                                                   -0.25
                                                                          0.098
                                                                                  0.32
                                                                                          0.2
                                                                                                                        0.043
                                                                                                                                        - -0.2
                             0.022
                                    0.14
                                            0.39
                                                    0.29
                   yr_built
                                                           0.051
                                                                   0.56
                                                                                          -0.4
                                                                                                 0.41
                                                                                                         -0.16
                                                                                                                        -0.32
                                            -0.19
                                                    -0.12
                                                                   -0.25
                                                                         0.0086 0.023
                                                                                          -0.19
                                                                                                 -0.16
                                                                                                                -0.32
              yr_renovated
                                                                                                        0.043
                                                                                                                 yr_built
                                                    sqft_living
                                                                                   view
                                      pedrooms
                                                                                           condition
                                                                                                  sqft_above
                                                                                                          basemen
                                                                                                                         renovated
                                                            adft
In [13]: | df['price'].replace(0,np.nan, inplace = True)
           df.fillna(df['price'].mean(),inplace=True)
In [14]: |df.drop(['sqft_above','sqft_lot','yr_built','yr_renovated','condition'], axis=1,inplace=True)
In [15]: df.columns
Out[15]: Index(['date', 'price', 'bedrooms', 'bathrooms', 'sqft_living', 'floors',
                    'waterfront', 'view', 'sqft_basement', 'street', 'city', 'statezip',
                    'country'],
                  dtype='object')
```

In [16]: sns.distplot(x = df.sqft_living)

C:\Users\99210\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep recated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms). warnings.warn(msg, FutureWarning)

Out[16]: <AxesSubplot:ylabel='Density'>

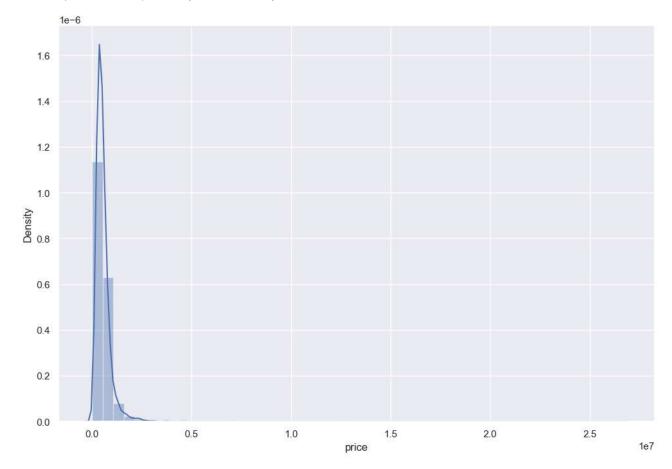


In [17]: sns.distplot(df.price)

C:\Users\99210\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep recated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

Out[17]: <AxesSubplot:xlabel='price', ylabel='Density'>

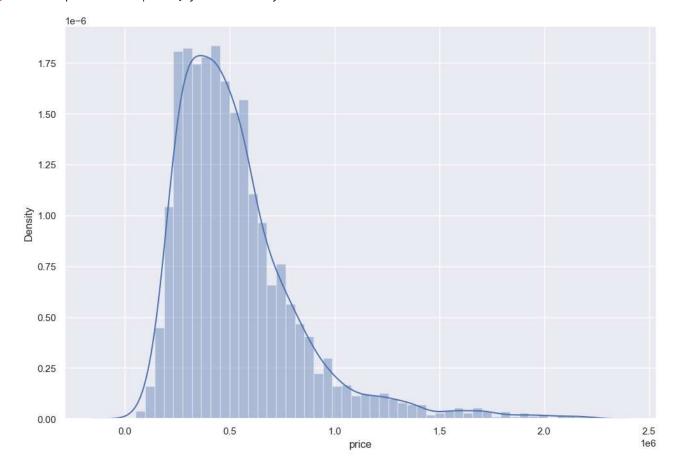


In [18]: from scipy import stats
df['price'] = df['price'].replace([df['price'][np.abs(stats.zscore(df['price'])) > 3]],np.median(df['price'])

In [19]: sns.distplot(df.price)

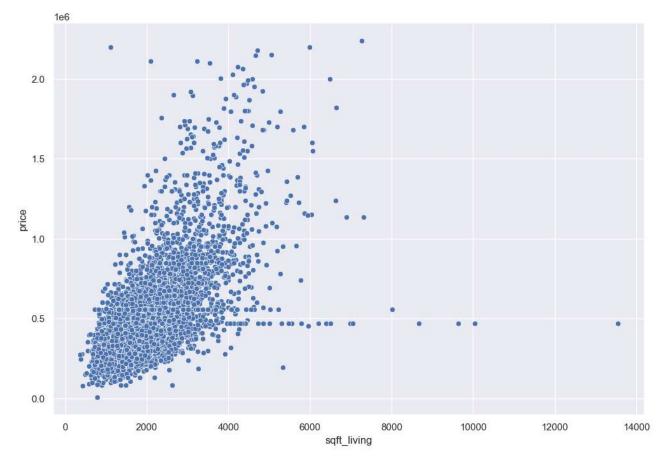
C:\Users\99210\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep recated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms). warnings.warn(msg, FutureWarning)

Out[19]: <AxesSubplot:xlabel='price', ylabel='Density'>



```
In [20]: sns.scatterplot(data = df, x='sqft_living',y='price')
```

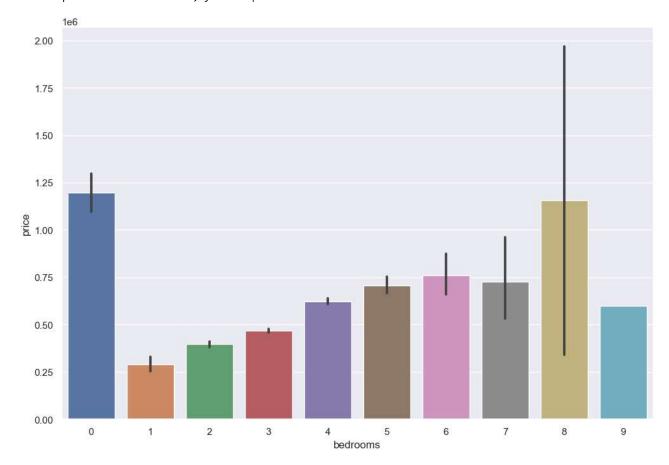
Out[20]: <AxesSubplot:xlabel='sqft_living', ylabel='price'>



```
In [21]: df.sqft_living.describe()
Out[21]: count
                   4600.000000
                   2139.346957
         mean
         std
                    963.206916
                    370.000000
         min
                   1460.000000
         25%
         50%
                   1980.000000
         75%
                   2620.000000
                  13540.000000
         max
         Name: sqft_living, dtype: float64
In [22]: df['sqft_living'] = np.where((df.sqft_living >6000 ), 6000, df.sqft_living)
```

In [23]: sns.barplot(x=df.bedrooms , y = df.price)

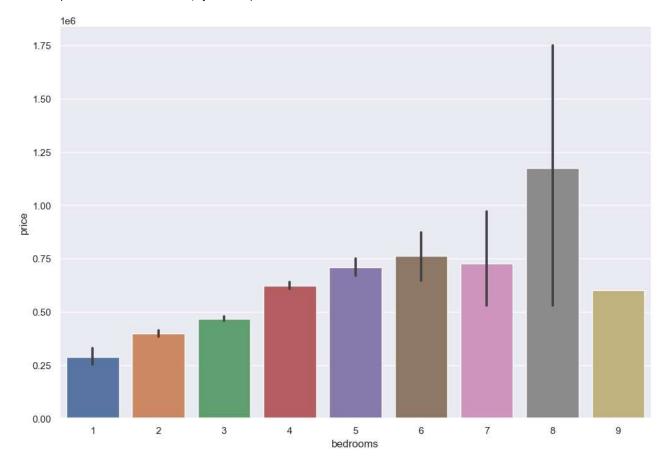
Out[23]: <AxesSubplot:xlabel='bedrooms', ylabel='price'>



In [24]: df.bedrooms.replace(0, 8,inplace = True)

