assigment2MLDat158

November 15, 2023

1 House pricing prediction

```
[1]: #biblotek

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import RandomizedSearchCV
```

2 Begynner med å importere dataen.

```
[3]: trainingData=pd.read_csv('train.csv')
     trainingData.head()
[3]:
            MSSubClass MSZoning
                                  LotFrontage
                                                 LotArea Street Alley LotShape
     0
         1
                     60
                               RL
                                           65.0
                                                     8450
                                                             Pave
                                                                    NaN
                                                                              Reg
         2
                               RL
     1
                     20
                                           0.08
                                                     9600
                                                             Pave
                                                                    NaN
                                                                              Reg
     2
         3
                     60
                               RL
                                           68.0
                                                    11250
                                                             Pave
                                                                    NaN
                                                                              IR1
     3
         4
                     70
                               RL
                                           60.0
                                                     9550
                                                             Pave
                                                                    NaN
                                                                              IR1
                     60
                               RL
                                           84.0
                                                    14260
                                                             Pave
                                                                              IR1
                                                                    NaN
       LandContour Utilities
                                ... PoolArea PoolQC Fence MiscFeature MiscVal MoSold
     0
                Lvl
                       AllPub
                                          0
                                               NaN
                                                      NaN
                                                                   NaN
                                                                              0
                                                                                      2
     1
                Lvl
                       AllPub
                                          0
                                                                              0
                                                                                      5
                                                NaN
                                                      NaN
                                                                   NaN
     2
                                                                              0
                                                                                      9
                Lvl
                       AllPub ...
                                          0
                                                                   NaN
                                                NaN
                                                      NaN
     3
                Lvl
                                                                                      2
                       AllPub
                                          0
                                                NaN
                                                      NaN
                                                                   NaN
                                                                              0
     4
                Lvl
                       AllPub
                                                NaN
                                                                              0
                                                                                     12
                                                      NaN
                                                                   NaN
       YrSold
                SaleType
                           SaleCondition SalePrice
         2008
                                  Normal
                                              208500
     0
                      WD
         2007
     1
                      WD
                                  Normal
                                              181500
     2
         2008
                      WD
                                  Normal
                                              223500
         2006
                                 Abnorml
     3
                      WD
                                              140000
         2008
                      WD
                                  Normal
                                              250000
```

[5 rows x 81 columns]

```
[4]: testData = pd.read_csv('test.csv')
     testData.head()
[4]:
               MSSubClass MSZoning
                                      LotFrontage LotArea Street Alley LotShape \
        1461
                                              80.0
     0
                        20
                                  RH
                                                       11622
                                                                Pave
                                                                        NaN
                                                                                  Reg
     1
        1462
                        20
                                  RL
                                              81.0
                                                       14267
                                                                        NaN
                                                                Pave
                                                                                  IR1
                                  RL
                                                       13830
       1463
                        60
                                              74.0
                                                                        {\tt NaN}
                                                                Pave
                                                                                  IR1
     3 1464
                                  RL
                        60
                                              78.0
                                                        9978
                                                                Pave
                                                                        NaN
                                                                                  IR1
        1465
                       120
                                  RL
                                              43.0
                                                        5005
                                                                Pave
                                                                        NaN
                                                                                  IR1
       LandContour Utilities
                                 ... ScreenPorch PoolArea PoolQC
                                                                   Fence MiscFeature
     0
                                            120
                                                        0
                                                                   MnPrv
                Lvl
                        AllPub
                                                              NaN
                                                                                   NaN
     1
                Lvl
                        AllPub
                                              0
                                                        0
                                                              NaN
                                                                                  Gar2
                                                                      NaN
     2
                Lvl
                        AllPub
                                              0
                                                        0
                                                              NaN
                                                                   MnPrv
                                                                                   NaN
     3
                Lvl
                        AllPub
                                              0
                                                        0
                                                              NaN
                                                                      NaN
                                                                                   NaN
                HLS
                        AllPub
                                            144
                                                              NaN
                                                                      NaN
                                                                                   NaN
       MiscVal MoSold YrSold
                                  SaleType
                                             SaleCondition
     0
                           2010
              0
                      6
                                         WD
                                                     Normal
         12500
                           2010
     1
                      6
                                         WD
                                                     Normal
     2
              0
                           2010
                                         WD
                                                     Normal
                      3
     3
              0
                      6
                           2010
                                         WD
                                                     Normal
     4
              0
                      1
                           2010
                                         WD
                                                     Normal
```

3 Begynner med å ryddde opp i dataen

[6]: trainingData.info()

[5 rows x 80 columns]

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1460 entries, 0 to 1459
Data columns (total 81 columns):

#	Column	Non-Null Count	Dtype
0	Id	1460 non-null	int64
1	MSSubClass	1460 non-null	int64
2	MSZoning	1460 non-null	object
3	LotFrontage	1201 non-null	float64
4	LotArea	1460 non-null	int64
5	Street	1460 non-null	object
6	Alley	91 non-null	object
7	LotShape	1460 non-null	object
8	LandContour	1460 non-null	object
9	Utilities	1460 non-null	object
10	LotConfig	1460 non-null	object
11	LandSlope	1460 non-null	object

12	Neighborhood	1460	non-null	object
13	Condition1	1460	non-null	object
14	Condition2	1460	non-null	object
15	BldgType	1460	non-null	object
16	HouseStyle	1460	non-null	object
17	OverallQual	1460	non-null	int64
18	OverallCond	1460	non-null	int64
19	YearBuilt	1460	non-null	int64
20	YearRemodAdd	1460	non-null	int64
21	RoofStyle	1460	non-null	object
22	RoofMatl	1460	non-null	object
23	Exterior1st	1460	non-null	object
24	Exterior2nd	1460	non-null	object
25	MasVnrType	588 r	non-null	object
26	MasVnrArea	1452	non-null	float64
27	ExterQual	1460	non-null	object
28	ExterCond	1460	non-null	object
29	Foundation	1460	non-null	object
30	BsmtQual	1423	non-null	object
31	BsmtCond	1423	non-null	object
32	BsmtExposure	1422	non-null	object
33	BsmtFinType1	1423	non-null	object
34	BsmtFinSF1	1460	non-null	int64
35	BsmtFinType2	1422	non-null	object
36	BsmtFinSF2	1460	non-null	int64
37	BsmtUnfSF	1460	non-null	int64
38	TotalBsmtSF	1460	non-null	int64
39	Heating	1460	non-null	object
40	HeatingQC	1460	non-null	object
41	CentralAir	1460	non-null	object
42	Electrical	1459	non-null	object
43	1stFlrSF	1460	non-null	int64
44	2ndFlrSF	1460	non-null	int64
45	LowQualFinSF	1460	non-null	int64
46	GrLivArea	1460	non-null	int64
47	BsmtFullBath	1460	non-null	int64
48	BsmtHalfBath	1460	non-null	int64
49	FullBath	1460	non-null	int64
50	HalfBath	1460	non-null	int64
51	BedroomAbvGr	1460	non-null	int64
52	KitchenAbvGr	1460	non-null	int64
53	KitchenQual	1460	non-null	object
54	TotRmsAbvGrd	1460	non-null	int64
55	Functional	1460	non-null	object
56	Fireplaces	1460	non-null	int64
57	FireplaceQu		non-null	object
58	GarageType	1379	non-null	object
59	GarageYrBlt	1379		float64
	_			

```
GarageFinish
                    1379 non-null
                                     object
 60
                                     int64
 61
     GarageCars
                    1460 non-null
 62
     GarageArea
                    1460 non-null
                                     int64
 63
     GarageQual
                    1379 non-null
                                     object
     GarageCond
                    1379 non-null
 64
                                     object
     PavedDrive
                    1460 non-null
                                     object
 66
     WoodDeckSF
                    1460 non-null
                                     int64
 67
     OpenPorchSF
                    1460 non-null
                                     int64
    EnclosedPorch
                    1460 non-null
                                     int64
     3SsnPorch
                    1460 non-null
 69
                                     int64
 70
     ScreenPorch
                    1460 non-null
                                     int64
 71 PoolArea
                    1460 non-null
                                     int64
 72 PoolQC
                    7 non-null
                                     object
                    281 non-null
 73
    Fence
                                     object
 74 MiscFeature
                    54 non-null
                                     object
    MiscVal
                    1460 non-null
                                     int64
 76
    MoSold
                    1460 non-null
                                     int64
 77
    YrSold
                    1460 non-null
                                     int64
 78
     SaleType
                    1460 non-null
                                     object
 79
     SaleCondition
                    1460 non-null
                                     object
 80
     SalePrice
                    1460 non-null
                                     int64
dtypes: float64(3), int64(35), object(43)
memory usage: 924.0+ KB
```

Dropper kolonner der nullverdiene utgjør mer en 50% av alle objektene. Så foreks kolonner som alley, PoolQc, Fence, og MiscFeature har veldig mangen null-verdier osm vil gå utover resultatet

```
[7]: trainingData.drop(['Alley', 'PoolQC', 'Fence', 

'MiscFeature', 'MasVnrType'], axis=1, inplace=True)

trainingData.drop(['Id'], axis=1, inplace=True)
```

Så må vi erstatte nullverdier i de andre kolonnene. Måten vi erstatter dem kommer litt ann på hvilke type kolonnen har. Med tall tar vi ofte bare å erstatter null-veridene med gjennomsnittet. Med kategoriske verdier er det kanskje lurere å erstatte dem med den mest hyppige typen.

Erstatter med gjennomsnitt:

```
[8]: for column in trainingData.columns:
    if trainingData[column].dtype in ['int64', 'float64']:
        column_mean = trainingData[column].mean()
        trainingData[column].fillna(column_mean, inplace=True)
```

Erstatter med hyppigeste type

```
[9]: for column in trainingData.columns:
    if trainingData[column].dtype == 'object':
        trainingData[column].fillna(trainingData[column].mode()[0],__
        inplace=True)
```

[10]: trainingData.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1460 entries, 0 to 1459
Data columns (total 75 columns):