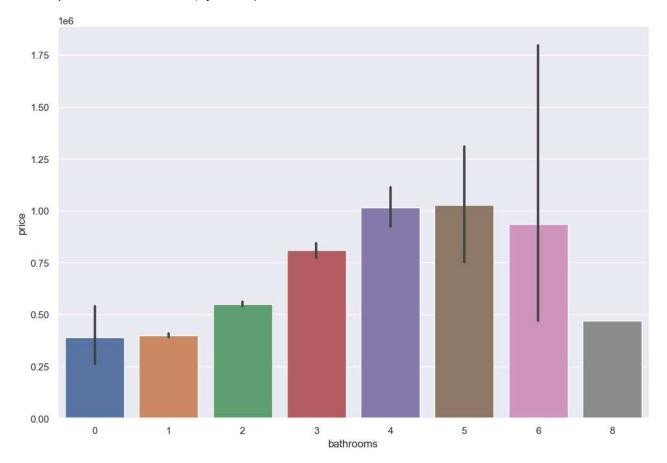
In [25]: sns.barplot(x=df.bedrooms , y = df.price)

Out[25]: <AxesSubplot:xlabel='bedrooms', ylabel='price'>



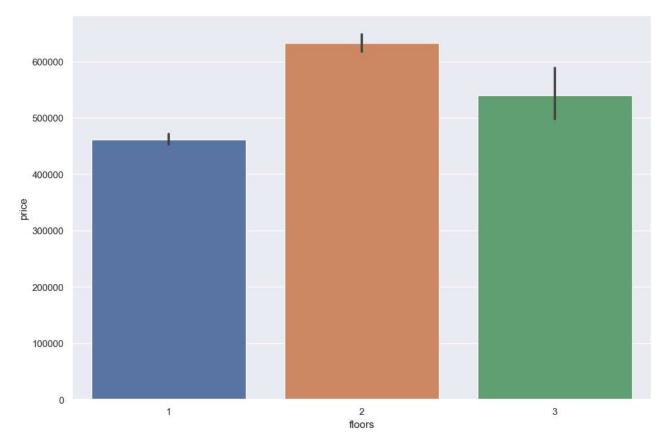
In [27]: sns.barplot(x=df.bathrooms , y = df.price)

Out[27]: <AxesSubplot:xlabel='bathrooms', ylabel='price'>



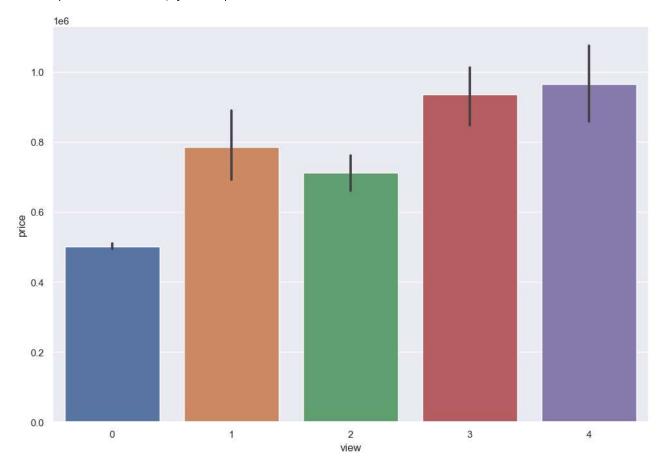
```
In [28]: sns.barplot(x=df.floors , y = df.price)
```

Out[28]: <AxesSubplot:xlabel='floors', ylabel='price'>



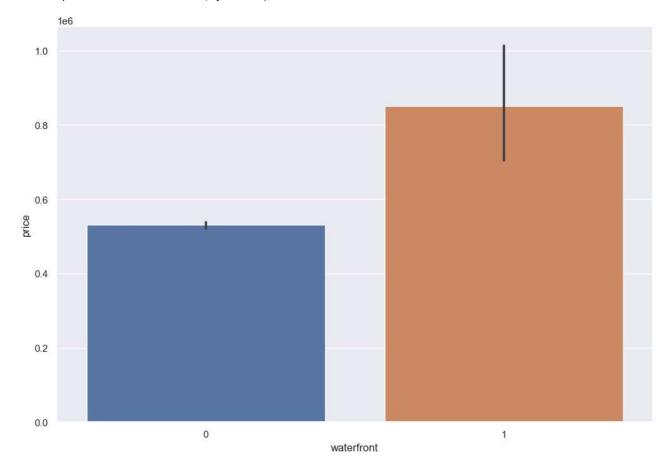
In [29]: sns.barplot(x=df.view , y = df.price)

Out[29]: <AxesSubplot:xlabel='view', ylabel='price'>



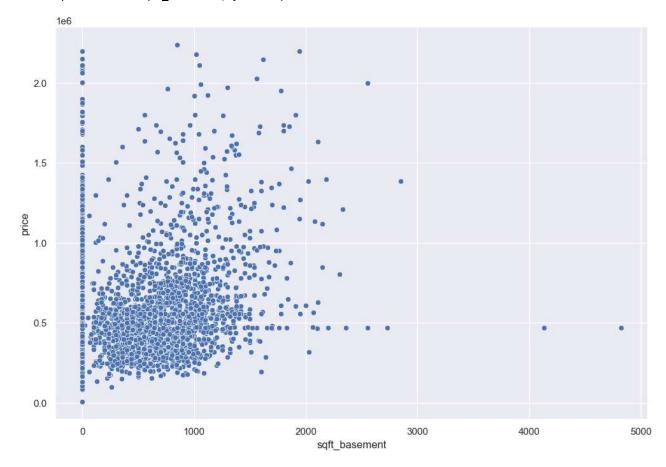
In [30]: sns.barplot(x=df.waterfront , y = df.price)

Out[30]: <AxesSubplot:xlabel='waterfront', ylabel='price'>



In [31]: sns.scatterplot(x='sqft_basement', y ='price',data= df)

Out[31]: <AxesSubplot:xlabel='sqft_basement', ylabel='price'>



In [32]: df['sqft_basement'] = np.where((df.sqft_basement >2000), 2000, df.sqft_basement)

In [33]: df.head()

Out[33]:

	date	price	bedrooms	bathrooms	sqft_living	floors	waterfront	view	sqft_basement	street	city	statezip	country
(02- 05- 2014 00:00	313000.0	3	1	1340	1	0	0	0	18810 Densmore Ave N	Shoreline	WA 98133	USA
	02- 05- 2014 00:00	468750.0	5	2	3650	2	0	4	280	709 W Blaine St	Seattle	WA 98119	USA
:	02- 05- 2014 00:00	342000.0	3	2	1930	1	0	0	0	26206- 26214 143rd Ave SE	Kent	WA 98042	USA
;	02- 05- 2014 00:00	420000.0	3	2	2000	1	0	0	1000	857 170th PI NE	Bellevue	WA 98008	USA
	02- 05- 2014 00:00	550000.0	4	2	1940	1	0	0	800	9105 170th Ave NE	Redmond	WA 98052	USA

```
In [34]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 4600 entries, 0 to 4599
         Data columns (total 13 columns):
          # Column
                             Non-Null Count Dtype
              -----
                             -----
          a
                             4600 non-null
              date
                                             object
              price
                             4600 non-null
                                             float64
          1
                             4600 non-null
                                             int32
              bedrooms
          3
              bathrooms
                             4600 non-null
                                             int32
              sqft_living
                             4600 non-null
                                            int64
              floors
                             4600 non-null
                                            int32
              waterfront
                             4600 non-null
                                            int64
                             4600 non-null
                                            int64
              view
          8
              sqft basement 4600 non-null
                                            int64
                             4600 non-null
                                             object
          9
              street
          10 city
                             4600 non-null
                                             object
                             4600 non-null
          11 statezip
                                            object
          12 country
                             4600 non-null
                                            object
         dtypes: float64(1), int32(3), int64(4), object(5)
         memory usage: 413.4+ KB
In [35]: | df = pd.get_dummies(df, columns=['city'], prefix = ['city'])
         X = df.drop(["date",'street', 'statezip', 'country','price'], axis=1)
         y = df[['price']]
In [36]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
In [37]: | 1r = linear_model.LinearRegression()
         lr.fit(X_train, y_train) #phase d'apprentissage
         print(lr.score(X train, y train))
         print(lr.score(X_test,y_test))
         0.6146718898676132
         0.5659003045853999
In [38]: | 1 = linear_model.Lasso()
         1.fit(X_train, y_train) #phase d'apprentissage
         predl = 1.predict(X test)
         scorel=r2_score(y_test,pred1)
         scorel
         C:\Users\99210\anaconda3\lib\site-packages\sklearn\linear_model\_coordinate_descent.py:647: ConvergenceWarni
         ng: Objective did not converge. You might want to increase the number of iterations, check the scale of the
         features or consider increasing regularisation. Duality gap: 2.025e+13, tolerance: 2.847e+10
           model = cd fast.enet coordinate descent(
Out[38]: 0.5661901911100673
In [39]: r = linear_model.Ridge()
         r.fit(X_train, y_train) #phase d'apprentissage
         predr = r.predict(X test)
         scorer=r2 score(y test,predr)
         scorer
Out[39]: 0.5638948252673996
```

```
In [40]: | 1r = linear model.LinearRegression()
         scores = cross_val_score(lr, X, y, cv=5)
         print("score", scores)
         print("moyenne :%0.03f , deviation: :%0.03f" % (scores.mean(), scores.std()))
         score [0.64991269 0.61701447 0.63198355 0.65896884 0.3101221 ]
         moyenne :0.574 , deviation: :0.133
In [41]: | from sklearn.ensemble import RandomForestRegressor
         # Define the model. Set random_state to 1
         rf_model = RandomForestRegressor()
         # fit your model