Data	columns (total	(5 columns):	
#	Column	Non-Null Count	Dtype
0	MSSubClass	1460 non-null	int64
1	MSZoning	1460 non-null	object
2	LotFrontage	1460 non-null	float64
3	LotArea	1460 non-null	int64
4	Street	1460 non-null	object
5	LotShape	1460 non-null	object
6	LandContour	1460 non-null	object
7	Utilities	1460 non-null	object
8	LotConfig	1460 non-null	object
9	LandSlope	1460 non-null	object
10	Neighborhood	1460 non-null	object
11	Condition1	1460 non-null	object
12	Condition2	1460 non-null	object
13	BldgType	1460 non-null	object
14	HouseStyle	1460 non-null	object
15	OverallQual	1460 non-null	int64
16	OverallCond	1460 non-null	int64
17	YearBuilt	1460 non-null	int64
18	YearRemodAdd	1460 non-null	int64
19	RoofStyle	1460 non-null	object
20	RoofMatl	1460 non-null	object
21	Exterior1st	1460 non-null	object
22	Exterior2nd	1460 non-null	object
23	MasVnrArea	1460 non-null	float64
24	ExterQual	1460 non-null	object
25	ExterCond	1460 non-null	object
26	Foundation	1460 non-null	object
27	BsmtQual	1460 non-null	object
28	BsmtCond	1460 non-null	object
29	BsmtExposure	1460 non-null	object
30	BsmtFinType1	1460 non-null	object
31	BsmtFinSF1	1460 non-null	int64
32	BsmtFinType2	1460 non-null	object
33	BsmtFinSF2	1460 non-null	int64
34	BsmtUnfSF	1460 non-null	int64
35	TotalBsmtSF	1460 non-null	int64
36	Heating	1460 non-null	object
37	${\tt HeatingQC}$	1460 non-null	object
38	CentralAir	1460 non-null	object
39	Electrical	1460 non-null	object
40	1stFlrSF	1460 non-null	int64

```
2ndFlrSF
                     1460 non-null
 41
                                      int64
 42
     LowQualFinSF
                     1460 non-null
                                      int64
 43
                     1460 non-null
                                      int64
     GrLivArea
     BsmtFullBath
                     1460 non-null
 44
                                      int64
 45
     BsmtHalfBath
                     1460 non-null
                                      int64
                     1460 non-null
     FullBath
                                      int64
 47
     HalfBath
                     1460 non-null
                                      int64
     BedroomAbvGr
 48
                     1460 non-null
                                      int64
     KitchenAbvGr
                     1460 non-null
 49
                                      int64
 50
     KitchenQual
                     1460 non-null
                                      object
 51
     {\tt TotRmsAbvGrd}
                     1460 non-null
                                      int64
 52
     Functional
                     1460 non-null
                                      object
 53
     Fireplaces
                     1460 non-null
                                      int64
 54
     FireplaceQu
                     1460 non-null
                                      object
 55
     GarageType
                     1460 non-null
                                      object
     GarageYrBlt
                     1460 non-null
                                      float64
 56
 57
     GarageFinish
                     1460 non-null
                                      object
 58
     GarageCars
                     1460 non-null
                                      int64
     GarageArea
                     1460 non-null
                                      int64
 59
 60
     GarageQual
                     1460 non-null
                                      object
 61
     GarageCond
                     1460 non-null
                                      object
     PavedDrive
 62
                     1460 non-null
                                      object
 63
     WoodDeckSF
                     1460 non-null
                                      int64
                     1460 non-null
                                      int64
 64
     OpenPorchSF
 65
     EnclosedPorch
                     1460 non-null
                                      int64
     3SsnPorch
                     1460 non-null
                                      int64
 66
     ScreenPorch
                     1460 non-null
 67
                                      int64
 68
     PoolArea
                     1460 non-null
                                      int64
                     1460 non-null
 69
     MiscVal
                                      int64
 70
     MoSold
                     1460 non-null
                                      int64
 71
     YrSold
                     1460 non-null
                                      int64
 72
     SaleType
                     1460 non-null
                                      object
 73
     SaleCondition
                     1460 non-null
                                      object
     SalePrice
                     1460 non-null
                                      int64
dtypes: float64(3), int64(34), object(38)
memory usage: 855.6+ KB
```

Da har vi blitt kvitt alle null-verdier trainingData. Nå må vi gjøre det samme for testData

[11]: testData.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1459 entries, 0 to 1458
Data columns (total 80 columns):

#	Column	Non-Null Count	Dtype
0	Id	1459 non-null	int64
1	MSSubClass	1459 non-null	int64
2	MSZoning	1455 non-null	object

_			
3	LotFrontage	1232 non-null	float64
4	LotArea	1459 non-null	int64
5	Street	1459 non-null	object
6	Alley	107 non-null	object
7	LotShape	1459 non-null	object
8	LandContour	1459 non-null	object
9	Utilities	1457 non-null	object
10	LotConfig	1459 non-null	object
11	LandSlope	1459 non-null	object
12	Neighborhood	1459 non-null	object
13	Condition1	1459 non-null	object
	Condition2	1459 non-null	object
15	BldgType	1459 non-null	object
16	HouseStyle	1459 non-null	object
17	OverallQual	1459 non-null	int64
18	OverallCond	1459 non-null	int64
19	YearBuilt	1459 non-null	int64
20	${\tt YearRemodAdd}$	1459 non-null	int64
21	RoofStyle	1459 non-null	object
22	RoofMatl	1459 non-null	object
23	Exterior1st	1458 non-null	object
24	Exterior2nd	1458 non-null	object
25	${ t MasVnrType}$	565 non-null	object
26	MasVnrArea	1444 non-null	float64
27	ExterQual	1459 non-null	object
28	ExterCond	1459 non-null	object
29	Foundation	1459 non-null	object
30	BsmtQual	1415 non-null	object
31	BsmtCond	1414 non-null	object
32	${\tt BsmtExposure}$	1415 non-null	object
33	${\tt BsmtFinType1}$	1417 non-null	object
34	BsmtFinSF1	1458 non-null	float64
35	${\tt BsmtFinType2}$	1417 non-null	object
36	BsmtFinSF2	1458 non-null	float64
37	${\tt BsmtUnfSF}$	1458 non-null	float64
38	TotalBsmtSF	1458 non-null	float64
39	Heating	1459 non-null	object
40	${\tt HeatingQC}$	1459 non-null	object
41	CentralAir	1459 non-null	object
42	Electrical	1459 non-null	object
43	1stFlrSF	1459 non-null	int64
44	2ndFlrSF	1459 non-null	int64
45	${\tt LowQualFinSF}$	1459 non-null	int64
46	GrLivArea	1459 non-null	int64
47	BsmtFullBath	1457 non-null	float64
48	BsmtHalfBath	1457 non-null	float64
49	FullBath	1459 non-null	int64
50	HalfBath	1459 non-null	int64

```
BedroomAbvGr
                        1459 non-null
                                        int64
      51
                        1459 non-null
                                        int64
      52 KitchenAbvGr
      53
         KitchenQual
                        1458 non-null
                                        object
      54 TotRmsAbvGrd
                        1459 non-null
                                        int64
      55 Functional
                        1457 non-null
                                        object
         Fireplaces
                        1459 non-null
                                        int64
         FireplaceQu
                        729 non-null
                                        object
      58
         GarageType
                        1383 non-null
                                        object
      59 GarageYrBlt
                        1381 non-null
                                        float64
         GarageFinish
      60
                        1381 non-null
                                        object
      61 GarageCars
                        1458 non-null
                                        float64
      62 GarageArea
                        1458 non-null
                                        float64
      63
         GarageQual
                        1381 non-null
                                        object
      64
         GarageCond
                        1381 non-null
                                        object
         PavedDrive
                        1459 non-null
                                        object
         WoodDeckSF
                        1459 non-null
                                        int64
      66
      67
         OpenPorchSF
                        1459 non-null
                                        int64
      68
         EnclosedPorch
                        1459 non-null
                                        int64
      69
         3SsnPorch
                        1459 non-null
                                        int64
      70
         ScreenPorch
                        1459 non-null
                                        int64
      71
         PoolArea
                        1459 non-null
                                        int64
      72 PoolQC
                        3 non-null
                                        object
      73 Fence
                        290 non-null
                                        object
      74 MiscFeature
                        51 non-null
                                        object
      75 MiscVal
                        1459 non-null
                                        int64
      76 MoSold
                        1459 non-null
                                        int64
      77
         YrSold
                        1459 non-null
                                        int64
      78
         SaleType
                        1458 non-null
                                        object
      79 SaleCondition 1459 non-null
                                        object
     dtypes: float64(11), int64(26), object(43)
     memory usage: 912.0+ KB
[12]: testData.drop(['Alley', 'PoolQC', 'Fence', L
       testData.drop(['Id'],axis=1,inplace=True)
[13]: for column in testData.columns:
         if testData[column].dtype in ['int64', 'float64']:
             column_mean = testData[column].mean()
             testData[column].fillna(column_mean, inplace=True)
[14]: for column in testData.columns:
         if testData[column].dtype == 'object':
             testData[column].fillna(testData[column].mode()[0], inplace=True)
[15]: testData.info()
     <class 'pandas.core.frame.DataFrame'>
```

RangeIndex: 1459 entries, 0 to 1458 Data columns (total 74 columns):

#	Column	Non-Null Count	Dtype
0	MSSubClass	1459 non-null	int64
1	MSZoning	1459 non-null	object
2	LotFrontage	1459 non-null	float64
3	LotArea	1459 non-null	int64
4	Street	1459 non-null	object
5	LotShape	1459 non-null	object
6	LandContour	1459 non-null	object
7	Utilities	1459 non-null	object
8	LotConfig	1459 non-null	object
9	LandSlope	1459 non-null	object
10	Neighborhood	1459 non-null	object
11	Condition1	1459 non-null	object
12	Condition2	1459 non-null	object
13	BldgType	1459 non-null	object
14	HouseStyle	1459 non-null	object
15	OverallQual	1459 non-null	int64
16	OverallCond	1459 non-null	int64
17	YearBuilt	1459 non-null	int64
18	YearRemodAdd	1459 non-null	int64
19	RoofStyle	1459 non-null	object
20	RoofMatl	1459 non-null	object
21	Exterior1st	1459 non-null	object
22	Exterior2nd	1459 non-null	object
23	MasVnrArea	1459 non-null	float64
24	ExterQual	1459 non-null	object
25	ExterCond	1459 non-null	object
26	Foundation	1459 non-null	object
27	BsmtQual	1459 non-null	object
28	BsmtCond	1459 non-null	object
29	BsmtExposure	1459 non-null	object
30	BsmtFinType1	1459 non-null	object
31	BsmtFinSF1	1459 non-null	float64
32	BsmtFinType2	1459 non-null	object
33	BsmtFinSF2	1459 non-null	float64
34	BsmtUnfSF	1459 non-null	float64
35	TotalBsmtSF	1459 non-null	float64
36	Heating	1459 non-null	object
37	HeatingQC	1459 non-null	object
38	CentralAir	1459 non-null	object
39	Electrical	1459 non-null	object
40	1stFlrSF	1459 non-null	int64
41	2ndFlrSF	1459 non-null	int64
42	LowQualFinSF	1459 non-null	int64
43	GrLivArea	1459 non-null	int64

```
44
     BsmtFullBath
                    1459 non-null
                                     float64
 45
     BsmtHalfBath
                    1459 non-null
                                     float64
 46
     FullBath
                    1459 non-null
                                     int64
 47
    HalfBath
                    1459 non-null
                                     int64
 48
     BedroomAbvGr
                    1459 non-null
                                     int64
    KitchenAbvGr
                    1459 non-null
                                     int64
 50
    KitchenQual
                    1459 non-null
                                     object
 51 TotRmsAbvGrd
                    1459 non-null
                                     int64
 52 Functional
                    1459 non-null
                                     object
 53
    Fireplaces
                    1459 non-null
                                     int64
 54
    FireplaceQu
                    1459 non-null
                                     object
     GarageType
 55
                    1459 non-null
                                     object
 56
     GarageYrBlt
                    1459 non-null
                                     float64
     GarageFinish
 57
                    1459 non-null
                                     object
 58
     GarageCars
                    1459 non-null
                                     float64
 59
     GarageArea
                    1459 non-null
                                     float64
 60
     GarageQual
                    1459 non-null
                                     object
     GarageCond
                    1459 non-null
 61
                                     object
 62
    PavedDrive
                    1459 non-null
                                     object
 63
     WoodDeckSF
                    1459 non-null
                                     int64
 64
     OpenPorchSF
                    1459 non-null
                                     int64
                    1459 non-null
     EnclosedPorch
                                     int64
     3SsnPorch
                    1459 non-null
                                     int64
     ScreenPorch
                    1459 non-null
                                     int64
 67
 68
    PoolArea
                    1459 non-null
                                     int64
 69
    MiscVal
                    1459 non-null
                                     int64
 70
    MoSold
                    1459 non-null
                                     int64
 71
    YrSold
                    1459 non-null
                                     int64
 72
     SaleType
                    1459 non-null
                                     object
     SaleCondition 1459 non-null
                                     object
dtypes: float64(11), int64(25), object(38)
memory usage: 843.6+ KB
```

Vi må håndtere alle kategoriske verdier. Dette er fordi maskinlærings modeller bruker ofte numeriske verdier. Lager så en metode som gjør dette med alle kolonnene som har kategoriske verdier.

```
def dummyGen(multcolumns):
    df_final=df
    i=0
    for fields in multcolumns:

    df1=pd.get_dummies(df[fields],drop_first=True)

    df.drop([fields],axis=1,inplace=True)
    if i==0:
        df_final=df1.copy()
```

```
else:
                   df_final=pd.concat([df_final,df1],axis=1)
               i=i+1
          df_final=pd.concat([df,df_final],axis=1)
          return df_final
[17]: kolonner = trainingData.select_dtypes(include=['object', 'category']).columns.
        →tolist()
[17]: 38
      testData.shape
[18]: (1459, 74)
     Vi vil unngå at test og data har ulike antall kategorier, det vil si at ikke alle kategorier av en type
     kommer opp. Det er viktig at test og training data har like verdier slik at modellen blir trent best
     mulig opp. Bruker da en concat å lage en ny tabell der trening og test data deler antall kolonner.
[19]: df=pd.concat([trainingData, testData],axis=0)
[20]: df=dummyGen(kolonner)
     MSZoning
     Street
     LotShape
     LandContour
     Utilities
     LotConfig
     LandSlope
     Neighborhood
     Condition1
     Condition2
     BldgType
     HouseStyle
     RoofStyle
     RoofMatl
     Exterior1st
     Exterior2nd
     ExterQual
     ExterCond
     Foundation
     BsmtQual
     BsmtCond
     BsmtExposure
```

```
BsmtFinType1
     BsmtFinType2
     Heating
     HeatingQC
     CentralAir
     Electrical
     KitchenQual
     Functional
     FireplaceQu
     GarageType
     GarageFinish
     GarageQual
     GarageCond
     PavedDrive
     SaleType
     SaleCondition
     Fjerner duplikater
[21]: df = df.loc[:,~df.columns.duplicated()]
      df.shape
[21]: (2919, 176)
[22]: df_Train=df.iloc[:1460,:]
      df_Test=df.iloc[1460:,:]
      df_Test.shape
[22]: (1459, 176)
[23]: df_Test.drop(['SalePrice'],axis=1,inplace=True)
```

/var/folders/xb/t_8qjfxd7wj4vhqfgglmrmy00000gn/T/ipykernel_23979/3985304647.py:1
: SettingWithCopyWarning:

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 df_Test.drop(['SalePrice'],axis=1,inplace=True)

Oppsummring: Da har vi ryddet i både training og test settet. fjernet NaN-verdier,laget dummies av kategoriske verdier. Vi har brukt concat for å få samme antall kategorier i test og training. Og så splittet det opp igjen.

Nå gjenstår det å velge algorithme og modell. Så kan gi den df_Train, trene den og så utføre modellen på df_Test data og få predikasjoner. Jeg velger å bruke XGboost.

Splitter opp df_train i X og y.

```
[24]: X_train = df_Train.drop(['SalePrice'],axis=1)
      y_train = df_Train['SalePrice']
[25]: !pip install xgboost
     Requirement already satisfied: xgboost in
     /Users/simonknutsson/anaconda3/lib/python3.11/site-packages (2.0.2)
     Requirement already satisfied: numpy in
     /Users/simonknutsson/anaconda3/lib/python3.11/site-packages (from xgboost)
     (1.24.3)
     Requirement already satisfied: scipy in
     /Users/simonknutsson/anaconda3/lib/python3.11/site-packages (from xgboost)
     (1.11.1)
     Som sagt har jeg tenkt til å bruke XGBboost algoritmen. For å optimalisere algoritmen vil gjøre jeg
     litt hyperparameter tuning. Det vil si å finne dem parameterne som retunerer det beste resultatet.
[26]: import xgboost
      regressor=xgboost.XGBRegressor()
 []:
[27]: from scipy.stats import randint, uniform
      # Definerer hvilke parametere vi vil endre og gjøre bedre.
      hyperparameter grid = {
          'n_estimators': randint(100, 1500),
          'max_depth': randint(2, 15),
          'learning_rate': uniform(0.05, 0.2),
          'subsample': uniform(0.5, 0.5),
          'min_child_weight': randint(1, 4),
          'gamma': uniform(0, 1)
[28]: random cv = RandomizedSearchCV(estimator=regressor,
                  param_distributions=hyperparameter_grid,
                  cv=5, n iter=50,
                  scoring = 'neg_mean_absolute_error',n_jobs = 4,
                  verbose = 5,
                  return_train_score = True,
                  random_state=42)
[29]: random_cv.fit(X_train,y_train)
     Fitting 5 folds for each of 50 candidates, totalling 250 fits
[29]: RandomizedSearchCV(cv=5,
                          estimator=XGBRegressor(base_score=None, booster=None,
                                                  callbacks=None,
```

```
colsample_bynode=None,
                                           colsample_bytree=None, device=None,
                                           early_stopping_rounds=None,
                                           enable_categorical=False,
                                           eval_metric=None, feature_types=None,
                                           gamma=None, grow_policy=None,
                                           importance_type=None,
                                           interaction constraints=None,
                                           learning rate=...
                                         'min child weight':
<scipy.stats._distn_infrastructure.rv_discrete_frozen object at 0x148ed8610>,
                                         'n estimators':
<scipy.stats._distn_infrastructure.rv_discrete_frozen object at 0x148ecab10>,
                                         'subsample':
<scipy.stats._distn_infrastructure.rv_continuous_frozen object at 0x148ed8110>},
                   random_state=42, return_train_score=True,
                   scoring='neg_mean_absolute_error', verbose=5)
```

colsample_bylevel=None,

random_cv blir trent med X_train og y_train. Så retunerer den en model med de beste estimatorene. Den lagrer vi variabelen best_model

```
[30]: best_model = random_cv.best_estimator_
```

Så trener vi den modelen med X train og y train

```
[31]: best_model.fit(X_train,y_train)
```

[31]: XGBRegressor(base_score=None, booster=None, callbacks=None, colsample_bylevel=None, colsample_bynode=None, colsample_bytree=None, device=None, early_stopping_rounds=None, enable_categorical=False, eval_metric=None, feature_types=None, gamma=0.009197051616629648, grow_policy=None, importance_type=None, interaction_constraints=None, learning_rate=0.07029430857320643, max_bin=None, max_cat_threshold=None, max_cat_to_onehot=None, max_delta_step=None, max_depth=4, max_leaves=None, min_child_weight=1, missing=nan, monotone_constraints=None, multi_strategy=None, n_estimators=1013, n_jobs=None, num_parallel_tree=None, random_state=None, ...)

Så kjører vi listen med modellen vi har trent og får resultatene

```
[32]: y_pred = best_model.predict(df_Test)
```

```
[33]: y_pred
```

[33]: array([120716.016, 158398.17 , 191163.56 , ..., 180101.25 , 115614.336, 233217.98], dtype=float32)

Så oppretter vi en csv fil med Id nr og SalePrice siden det er den vi har trent modellen til å retunere.

```
[34]: pred = pd.DataFrame(y_pred)
      sub_df=pd.read_csv('sample_submission.csv')
      datasets=pd.concat([sub_df['Id'], pred],axis=1)
      datasets.columns=['Id', 'SalePrice']
      datasets.to_csv('sample_submission.csv',index=False)
     [CV 1/5] END gamma=0.3745401188473625, learning_rate=0.24014286128198326,
     max_depth=12, min_child_weight=1, n_estimators=1144,
     subsample=0.5780093202212182;, score=(train=-8.251, test=-17206.340) total time=
     3.3s
     [CV 3/5] END gamma=0.15599452033620265, learning rate=0.061616722433639894,
     max_depth=9, min_child_weight=1, n_estimators=1223,
     subsample=0.5714334089609704;, score=(train=-0.171, test=-16572.982) total time=
     5.0s
     [CV 2/5] END gamma=0.6508884729488529, learning_rate=0.061282315805420054,
     max_depth=9, min_child_weight=2, n_estimators=905,
     subsample=0.5003893829205072;, score=(train=-81.072, test=-17704.979) total
     time=
             2.4s
     [CV 1/5] END gamma=0.9922115592912175, learning rate=0.17349630192554333,
     max_depth=11, min_child_weight=2, n_estimators=352,
     subsample=0.7159725093210578;, score=(train=-9.641, test=-16674.725) total time=
     1.3s
     [CV 2/5] END gamma=0.9922115592912175, learning_rate=0.17349630192554333,
     max_depth=11, min_child_weight=2, n_estimators=352,
     subsample=0.7159725093210578;, score=(train=-10.293, test=-18502.063) total
     time=
             1.3s
     [CV 1/5] END gamma=0.2912291401980419, learning rate=0.17237057894447588,
     max_depth=11, min_child_weight=3, n_estimators=289,
     subsample=0.5453032172664104;, score=(train=-107.979, test=-18157.299) total
     time=
             0.8s
     [CV 4/5] END gamma=0.2912291401980419, learning_rate=0.17237057894447588,
     max_depth=11, min_child_weight=3, n_estimators=289,
     subsample=0.5453032172664104;, score=(train=-80.226, test=-15492.485) total
     time= 0.8s
     [CV 3/5] END gamma=0.6183860093330873, learning_rate=0.12649239825343256,
     max depth=5, min child weight=1, n estimators=1254,
     subsample=0.9299702033681603;, score=(train=-4.480, test=-17228.134) total time=
     1.7s
     [CV 2/5] END gamma=0.6803075385877797, learning_rate=0.1400998503939086,
     max_depth=3, min_child_weight=1, n_estimators=415,
     subsample=0.7816441089227697;, score=(train=-4035.259, test=-17341.672) total
     time=
             0.3s
     [CV 5/5] END gamma=0.6803075385877797, learning rate=0.1400998503939086,
     max_depth=3, min_child_weight=1, n_estimators=415,
     subsample=0.7816441089227697;, score=(train=-4017.569, test=-16149.804) total
     time=
             0.3s
     [CV 3/5] END gamma=0.3854165025399161, learning rate=0.05319325044404284,
     max_depth=3, min_child_weight=3, n_estimators=1055,
```

```
subsample=0.5610191174223894;, score=(train=-4355.286, test=-15818.615) total
time=
       0.7s
[CV 1/5] END gamma=0.4951769101112702, learning rate=0.05687770422304368,
max_depth=2, min_child_weight=2, n_estimators=1125,
subsample=0.7125779372456224;, score=(train=-7541.582, test=-15165.893) total
time=
        0.6s
[CV 4/5] END gamma=0.4951769101112702, learning rate=0.05687770422304368,
max depth=2, min child weight=2, n estimators=1125,
subsample=0.7125779372456224;, score=(train=-7851.267, test=-13223.467) total
time=
       0.6s
[CV 3/5] END gamma=0.20794166286818883, learning rate=0.1635400655639983,
max_depth=14, min_child_weight=3, n_estimators=501,
subsample=0.8875664116805573;, score=(train=-0.309, test=-17868.612) total time=
1.8s
[CV 3/5] END gamma=0.9394989415641891, learning_rate=0.22896547008552975,
max_depth=9, min_child_weight=2, n_estimators=1375,
subsample=0.522613644455269;, score=(train=-0.282, test=-18248.251) total time=
1.9s
[CV 5/5] END gamma=0.9394989415641891, learning_rate=0.22896547008552975,
max depth=9, min child weight=2, n estimators=1375,
subsample=0.522613644455269;, score=(train=-0.270, test=-19045.187) total time=
1.7s
[CV 1/5] END gamma=0.965255307264138, learning rate=0.17140684953733692,
max_depth=14, min_child_weight=1, n_estimators=1280,
subsample=0.9010984903770198;, score=(train=-7.642, test=-18918.269) total time=
2.4s
[CV 5/5] END gamma=0.965255307264138, learning rate=0.17140684953733692,
max_depth=14, min_child_weight=1, n_estimators=1280,
subsample=0.9010984903770198;, score=(train=-0.366, test=-19910.246) total time=
1.9s
[CV 4/5] END gamma=0.07455064367977082, learning rate=0.24737738732010345,
max_depth=9, min_child_weight=1, n_estimators=747,
subsample=0.5027610585618012;, score=(train=-8.478, test=-16916.420) total time=
2.4s
[CV 1/5] END gamma=0.926300878513349, learning rate=0.18021540510038891,
max depth=13, min child weight=3, n estimators=300,
subsample=0.811649063413779;, score=(train=-17.424, test=-18528.755) total time=
1.0s
[CV 5/5] END gamma=0.926300878513349, learning_rate=0.18021540510038891,
max_depth=13, min_child_weight=3, n_estimators=300,
subsample=0.811649063413779;, score=(train=-9.626, test=-19023.190) total time=
1.1s
[CV 4/5] END gamma=0.3308980248526492, learning rate=0.06271167005720474,
max depth=8, min_child_weight=2, n_estimators=904, subsample=0.864803089169032;,
score=(train=-10.213, test=-15028.636) total time=
[CV 5/5] END gamma=0.6375574713552131, learning rate=0.2274425485152653,
max_depth=2, min_child_weight=3, n_estimators=712,
```