         rf_model.fit(X_train, y_train)
         rf_model.score(X_train, y_train)
         C:\Users\99210\AppData\Local\Temp\ipykernel_30764\2626062114.py:7: DataConversionWarning: A column-vector y
         was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using rav
         el().
           rf_model.fit(X_train, y_train)
Out[41]: 0.9351156289941331
In [42]: rf = RandomForestRegressor()
         scores = cross_val_score(rf, X, y, cv=5)
         print("score", scores)
         print("moyenne :%0.03f , deviation: :%0.03f" % (scores.mean(), scores.std()))
         C:\Users\99210\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:680: DataConversionWarnin
         g: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,),
         for example using ravel().
           estimator.fit(X train, y train, **fit params)
         C:\Users\99210\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:680: DataConversionWarnin
         g: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,),
         for example using ravel().
           estimator.fit(X_train, y_train, **fit_params)
         C:\Users\99210\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:680: DataConversionWarnin
         g: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,),
         for example using ravel().
           estimator.fit(X train, y train, **fit params)
         C:\Users\99210\anaconda3\lib\site-packages\sklearn\model selection\ validation.py:680: DataConversionWarnin
         g: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,),
         for example using ravel().
           estimator.fit(X_train, y_train, **fit_params)
         C:\Users\99210\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:680: DataConversionWarnin
         g: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n samples,),
         for example using ravel().
           estimator.fit(X_train, y_train, **fit_params)
         score [0.61341309 0.61893321 0.61335355 0.58788427 0.34777901]
         moyenne: 0.556, deviation:: 0.105
```

In [43]:	rf_model = RandomForestRegressor(n_estimators=5, max_depth=5)										
	# fit your model										
	rf_model.fit(X_train, y_train)										
	<pre>print(rf_model.score(X_train, y_train)) print(rf model.score(X test, y test))</pre>										
	0.6045465290011636										
	0.4986676277428299										
	C.\\\\00040\AP-t-\\\\\\\\\-\\\										
	C:\Users\99210\AppData\Local\Temp\ipykernel_30764\1519264609.py:4: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n samples,), for example using ray										
	el().										
	rf_model.fit(X_train, y_train)										
In [*]:											
In []:											
±11 [].											