subsample=0.9858560476945519;, score=(train=-4496.686, test=-18655.218) total

```
time=
        0.4s
[CV 3/5] END gamma=0.8489138242660839, learning_rate=0.19434590423297465,
max_depth=8, min_child_weight=3, n_estimators=497,
subsample=0.7468977981821954;, score=(train=-1.002, test=-18537.234) total time=
1.3s
[CV 2/5] END gamma=0.5227328293819941, learning rate=0.13550820367170993,
max_depth=11, min_child_weight=1, n_estimators=963,
subsample=0.8182052056318903;, score=(train=-7.565, test=-17449.671) total time=
2.7s
[CV 1/5] END gamma=0.3143559810763267, learning_rate=0.15171413823294055,
max_depth=6, min_child_weight=3, n_estimators=330,
subsample=0.7051914615178149;, score=(train=-274.142, test=-15740.528) total
time=
       0.7s
[CV 3/5] END gamma=0.3143559810763267, learning rate=0.15171413823294055,
max_depth=6, min_child_weight=3, n_estimators=330,
subsample=0.7051914615178149;, score=(train=-259.507, test=-17858.372) total
time=
       0.6s
[CV 1/5] END gamma=0.7555511385430487, learning rate=0.0957596330983245,
max_depth=8, min_child_weight=3, n_estimators=1366,
subsample=0.5806106436270022;, score=(train=-8.702, test=-16033.333) total time=
3.0s
[CV 5/5] END gamma=0.7555511385430487, learning rate=0.0957596330983245,
max_depth=8, min_child_weight=3, n_estimators=1366,
subsample=0.5806106436270022;, score=(train=-0.721, test=-17656.152) total time=
3.3s
[CV 2/5] END gamma=0.2184404372168336, learning rate=0.13330198957407324,
max_depth=12, min_child_weight=2, n_estimators=1471,
subsample=0.6781489190384875;, score=(train=-7.988, test=-18155.612) total time=
3.7s
[CV 2/5] END gamma=0.906828441545754, learning rate=0.10442644987692706,
max_depth=12, min_child_weight=2, n_estimators=732,
subsample=0.9090073829612466;, score=(train=-7.523, test=-18486.229) total time=
2.6s
[CV 1/5] END gamma=0.8607305832563434, learning_rate=0.051390426106238146,
max depth=9, min child weight=3, n estimators=748,
subsample=0.7424149856794916;, score=(train=-104.696, test=-16393.839) total
time=
       2.1s
[CV 5/5] END gamma=0.8607305832563434, learning rate=0.051390426106238146,
max_depth=9, min_child_weight=3, n_estimators=748,
subsample=0.7424149856794916;, score=(train=-221.081, test=-17075.460) total
time=
       1.8s
[CV 4/5] END gamma=0.6924360328902703, learning rate=0.1038824667597043,
max_depth=9, min_child_weight=3, n_estimators=1497,
subsample=0.6093821097865351;, score=(train=-8.206, test=-14874.525) total time=
3.3s
[CV 2/5] END gamma=0.5581020020173412, learning rate=0.13076723421160819,
max_depth=2, min_child_weight=2, n_estimators=663,
```

subsample=0.6234380314193007;, score=(train=-6983.491, test=-17704.117) total

```
time=
       0.4s
[CV 4/5] END gamma=0.5581020020173412, learning_rate=0.13076723421160819,
max_depth=2, min_child_weight=2, n_estimators=663,
subsample=0.6234380314193007;, score=(train=-7076.117, test=-14197.712) total
time=
      0.4s
[CV 3/5] END gamma=0.6963042728397884, learning rate=0.19245411798488843,
max_depth=6, min_child_weight=1, n_estimators=1051,
subsample=0.5184434736772664;, score=(train=-0.271, test=-18878.553) total time=
2.0s
[CV 1/5] END gamma=0.6095643339798968, learning_rate=0.15053580464577232,
max_depth=7, min_child_weight=3, n_estimators=1348,
subsample=0.8171756723506819;, score=(train=-8.379, test=-16058.925) total time=
2.4s
[CV 5/5] END gamma=0.6095643339798968, learning rate=0.15053580464577232,
max_depth=7, min_child_weight=3, n_estimators=1348,
subsample=0.8171756723506819;, score=(train=-0.280, test=-17189.052) total time=
2.1s
[CV 5/5] END gamma=0.6807054515547668, learning rate=0.15618691666342727,
max_depth=11, min_child_weight=2, n_estimators=346,
subsample=0.8360677737029393;, score=(train=-0.668, test=-18589.343) total time=
1.5s
[CV 4/5] END gamma=0.7616196153287176, learning rate=0.09752750879847993,
max_depth=7, min_child_weight=1, n_estimators=983,
subsample=0.9917115704474215;, score=(train=-8.084, test=-14823.999) total time=
2.8s
[CV 3/5] END gamma=0.3988244424455306, learning rate=0.21328637464387679,
max_depth=2, min_child_weight=1, n_estimators=1247,
subsample=0.7540993883703593;, score=(train=-2723.165, test=-16871.864) total
time=
       0.8s
[CV 1/5] END gamma=0.6958128067908819, learning_rate=0.221671760962744,
max_depth=13, min_child_weight=3, n_estimators=502,
subsample=0.855574766219009;, score=(train=-7.767, test=-18240.581) total time=
1.6s
[CV 5/5] END gamma=0.6958128067908819, learning_rate=0.221671760962744,
max depth=13, min child weight=3, n estimators=502,
subsample=0.855574766219009;, score=(train=-0.300, test=-19053.487) total time=
1.6s
[CV 4/5] END gamma=0.8095010461397154, learning_rate=0.11973319745834587,
max_depth=8, min_child_weight=1, n_estimators=1028,
subsample=0.6987860105437611;, score=(train=-8.208, test=-14674.580) total time=
3.3s
[CV 3/5] END gamma=0.5177513505274801, learning rate=0.2175420211814656,
max_depth=12, min_child_weight=1, n_estimators=572,
subsample=0.6045358103688568;, score=(train=-0.222, test=-18464.606) total time=
1.8s
[CV 1/5] END gamma=0.5414479738275658, learning rate=0.18915687986901647,
max_depth=8, min_child_weight=2, n_estimators=723,
subsample=0.7776004057997312;, score=(train=-7.818, test=-17652.267) total time=
```

```
2.0s
[CV 4/5] END gamma=0.5414479738275658, learning_rate=0.18915687986901647,
max_depth=8, min_child_weight=2, n_estimators=723,
subsample=0.7776004057997312;, score=(train=-7.783, test=-15660.302) total time=
2.0s
[CV 4/5] END gamma=0.5296505783560065, learning rate=0.09837045818009034,
max_depth=13, min_child_weight=1, n_estimators=1160,
subsample=0.9413181715946699;, score=(train=-7.411, test=-16540.738) total time=
3.6s
[CV 3/5] END gamma=0.18870710834137938, learning_rate=0.10577427051843638,
max depth=4, min_child_weight=3, n_estimators=503,
subsample=0.9435432121325587;, score=(train=-1832.470, test=-17156.057) total
time=
       0.6s
[CV 1/5] END gamma=0.7798755458576239, learning rate=0.17840632923085759,
max_depth=12, min_child_weight=1, n_estimators=203,
subsample=0.803214529829795;, score=(train=-8.432, test=-17036.802) total time=
1.6s
[CV 5/5] END gamma=0.7798755458576239, learning rate=0.17840632923085759,
max_depth=12, min_child_weight=1, n_estimators=203,
subsample=0.803214529829795;, score=(train=-0.352, test=-18282.833) total time=
1.5s
[CV 4/5] END gamma=0.009197051616629648, learning rate=0.07029430857320643,
max_depth=4, min_child_weight=1, n_estimators=1013,
subsample=0.5804040257087493;, score=(train=-1139.123, test=-13647.389) total
time=
       1.3s
[CV 3/5] END gamma=0.5487337893665861, learning rate=0.18837903953853868,
max_depth=14, min_child_weight=1, n_estimators=260,
subsample=0.5090376818077604;, score=(train=-3.055, test=-18557.514) total time=
1.7s
[CV 2/5] END gamma=0.4938937151834346, learning rate=0.08576454184426577,
max_depth=9, min_child_weight=3, n_estimators=1318,
subsample=0.8288064461501716;, score=(train=-7.667, test=-17253.853) total time=
3.8s
[CV 2/5] END gamma=0.5683086033354716, learning_rate=0.0687349535656185,
max depth=4, min child weight=2, n estimators=1498,
subsample=0.6252309093027921;, score=(train=-484.948, test=-16473.002) total
time=
      1.8s
[CV 4/5] END gamma=0.5683086033354716, learning_rate=0.0687349535656185,
max_depth=4, min_child_weight=2, n_estimators=1498,
subsample=0.6252309093027921;, score=(train=-484.350, test=-13912.506) total
time=
       1.7s
[CV 3/5] END gamma=0.5898708475605439, learning rate=0.24577857165500183,
max_depth=4, min_child_weight=2, n_estimators=1406,
subsample=0.7171971827552144;, score=(train=-2.920, test=-18455.533) total time=
1.7s
[CV 3/5] END gamma=0.3500784076946757, learning rate=0.17902067240611297,
max_depth=2, min_child_weight=1, n_estimators=1119,
subsample=0.8612260576307527;, score=(train=-3416.769, test=-16316.748) total
```

```
time=
       0.7s
[CV 2/5] END gamma=0.2807723624408558, learning_rate=0.05486319328629077,
max_depth=2, min_child_weight=1, n_estimators=1295,
subsample=0.9702292921764571;, score=(train=-7508.306, test=-16868.996) total
time=
      0.8s
[CV 1/5] END gamma=0.9539285770025874, learning rate=0.23297287804408973,
max_depth=4, min_child_weight=1, n_estimators=1160,
subsample=0.9641592812938626;, score=(train=-22.714, test=-16133.214) total
       1.4s
[CV 5/5] END gamma=0.9539285770025874, learning_rate=0.23297287804408973,
max_depth=4, min_child_weight=1, n_estimators=1160,
subsample=0.9641592812938626;, score=(train=-7.199, test=-17761.710) total time=
1.4s
[CV 4/5] END gamma=0.42818414831731433, learning rate=0.24333096380873392,
max_depth=4, min_child_weight=1, n_estimators=1239,
subsample=0.6472244460347929;, score=(train=-36.771, test=-15414.345) total
time=
        1.5s
[CV 5/5] END gamma=0.38509772860192526, learning rate=0.22022733430337138,
max_depth=4, min_child_weight=2, n_estimators=665,
subsample=0.8480148983374864;, score=(train=-145.075, test=-17521.729) total
time=
       0.7s
[CV 3/5] END gamma=0.3745401188473625, learning rate=0.24014286128198326,
max_depth=12, min_child_weight=1, n_estimators=1144,
subsample=0.5780093202212182;, score=(train=-0.177, test=-19229.299) total time=
2.0s
[CV 5/5] END gamma=0.3745401188473625, learning rate=0.24014286128198326,
max_depth=12, min_child_weight=1, n_estimators=1144,
subsample=0.5780093202212182;, score=(train=-0.173, test=-18574.294) total time=
1.7s
[CV 4/5] END gamma=0.15599452033620265, learning rate=0.061616722433639894,
max_depth=9, min_child_weight=1, n_estimators=1223,
subsample=0.5714334089609704;, score=(train=-8.440, test=-13681.908) total time=
4.8s
[CV 3/5] END gamma=0.6508884729488529, learning_rate=0.061282315805420054,
max depth=9, min child weight=2, n estimators=905,
subsample=0.5003893829205072;, score=(train=-76.600, test=-16401.367) total
time=
       2.2s
[CV 5/5] END gamma=0.6508884729488529, learning_rate=0.061282315805420054,
max_depth=9, min_child_weight=2, n_estimators=905,
subsample=0.5003893829205072;, score=(train=-81.208, test=-16784.974) total
time=
        2.4s
[CV 5/5] END gamma=0.9922115592912175, learning rate=0.17349630192554333,
max_depth=11, min_child_weight=2, n_estimators=352,
subsample=0.7159725093210578;, score=(train=-1.254, test=-19300.795) total time=
1.2s
[CV 5/5] END gamma=0.2912291401980419, learning rate=0.17237057894447588,
max_depth=11, min_child_weight=3, n_estimators=289,
```

subsample=0.5453032172664104;, score=(train=-113.658, test=-18606.301) total

```
time=
       0.7s
[CV 4/5] END gamma=0.6183860093330873, learning_rate=0.12649239825343256,
max_depth=5, min_child_weight=1, n_estimators=1254,
subsample=0.9299702033681603;, score=(train=-16.805, test=-14868.344) total
time=
      1.7s
[CV 3/5] END gamma=0.6803075385877797, learning_rate=0.1400998503939086,
max depth=3, min child weight=1, n estimators=415,
subsample=0.7816441089227697;, score=(train=-3860.242, test=-17108.439) total
       0.3s
[CV 1/5] END gamma=0.3854165025399161, learning_rate=0.05319325044404284,
max_depth=3, min_child_weight=3, n_estimators=1055,
subsample=0.5610191174223894;, score=(train=-4648.438, test=-15167.716) total
time=
       0.7s
[CV 4/5] END gamma=0.3854165025399161, learning rate=0.05319325044404284,
max_depth=3, min_child_weight=3, n_estimators=1055,
subsample=0.5610191174223894;, score=(train=-4772.729, test=-13173.534) total
time=
       0.7s
[CV 3/5] END gamma=0.4951769101112702, learning rate=0.05687770422304368,
max_depth=2, min_child_weight=2, n_estimators=1125,
subsample=0.7125779372456224;, score=(train=-7329.754, test=-15741.460) total
time=
       0.6s
[CV 2/5] END gamma=0.20794166286818883, learning rate=0.1635400655639983,
max_depth=14, min_child_weight=3, n_estimators=501,
subsample=0.8875664116805573;, score=(train=-7.527, test=-17821.464) total time=
1.8s
[CV 1/5] END gamma=0.9394989415641891, learning rate=0.22896547008552975,
max_depth=9, min_child_weight=2, n_estimators=1375,
subsample=0.522613644455269;, score=(train=-10.022, test=-18665.805) total time=
2.7s
[CV 1/5] END gamma=0.32533033076326434, learning rate=0.12773545793789642,
max_depth=3, min_child_weight=3, n_estimators=1176,
subsample=0.7933755828319241;, score=(train=-1154.765, test=-15264.382) total
time=
       0.8s
[CV 3/5] END gamma=0.32533033076326434, learning_rate=0.12773545793789642,
max depth=3, min child weight=3, n estimators=1176,
subsample=0.7933755828319241;, score=(train=-1088.002, test=-17568.772) total
time=
      1.0s
[CV 2/5] END gamma=0.965255307264138, learning_rate=0.17140684953733692,
max_depth=14, min_child_weight=1, n_estimators=1280,
subsample=0.9010984903770198;, score=(train=-7.657, test=-18913.546) total time=
2.4s
[CV 2/5] END gamma=0.07455064367977082, learning rate=0.24737738732010345,
max_depth=9, min_child_weight=1, n_estimators=747,
subsample=0.5027610585618012;, score=(train=-9.352, test=-19186.417) total time=
2.4s
[CV 2/5] END gamma=0.8154614284548342, learning rate=0.19137146876952343,
max_depth=4, min_child_weight=3, n_estimators=388,
subsample=0.8029799873905057;, score=(train=-1043.022, test=-17136.960) total
```

```
time=
        0.6s
[CV 4/5] END gamma=0.8154614284548342, learning_rate=0.19137146876952343,
max_depth=4, min_child_weight=3, n_estimators=388,
subsample=0.8029799873905057;, score=(train=-1097.159, test=-14638.313) total
time=
      0.4s
[CV 2/5] END gamma=0.926300878513349, learning_rate=0.18021540510038891,
max_depth=13, min_child_weight=3, n_estimators=300,
subsample=0.811649063413779;, score=(train=-10.734, test=-18533.447) total time=
1.2s
[CV 1/5] END gamma=0.3308980248526492, learning_rate=0.06271167005720474,
max depth=8, min child weight=2, n estimators=904, subsample=0.864803089169032;,
score=(train=-11.913, test=-16565.598) total time=
                                                     2.9s
[CV 5/5] END gamma=0.3308980248526492, learning rate=0.06271167005720474,
max depth=8, min_child_weight=2, n_estimators=904, subsample=0.864803089169032;,
score=(train=-3.874, test=-17270.018) total time=
[CV 1/5] END gamma=0.5227328293819941, learning rate=0.13550820367170993,
max_depth=11, min_child_weight=1, n_estimators=963,
subsample=0.8182052056318903;, score=(train=-7.634, test=-18047.574) total time=
2.7s
[CV 5/5] END gamma=0.5227328293819941, learning rate=0.13550820367170993,
max_depth=11, min_child_weight=1, n_estimators=963,
subsample=0.8182052056318903;, score=(train=-0.264, test=-18341.384) total time=
2.1s
[CV 4/5] END gamma=0.7555511385430487, learning_rate=0.0957596330983245,
max_depth=8, min_child_weight=3, n_estimators=1366,
subsample=0.5806106436270022;, score=(train=-8.516, test=-14671.481) total time=
3.0s
[CV 3/5] END gamma=0.9296976523425731, learning rate=0.2116240759128834,
max_depth=10, min_child_weight=1, n_estimators=745,
subsample=0.7282672852414551;, score=(train=-0.316, test=-18294.252) total time=
1.5s
[CV 4/5] END gamma=0.9296976523425731, learning rate=0.2116240759128834,
max_depth=10, min_child_weight=1, n_estimators=745,
subsample=0.7282672852414551;, score=(train=-7.751, test=-15654.346) total time=
2.0s
[CV 4/5] END gamma=0.2184404372168336, learning rate=0.13330198957407324,
max_depth=12, min_child_weight=2, n_estimators=1471,
subsample=0.6781489190384875;, score=(train=-7.724, test=-14986.406) total time=
3.6s
[CV 3/5] END gamma=0.906828441545754, learning_rate=0.10442644987692706,
max_depth=12, min_child_weight=2, n_estimators=732,
subsample=0.9090073829612466;, score=(train=-0.406, test=-17659.139) total time=
2.5s
[CV 2/5] END gamma=0.8607305832563434, learning_rate=0.051390426106238146,
max_depth=9, min_child_weight=3, n_estimators=748,
subsample=0.7424149856794916;, score=(train=-139.481, test=-17712.274) total
time=
        2.0s
[CV 1/5] END gamma=0.6924360328902703, learning rate=0.1038824667597043,
```

```
max_depth=9, min_child_weight=3, n_estimators=1497,
subsample=0.6093821097865351;, score=(train=-8.404, test=-16542.808) total time=
3.5s
[CV 5/5] END gamma=0.6924360328902703, learning_rate=0.1038824667597043,
max depth=9, min child weight=3, n estimators=1497,
subsample=0.6093821097865351;, score=(train=-0.283, test=-17685.234) total time=
3.4s
[CV 4/5] END gamma=0.6963042728397884, learning_rate=0.19245411798488843,
max_depth=6, min_child_weight=1, n_estimators=1051,
subsample=0.5184434736772664;, score=(train=-10.208, test=-15383.016) total
time=
        2.2s
[CV 3/5] END gamma=0.6095643339798968, learning rate=0.15053580464577232,
max_depth=7, min_child_weight=3, n_estimators=1348,
subsample=0.8171756723506819;, score=(train=-0.288, test=-17470.886) total time=
2.2s
[CV 2/5] END gamma=0.6807054515547668, learning rate=0.15618691666342727,
max_depth=11, min_child_weight=2, n_estimators=346,
subsample=0.8360677737029393;, score=(train=-8.717, test=-18870.406) total time=
1.6s
[CV 2/5] END gamma=0.7616196153287176, learning rate=0.09752750879847993,
max_depth=7, min_child_weight=1, n_estimators=983,
subsample=0.9917115704474215;, score=(train=-8.578, test=-17529.957) total time=
3.0s
[CV 1/5] END gamma=0.3988244424455306, learning_rate=0.21328637464387679,
max_depth=2, min_child_weight=1, n_estimators=1247,
subsample=0.7540993883703593;, score=(train=-2802.375, test=-15717.210) total
time=
       0.8s
[CV 4/5] END gamma=0.3988244424455306, learning rate=0.21328637464387679,
max_depth=2, min_child_weight=1, n_estimators=1247,
subsample=0.7540993883703593;, score=(train=-2893.501, test=-14480.384) total
       0.8s
time=
[CV 2/5] END gamma=0.6958128067908819, learning rate=0.221671760962744,
max_depth=13, min_child_weight=3, n_estimators=502,
subsample=0.855574766219009;, score=(train=-7.647, test=-19481.987) total time=
1.7s
[CV 1/5] END gamma=0.8095010461397154, learning_rate=0.11973319745834587,
max depth=8, min child weight=1, n estimators=1028,
subsample=0.6987860105437611;, score=(train=-8.272, test=-16324.013) total time=
3.5s
[CV 5/5] END gamma=0.8095010461397154, learning_rate=0.11973319745834587,
max_depth=8, min_child_weight=1, n_estimators=1028,
subsample=0.6987860105437611;, score=(train=-0.314, test=-17847.357) total time=
2.9s
[CV 2/5] END gamma=0.5414479738275658, learning rate=0.18915687986901647,
max_depth=8, min_child_weight=2, n_estimators=723,
subsample=0.7776004057997312;, score=(train=-8.151, test=-18131.279) total time=
2.1s
```

[CV 1/5] END gamma=0.5296505783560065, learning rate=0.09837045818009034,

```
max_depth=13, min_child_weight=1, n_estimators=1160,
subsample=0.9413181715946699;, score=(train=-7.432, test=-18299.112) total time=
3.8s
[CV 5/5] END gamma=0.5296505783560065, learning_rate=0.09837045818009034,
max depth=13, min child weight=1, n estimators=1160,
subsample=0.9413181715946699;, score=(train=-0.319, test=-18612.162) total time=
3.0s
[CV 4/5] END gamma=0.7798755458576239, learning_rate=0.17840632923085759,
max_depth=12, min_child_weight=1, n_estimators=203,
subsample=0.803214529829795;, score=(train=-7.887, test=-15139.534) total time=
1.6s
[CV 3/5] END gamma=0.009197051616629648, learning rate=0.07029430857320643,
max_depth=4, min_child_weight=1, n_estimators=1013,
subsample=0.5804040257087493;, score=(train=-1102.876, test=-16036.844) total
time=
        1.3s
[CV 2/5] END gamma=0.5487337893665861, learning rate=0.18837903953853868,
max_depth=14, min_child_weight=1, n_estimators=260,
subsample=0.5090376818077604;, score=(train=-13.034, test=-18660.409) total
time=
       1.7s
[CV 1/5] END gamma=0.4938937151834346, learning rate=0.08576454184426577,
max_depth=9, min_child_weight=3, n_estimators=1318,
subsample=0.8288064461501716;, score=(train=-7.552, test=-16183.125) total time=
3.6s
[CV 5/5] END gamma=0.4938937151834346, learning rate=0.08576454184426577,
max_depth=9, min_child_weight=3, n_estimators=1318,
subsample=0.8288064461501716;, score=(train=-0.269, test=-17922.210) total time=
3.1s
[CV 1/5] END gamma=0.5898708475605439, learning rate=0.24577857165500183,
max_depth=4, min_child_weight=2, n_estimators=1406,
subsample=0.7171971827552144;, score=(train=-22.588, test=-16983.736) total
       1.7s
time=
[CV 5/5] END gamma=0.5898708475605439, learning_rate=0.24577857165500183,
max_depth=4, min_child_weight=2, n_estimators=1406,
subsample=0.7171971827552144;, score=(train=-2.573, test=-18389.611) total time=
1.7s
[CV 1/5] END gamma=0.2807723624408558, learning_rate=0.05486319328629077,
max_depth=2, min_child_weight=1, n_estimators=1295,
subsample=0.9702292921764571;, score=(train=-7486.787, test=-14673.024) total
time=
       0.8s
[CV 5/5] END gamma=0.2807723624408558, learning_rate=0.05486319328629077,
max_depth=2, min_child_weight=1, n_estimators=1295,
subsample=0.9702292921764571;, score=(train=-7176.337, test=-16399.461) total
time=
       0.8s
[CV 4/5] END gamma=0.9539285770025874, learning rate=0.23297287804408973,
max_depth=4, min_child_weight=1, n_estimators=1160,
subsample=0.9641592812938626;, score=(train=-22.711, test=-14276.043) total
time=
       1.4s
```

[CV 3/5] END gamma=0.42818414831731433, learning rate=0.24333096380873392,

```
max_depth=4, min_child_weight=1, n_estimators=1239,
subsample=0.6472244460347929;, score=(train=-5.511, test=-18686.280) total time=
1.5s
[CV 2/5] END gamma=0.38509772860192526, learning_rate=0.22022733430337138,
max depth=4, min child weight=2, n estimators=665,
subsample=0.8480148983374864;, score=(train=-201.585, test=-18035.469) total
time=
       0.8s
[CV 4/5] END gamma=0.38509772860192526, learning_rate=0.22022733430337138,
max_depth=4, min_child_weight=2, n_estimators=665,
subsample=0.8480148983374864;, score=(train=-183.737, test=-15151.213) total
time=
        0.7s
[CV 4/5] END gamma=0.3745401188473625, learning rate=0.24014286128198326,
max_depth=12, min_child_weight=1, n_estimators=1144,
subsample=0.5780093202212182;, score=(train=-8.526, test=-16508.045) total time=
3.0s
[CV 1/5] END gamma=0.15599452033620265, learning rate=0.061616722433639894,
max_depth=9, min_child_weight=1, n_estimators=1223,
subsample=0.5714334089609704;, score=(train=-8.041, test=-16033.350) total time=
4.7s
[CV 1/5] END gamma=0.6508884729488529, learning rate=0.061282315805420054,
max_depth=9, min_child_weight=2, n_estimators=905,
subsample=0.5003893829205072;, score=(train=-59.775, test=-15802.380) total
time=
       2.5s
[CV 4/5] END gamma=0.6508884729488529, learning_rate=0.061282315805420054,
max_depth=9, min_child_weight=2, n_estimators=905,
subsample=0.5003893829205072;, score=(train=-81.554, test=-13678.064) total
time=
        2.4s
[CV 4/5] END gamma=0.9922115592912175, learning rate=0.17349630192554333,
max_depth=11, min_child_weight=2, n_estimators=352,
subsample=0.7159725093210578;, score=(train=-8.473, test=-15488.546) total time=
1.3s
[CV 3/5] END gamma=0.2912291401980419, learning_rate=0.17237057894447588,
max_depth=11, min_child_weight=3, n_estimators=289,
subsample=0.5453032172664104;, score=(train=-165.462, test=-16936.569) total
time=
       0.7s
[CV 2/5] END gamma=0.6183860093330873, learning rate=0.12649239825343256,
max depth=5, min child weight=1, n estimators=1254,
subsample=0.9299702033681603;, score=(train=-17.466, test=-16652.067) total
time=
       1.8s
[CV 1/5] END gamma=0.6803075385877797, learning_rate=0.1400998503939086,
max_depth=3, min_child_weight=1, n_estimators=415,
subsample=0.7816441089227697;, score=(train=-4136.572, test=-15306.927) total
time=
       0.3s
[CV 4/5] END gamma=0.6803075385877797, learning rate=0.1400998503939086,
max_depth=3, min_child_weight=1, n_estimators=415,
subsample=0.7816441089227697;, score=(train=-4198.829, test=-13779.869) total
time=
       0.3s
[CV 2/5] END gamma=0.3854165025399161, learning rate=0.05319325044404284,
```

```
max_depth=3, min_child_weight=3, n_estimators=1055,
subsample=0.5610191174223894;, score=(train=-4665.175, test=-16869.708) total
time=
       0.7s
[CV 5/5] END gamma=0.3854165025399161, learning_rate=0.05319325044404284,
max depth=3, min child weight=3, n estimators=1055,
subsample=0.5610191174223894;, score=(train=-4291.571, test=-17106.684) total
time=
       0.7s
[CV 1/5] END gamma=0.20794166286818883, learning_rate=0.1635400655639983,
max_depth=14, min_child_weight=3, n_estimators=501,
subsample=0.8875664116805573;, score=(train=-7.584, test=-17945.894) total time=
1.7s
[CV 5/5] END gamma=0.20794166286818883, learning rate=0.1635400655639983,
max_depth=14, min_child_weight=3, n_estimators=501,
subsample=0.8875664116805573;, score=(train=-0.236, test=-18819.147) total time=
1.6s
[CV 4/5] END gamma=0.9394989415641891, learning rate=0.22896547008552975,
max_depth=9, min_child_weight=2, n_estimators=1375,
subsample=0.522613644455269;, score=(train=-10.595, test=-16111.482) total time=
2.5s
[CV 5/5] END gamma=0.32533033076326434, learning rate=0.12773545793789642,
max_depth=3, min_child_weight=3, n_estimators=1176,
subsample=0.7933755828319241;, score=(train=-1158.502, test=-17572.620) total
       1.0s
[CV 3/5] END gamma=0.965255307264138, learning_rate=0.17140684953733692,
max_depth=14, min_child_weight=1, n_estimators=1280,
subsample=0.9010984903770198;, score=(train=-0.357, test=-18516.841) total time=
1.9s
[CV 1/5] END gamma=0.07455064367977082, learning rate=0.24737738732010345,
max_depth=9, min_child_weight=1, n_estimators=747,
subsample=0.5027610585618012;, score=(train=-10.137, test=-17815.542) total
time=
        2.3s
[CV 5/5] END gamma=0.07455064367977082, learning rate=0.24737738732010345,
max_depth=9, min_child_weight=1, n_estimators=747,
subsample=0.5027610585618012;, score=(train=-0.082, test=-18542.690) total time=
1.8s
[CV 4/5] END gamma=0.926300878513349, learning_rate=0.18021540510038891,
max depth=13, min child weight=3, n estimators=300,
subsample=0.811649063413779;, score=(train=-10.992, test=-15605.162) total time=
1.1s
[CV 3/5] END gamma=0.3308980248526492, learning_rate=0.06271167005720474,
max_depth=8, min_child_weight=2, n_estimators=904, subsample=0.864803089169032;,
score=(train=-3.475, test=-16521.377) total time=
[CV 1/5] END gamma=0.6375574713552131, learning rate=0.2274425485152653,
max_depth=2, min_child_weight=3, n_estimators=712,
subsample=0.9858560476945519;, score=(train=-4702.805, test=-15605.749) total
[CV 3/5] END gamma=0.6375574713552131, learning_rate=0.2274425485152653,
max_depth=2, min_child_weight=3, n_estimators=712,
```

```
subsample=0.9858560476945519;, score=(train=-4540.175, test=-17484.260) total
time=
       0.5s
[CV 2/5] END gamma=0.8489138242660839, learning rate=0.19434590423297465,
max_depth=8, min_child_weight=3, n_estimators=497,
subsample=0.7468977981821954;, score=(train=-10.218, test=-17458.190) total
time=
        1.2s
[CV 4/5] END gamma=0.8489138242660839, learning rate=0.19434590423297465,
max depth=8, min child weight=3, n estimators=497,
subsample=0.7468977981821954;, score=(train=-8.533, test=-15075.076) total time=
1.3s
[CV 3/5] END gamma=0.5227328293819941, learning rate=0.13550820367170993,
max_depth=11, min_child_weight=1, n_estimators=963,
subsample=0.8182052056318903;, score=(train=-0.261, test=-18021.616) total time=
2.2s
[CV 2/5] END gamma=0.3143559810763267, learning_rate=0.15171413823294055,
max_depth=6, min_child_weight=3, n_estimators=330,
subsample=0.7051914615178149;, score=(train=-274.921, test=-17976.325) total
time=
       0.6s
[CV 5/5] END gamma=0.3143559810763267, learning_rate=0.15171413823294055,
max depth=6, min child weight=3, n estimators=330,
subsample=0.7051914615178149;, score=(train=-323.947, test=-17857.999) total
time=
        0.6s
[CV 3/5] END gamma=0.7555511385430487, learning_rate=0.0957596330983245,
max_depth=8, min_child_weight=3, n_estimators=1366,
subsample=0.5806106436270022;, score=(train=-1.401, test=-16838.658) total time=
2.9s
[CV 2/5] END gamma=0.9296976523425731, learning rate=0.2116240759128834,
max_depth=10, min_child_weight=1, n_estimators=745,
subsample=0.7282672852414551;, score=(train=-7.611, test=-18807.161) total time=
1.9s
[CV 5/5] END gamma=0.9296976523425731, learning rate=0.2116240759128834,
max_depth=10, min_child_weight=1, n_estimators=745,
subsample=0.7282672852414551;, score=(train=-0.316, test=-17731.031) total time=
1.5s
[CV 3/5] END gamma=0.2184404372168336, learning rate=0.13330198957407324,
max_depth=12, min_child_weight=2, n_estimators=1471,
subsample=0.6781489190384875;, score=(train=-0.150, test=-17617.515) total time=
2.9s
[CV 1/5] END gamma=0.906828441545754, learning_rate=0.10442644987692706,
max_depth=12, min_child_weight=2, n_estimators=732,
subsample=0.9090073829612466;, score=(train=-7.564, test=-17437.885) total time=
2.4s
[CV 5/5] END gamma=0.906828441545754, learning rate=0.10442644987692706,
max_depth=12, min_child_weight=2, n_estimators=732,
subsample=0.9090073829612466;, score=(train=-0.388, test=-18312.694) total time=
[CV 3/5] END gamma=0.8607305832563434, learning_rate=0.051390426106238146,
```

max_depth=9, min_child_weight=3, n_estimators=748,

```
subsample=0.7424149856794916;, score=(train=-186.010, test=-16741.309) total
time=
       1.9s
[CV 2/5] END gamma=0.6924360328902703, learning rate=0.1038824667597043,
max_depth=9, min_child_weight=3, n_estimators=1497,
subsample=0.6093821097865351;, score=(train=-8.297, test=-16606.396) total time=
3.6s
[CV 3/5] END gamma=0.5581020020173412, learning rate=0.13076723421160819,
max depth=2, min child weight=2, n estimators=663,
subsample=0.6234380314193007;, score=(train=-6627.724, test=-15869.402) total
time=
       0.4s
[CV 1/5] END gamma=0.6963042728397884, learning rate=0.19245411798488843,
max_depth=6, min_child_weight=1, n_estimators=1051,
subsample=0.5184434736772664;, score=(train=-11.363, test=-17752.577) total
time=
        2.1s
[CV 5/5] END gamma=0.6963042728397884, learning_rate=0.19245411798488843,
max_depth=6, min_child_weight=1, n_estimators=1051,
subsample=0.5184434736772664;, score=(train=-0.266, test=-17501.773) total time=
2.0s
[CV 4/5] END gamma=0.6095643339798968, learning_rate=0.15053580464577232,
max depth=7, min child weight=3, n estimators=1348,
subsample=0.8171756723506819;, score=(train=-8.095, test=-14119.953) total time=
2.5s
[CV 4/5] END gamma=0.6807054515547668, learning rate=0.15618691666342727,
max_depth=11, min_child_weight=2, n_estimators=346,
subsample=0.8360677737029393;, score=(train=-8.345, test=-16324.876) total time=
1.6s
[CV 3/5] END gamma=0.7616196153287176, learning rate=0.09752750879847993,
max_depth=7, min_child_weight=1, n_estimators=983,
subsample=0.9917115704474215;, score=(train=-0.931, test=-17465.821) total time=
2.9s
[CV 2/5] END gamma=0.3988244424455306, learning rate=0.21328637464387679,
max_depth=2, min_child_weight=1, n_estimators=1247,
subsample=0.7540993883703593;, score=(train=-2816.162, test=-17115.561) total
time=
       0.9s
[CV 5/5] END gamma=0.3988244424455306, learning rate=0.21328637464387679,
max depth=2, min child weight=1, n estimators=1247,
subsample=0.7540993883703593;, score=(train=-2650.287, test=-17481.876) total
time=
       0.8s
[CV 3/5] END gamma=0.6958128067908819, learning_rate=0.221671760962744,
max_depth=13, min_child_weight=3, n_estimators=502,
subsample=0.855574766219009;, score=(train=-0.305, test=-18501.488) total time=
1.7s
[CV 2/5] END gamma=0.8095010461397154, learning rate=0.11973319745834587,
max depth=8, min_child_weight=1, n_estimators=1028,
subsample=0.6987860105437611;, score=(train=-7.930, test=-17313.947) total time=
[CV 2/5] END gamma=0.5177513505274801, learning_rate=0.2175420211814656,
```

max_depth=12, min_child_weight=1, n_estimators=572,

```
subsample=0.6045358103688568;, score=(train=-9.582, test=-18179.784) total time=
2.5s
[CV 5/5] END gamma=0.5177513505274801, learning rate=0.2175420211814656,
max_depth=12, min_child_weight=1, n_estimators=572,
subsample=0.6045358103688568;, score=(train=-0.221, test=-17770.187) total time=
1.8s
[CV 3/5] END gamma=0.5414479738275658, learning rate=0.18915687986901647,
max depth=8, min child weight=2, n estimators=723,
subsample=0.7776004057997312;, score=(train=-0.262, test=-17828.434) total time=
1.9s
[CV 2/5] END gamma=0.5296505783560065, learning rate=0.09837045818009034,
max_depth=13, min_child_weight=1, n_estimators=1160,
subsample=0.9413181715946699;, score=(train=-7.425, test=-18494.403) total time=
3.7s
[CV 2/5] END gamma=0.18870710834137938, learning_rate=0.10577427051843638,
max_depth=4, min_child_weight=3, n_estimators=503,
subsample=0.9435432121325587;, score=(train=-1888.619, test=-16854.215) total
time=
       0.6s
[CV 5/5] END gamma=0.18870710834137938, learning_rate=0.10577427051843638,
max depth=4, min child weight=3, n estimators=503,
subsample=0.9435432121325587;, score=(train=-1933.859, test=-16054.185) total
        0.6s
time=
[CV 3/5] END gamma=0.7798755458576239, learning rate=0.17840632923085759,
max_depth=12, min_child_weight=1, n_estimators=203,
subsample=0.803214529829795;, score=(train=-0.372, test=-18634.650) total time=
1.7s
[CV 2/5] END gamma=0.009197051616629648, learning rate=0.07029430857320643,
max_depth=4, min_child_weight=1, n_estimators=1013,
subsample=0.5804040257087493;, score=(train=-1170.283, test=-16087.775) total
time=
       1.3s
[CV 1/5] END gamma=0.5487337893665861, learning rate=0.18837903953853868,
max_depth=14, min_child_weight=1, n_estimators=260,
subsample=0.5090376818077604;, score=(train=-10.730, test=-17853.950) total
time=
        1.7s
[CV 5/5] END gamma=0.5487337893665861, learning rate=0.18837903953853868,
max depth=14, min child weight=1, n estimators=260,
subsample=0.5090376818077604;, score=(train=-2.808, test=-17840.679) total time=
1.8s
[CV 4/5] END gamma=0.4938937151834346, learning_rate=0.08576454184426577,
max_depth=9, min_child_weight=3, n_estimators=1318,
subsample=0.8288064461501716;, score=(train=-7.678, test=-14832.204) total time=
3.6s
[CV 3/5] END gamma=0.5683086033354716, learning rate=0.0687349535656185,
max_depth=4, min_child_weight=2, n_estimators=1498,
subsample=0.6252309093027921;, score=(train=-452.756, test=-16697.572) total
[CV 2/5] END gamma=0.5898708475605439, learning_rate=0.24577857165500183,
```

max_depth=4, min_child_weight=2, n_estimators=1406,

```
subsample=0.7171971827552144;, score=(train=-29.759, test=-16567.997) total
time=
       1.7s
[CV 1/5] END gamma=0.3500784076946757, learning rate=0.17902067240611297,
max_depth=2, min_child_weight=1, n_estimators=1119,
subsample=0.8612260576307527;, score=(train=-3583.756, test=-15208.171) total
time=
        0.7s
[CV 2/5] END gamma=0.3500784076946757, learning rate=0.17902067240611297,
max depth=2, min child weight=1, n estimators=1119,
subsample=0.8612260576307527;, score=(train=-3642.883, test=-17315.397) total
time=
       0.8s
[CV 5/5] END gamma=0.3500784076946757, learning rate=0.17902067240611297,
max_depth=2, min_child_weight=1, n_estimators=1119,
subsample=0.8612260576307527;, score=(train=-3505.888, test=-16095.607) total
time=
       0.8s
[CV 4/5] END gamma=0.2807723624408558, learning_rate=0.05486319328629077,
max_depth=2, min_child_weight=1, n_estimators=1295,
subsample=0.9702292921764571;, score=(train=-7694.487, test=-13394.297) total
time=
       0.8s
[CV 3/5] END gamma=0.9539285770025874, learning_rate=0.23297287804408973,
max depth=4, min child weight=1, n estimators=1160,
subsample=0.9641592812938626;, score=(train=-9.409, test=-17532.007) total time=
1.4s
[CV 2/5] END gamma=0.42818414831731433, learning_rate=0.24333096380873392,
max_depth=4, min_child_weight=1, n_estimators=1239,
subsample=0.6472244460347929;, score=(train=-39.222, test=-18300.671) total
time=
       1.5s
[CV 1/5] END gamma=0.38509772860192526, learning rate=0.22022733430337138,
max_depth=4, min_child_weight=2, n_estimators=665,
subsample=0.8480148983374864;, score=(train=-181.939, test=-16986.356) total
time=
       0.8s
[CV 3/5] END gamma=0.38509772860192526, learning rate=0.22022733430337138,
max_depth=4, min_child_weight=2, n_estimators=665,
subsample=0.8480148983374864;, score=(train=-129.068, test=-17333.750) total
time=
       0.8s
[CV 2/5] END gamma=0.3745401188473625, learning rate=0.24014286128198326,
max_depth=12, min_child_weight=1, n_estimators=1144,
subsample=0.5780093202212182;, score=(train=-8.496, test=-18184.567) total time=
[CV 2/5] END gamma=0.15599452033620265, learning_rate=0.061616722433639894,
max_depth=9, min_child_weight=1, n_estimators=1223,
subsample=0.5714334089609704;, score=(train=-8.764, test=-17313.866) total time=
4.6s
[CV 5/5] END gamma=0.15599452033620265, learning_rate=0.061616722433639894,
max_depth=9, min_child_weight=1, n_estimators=1223,
subsample=0.5714334089609704;, score=(train=-0.161, test=-16460.813) total time=
[CV 3/5] END gamma=0.9922115592912175, learning_rate=0.17349630192554333,
```

max_depth=11, min_child_weight=2, n_estimators=352,

```
subsample=0.7159725093210578;, score=(train=-1.640, test=-18242.181) total time=
1.3s
[CV 2/5] END gamma=0.2912291401980419, learning rate=0.17237057894447588,
max_depth=11, min_child_weight=3, n_estimators=289,
subsample=0.5453032172664104;, score=(train=-113.174, test=-19292.055) total
time=
        0.8s
[CV 1/5] END gamma=0.6183860093330873, learning rate=0.12649239825343256,
max depth=5, min child weight=1, n estimators=1254,
subsample=0.9299702033681603;, score=(train=-18.429, test=-15631.360) total
time=
       1.8s
[CV 5/5] END gamma=0.6183860093330873, learning rate=0.12649239825343256,
max_depth=5, min_child_weight=1, n_estimators=1254,
subsample=0.9299702033681603;, score=(train=-5.693, test=-17107.346) total time=
1.8s
[CV 2/5] END gamma=0.4951769101112702, learning_rate=0.05687770422304368,
max_depth=2, min_child_weight=2, n_estimators=1125,
subsample=0.7125779372456224;, score=(train=-7710.327, test=-16930.351) total
time=
       0.6s
[CV 5/5] END gamma=0.4951769101112702, learning_rate=0.05687770422304368,
max depth=2, min child weight=2, n estimators=1125,
subsample=0.7125779372456224;, score=(train=-7306.251, test=-16597.868) total
time=
       0.6s
[CV 4/5] END gamma=0.20794166286818883, learning rate=0.1635400655639983,
max_depth=14, min_child_weight=3, n_estimators=501,
subsample=0.8875664116805573;, score=(train=-7.518, test=-16487.267) total time=
1.6s
[CV 2/5] END gamma=0.9394989415641891, learning rate=0.22896547008552975,
max_depth=9, min_child_weight=2, n_estimators=1375,
subsample=0.522613644455269;, score=(train=-9.360, test=-18599.356) total time=
2.7s
[CV 2/5] END gamma=0.32533033076326434, learning rate=0.12773545793789642,
max_depth=3, min_child_weight=3, n_estimators=1176,
subsample=0.7933755828319241;, score=(train=-1129.420, test=-17608.984) total
time=
       0.8s
[CV 4/5] END gamma=0.32533033076326434, learning rate=0.12773545793789642,
max_depth=3, min_child_weight=3, n_estimators=1176,
subsample=0.7933755828319241;, score=(train=-1171.407, test=-13675.421) total
time=
       1.1s
[CV 4/5] END gamma=0.965255307264138, learning_rate=0.17140684953733692,
max_depth=14, min_child_weight=1, n_estimators=1280,
subsample=0.9010984903770198;, score=(train=-7.472, test=-16749.113) total time=
2.4s
[CV 3/5] END gamma=0.07455064367977082, learning rate=0.24737738732010345,
max depth=9, min_child_weight=1, n_estimators=747,
subsample=0.5027610585618012;, score=(train=-0.080, test=-19118.136) total time=
[CV 1/5] END gamma=0.8154614284548342, learning_rate=0.19137146876952343,
```

max_depth=4, min_child_weight=3, n_estimators=388,

```
subsample=0.8029799873905057;, score=(train=-1043.652, test=-16036.346) total
time=
       0.6s
[CV 3/5] END gamma=0.8154614284548342, learning rate=0.19137146876952343,
max_depth=4, min_child_weight=3, n_estimators=388,
subsample=0.8029799873905057;, score=(train=-1053.029, test=-17156.931) total
time=
        0.5s
[CV 5/5] END gamma=0.8154614284548342, learning rate=0.19137146876952343,
max depth=4, min child weight=3, n estimators=388,
subsample=0.8029799873905057;, score=(train=-1131.524, test=-17840.111) total
time=
       0.5s
[CV 3/5] END gamma=0.926300878513349, learning rate=0.18021540510038891,
max_depth=13, min_child_weight=3, n_estimators=300,
subsample=0.811649063413779;, score=(train=-14.708, test=-18026.014) total time=
1.1s
[CV 2/5] END gamma=0.3308980248526492, learning rate=0.06271167005720474,
max_depth=8, min_child_weight=2, n_estimators=904, subsample=0.864803089169032;,
score=(train=-12.890, test=-17092.070) total time=
[CV 2/5] END gamma=0.6375574713552131, learning rate=0.2274425485152653,
max_depth=2, min_child_weight=3, n_estimators=712,
subsample=0.9858560476945519;, score=(train=-4708.874, test=-16864.719) total
       0.4s
[CV 4/5] END gamma=0.6375574713552131, learning rate=0.2274425485152653,
max_depth=2, min_child_weight=3, n_estimators=712,
subsample=0.9858560476945519;, score=(train=-4901.395, test=-14005.792) total
time=
       0.5s
[CV 1/5] END gamma=0.8489138242660839, learning rate=0.19434590423297465,
max_depth=8, min_child_weight=3, n_estimators=497,
subsample=0.7468977981821954;, score=(train=-8.136, test=-16591.328) total time=
1.3s
[CV 5/5] END gamma=0.8489138242660839, learning rate=0.19434590423297465,
max_depth=8, min_child_weight=3, n_estimators=497,
subsample=0.7468977981821954;, score=(train=-0.733, test=-17904.247) total time=
1.4s
[CV 4/5] END gamma=0.5227328293819941, learning_rate=0.13550820367170993,
max depth=11, min child weight=1, n estimators=963,
subsample=0.8182052056318903;, score=(train=-8.115, test=-15214.355) total time=
2.6s
[CV 4/5] END gamma=0.3143559810763267, learning_rate=0.15171413823294055,
max_depth=6, min_child_weight=3, n_estimators=330,
subsample=0.7051914615178149;, score=(train=-276.073, test=-14349.072) total
time=
       0.6s
[CV 2/5] END gamma=0.7555511385430487, learning_rate=0.0957596330983245,
max_depth=8, min_child_weight=3, n_estimators=1366,
subsample=0.5806106436270022;, score=(train=-10.059, test=-17220.166) total
time=
        2.9s
[CV 1/5] END gamma=0.9296976523425731, learning rate=0.2116240759128834,
max_depth=10, min_child_weight=1, n_estimators=745,
subsample=0.7282672852414551;, score=(train=-7.772, test=-17332.680) total time=
```

```
2.0s
[CV 1/5] END gamma=0.2184404372168336, learning_rate=0.13330198957407324,
max_depth=12, min_child_weight=2, n_estimators=1471,
subsample=0.6781489190384875;, score=(train=-8.168, test=-17099.834) total time=
3.7s
[CV 5/5] END gamma=0.2184404372168336, learning rate=0.13330198957407324,
max_depth=12, min_child_weight=2, n_estimators=1471,
subsample=0.6781489190384875;, score=(train=-0.150, test=-18152.537) total time=
3.0s
[CV 4/5] END gamma=0.906828441545754, learning_rate=0.10442644987692706,
max_depth=12, min_child_weight=2, n_estimators=732,
subsample=0.9090073829612466;, score=(train=-7.602, test=-15784.241) total time=
2.4s
[CV 4/5] END gamma=0.8607305832563434, learning_rate=0.051390426106238146,
max_depth=9, min_child_weight=3, n_estimators=748,
subsample=0.7424149856794916;, score=(train=-88.544, test=-14775.547) total
time=
        2.1s
[CV 3/5] END gamma=0.6924360328902703, learning rate=0.1038824667597043,
max_depth=9, min_child_weight=3, n_estimators=1497,
subsample=0.6093821097865351;, score=(train=-0.312, test=-17118.865) total time=
3.3s
[CV 1/5] END gamma=0.5581020020173412, learning rate=0.13076723421160819,
max_depth=2, min_child_weight=2, n_estimators=663,
subsample=0.6234380314193007;, score=(train=-6894.284, test=-15704.855) total
time=
       0.4s
[CV 5/5] END gamma=0.5581020020173412, learning rate=0.13076723421160819,
max_depth=2, min_child_weight=2, n_estimators=663,
subsample=0.6234380314193007;, score=(train=-6704.493, test=-17120.593) total
time=
       0.4s
[CV 2/5] END gamma=0.6963042728397884, learning rate=0.19245411798488843,
max_depth=6, min_child_weight=1, n_estimators=1051,
subsample=0.5184434736772664;, score=(train=-10.116, test=-18205.597) total
time=
       2.1s
[CV 2/5] END gamma=0.6095643339798968, learning_rate=0.15053580464577232,
max depth=7, min child weight=3, n estimators=1348,
subsample=0.8171756723506819;, score=(train=-8.336, test=-17091.870) total time=
2.5s
[CV 1/5] END gamma=0.6807054515547668, learning_rate=0.15618691666342727,
max_depth=11, min_child_weight=2, n_estimators=346,
subsample=0.8360677737029393;, score=(train=-8.028, test=-17680.011) total time=
1.5s
[CV 3/5] END gamma=0.6807054515547668, learning rate=0.15618691666342727,
max_depth=11, min_child_weight=2, n_estimators=346,
subsample=0.8360677737029393;, score=(train=-3.888, test=-18029.026) total time=
1.3s
[CV 1/5] END gamma=0.7616196153287176, learning rate=0.09752750879847993,
max_depth=7, min_child_weight=1, n_estimators=983,
```

subsample=0.9917115704474215;, score=(train=-8.945, test=-16678.254) total time=

```
2.8s
[CV 5/5] END gamma=0.7616196153287176, learning_rate=0.09752750879847993,
max_depth=7, min_child_weight=1, n_estimators=983,
subsample=0.9917115704474215;, score=(train=-0.899, test=-17685.678) total time=
2.8s
[CV 4/5] END gamma=0.6958128067908819, learning rate=0.221671760962744,
max_depth=13, min_child_weight=3, n_estimators=502,
subsample=0.855574766219009;, score=(train=-7.762, test=-16212.478) total time=
1.6s
[CV 3/5] END gamma=0.8095010461397154, learning_rate=0.11973319745834587,
max_depth=8, min_child_weight=1, n_estimators=1028,
subsample=0.6987860105437611;, score=(train=-0.309, test=-16625.633) total time=
2.8s
[CV 1/5] END gamma=0.5177513505274801, learning rate=0.2175420211814656,
max_depth=12, min_child_weight=1, n_estimators=572,
subsample=0.6045358103688568;, score=(train=-10.052, test=-17482.793) total
time=
        2.4s
[CV 4/5] END gamma=0.5177513505274801, learning rate=0.2175420211814656,
max_depth=12, min_child_weight=1, n_estimators=572,
subsample=0.6045358103688568;, score=(train=-9.734, test=-15736.554) total time=
2.2s
[CV 5/5] END gamma=0.5414479738275658, learning rate=0.18915687986901647,
max_depth=8, min_child_weight=2, n_estimators=723,
subsample=0.7776004057997312;, score=(train=-0.262, test=-19016.782) total time=
1.9s
[CV 3/5] END gamma=0.5296505783560065, learning_rate=0.09837045818009034,
max_depth=13, min_child_weight=1, n_estimators=1160,
subsample=0.9413181715946699;, score=(train=-0.309, test=-18160.633) total time=
3.3s
[CV 1/5] END gamma=0.18870710834137938, learning rate=0.10577427051843638,
max_depth=4, min_child_weight=3, n_estimators=503,
subsample=0.9435432121325587;, score=(train=-2012.558, test=-15298.186) total
time=
       0.6s
[CV 4/5] END gamma=0.18870710834137938, learning_rate=0.10577427051843638,
max depth=4, min child weight=3, n estimators=503,
subsample=0.9435432121325587;, score=(train=-1900.865, test=-14080.135) total
time= 0.6s
[CV 2/5] END gamma=0.7798755458576239, learning_rate=0.17840632923085759,
max_depth=12, min_child_weight=1, n_estimators=203,
subsample=0.803214529829795;, score=(train=-7.836, test=-18686.062) total time=
1.7s
[CV 1/5] END gamma=0.009197051616629648, learning_rate=0.07029430857320643,
max_depth=4, min_child_weight=1, n_estimators=1013,
subsample=0.5804040257087493;, score=(train=-1090.900, test=-14663.066) total
time=
        1.3s
[CV 5/5] END gamma=0.009197051616629648, learning rate=0.07029430857320643,
max_depth=4, min_child_weight=1, n_estimators=1013,
subsample=0.5804040257087493;, score=(train=-1073.945, test=-16565.489) total
```

```
time=
             1.3s
     [CV 4/5] END gamma=0.5487337893665861, learning_rate=0.18837903953853868,
     max_depth=14, min_child_weight=1, n_estimators=260,
     subsample=0.5090376818077604;, score=(train=-11.355, test=-15938.775) total
     time=
            1.8s
     [CV 3/5] END gamma=0.4938937151834346, learning_rate=0.08576454184426577,
     max_depth=9, min_child_weight=3, n_estimators=1318,
     subsample=0.8288064461501716;, score=(train=-0.277, test=-17012.799) total time=
     3.6s
     [CV 1/5] END gamma=0.5683086033354716, learning_rate=0.0687349535656185,
     max_depth=4, min_child_weight=2, n_estimators=1498,
     subsample=0.6252309093027921;, score=(train=-501.516, test=-14662.001) total
     time=
             1.8s
     [CV 5/5] END gamma=0.5683086033354716, learning rate=0.0687349535656185,
     max_depth=4, min_child_weight=2, n_estimators=1498,
     subsample=0.6252309093027921;, score=(train=-444.461, test=-16879.124) total
     time=
             1.8s
     [CV 4/5] END gamma=0.5898708475605439, learning rate=0.24577857165500183,
     max_depth=4, min_child_weight=2, n_estimators=1406,
     subsample=0.7171971827552144;, score=(train=-24.208, test=-15396.716) total
     time=
            1.7s
     [CV 4/5] END gamma=0.3500784076946757, learning rate=0.17902067240611297,
     max_depth=2, min_child_weight=1, n_estimators=1119,
     subsample=0.8612260576307527;, score=(train=-3792.814, test=-13736.330) total
     time=
             0.7s
     [CV 3/5] END gamma=0.2807723624408558, learning rate=0.05486319328629077,
     max_depth=2, min_child_weight=1, n_estimators=1295,
     subsample=0.9702292921764571;, score=(train=-7277.462, test=-16010.043) total
     time=
             0.8s
     [CV 2/5] END gamma=0.9539285770025874, learning_rate=0.23297287804408973,
     max_depth=4, min_child_weight=1, n_estimators=1160,
     subsample=0.9641592812938626;, score=(train=-25.959, test=-17417.986) total
     time=
             1.4s
     [CV 1/5] END gamma=0.42818414831731433, learning_rate=0.24333096380873392,
     max depth=4, min child weight=1, n estimators=1239,
     subsample=0.6472244460347929;, score=(train=-36.121, test=-16267.581) total
     time=
            1.6s
     [CV 5/5] END gamma=0.42818414831731433, learning_rate=0.24333096380873392,
     max_depth=4, min_child_weight=1, n_estimators=1239,
     subsample=0.6472244460347929;, score=(train=-5.010, test=-17668.675) total time=
     1.5s
[57]: import gradio as gr
      def create_dummies(df, columns):
          for column in columns:
              if column in df.columns:
```

```
dummies = pd.get_dummies(df[column], prefix=column, drop_first=True)
                  df = pd.concat([df, dummies], axis=1)
                  df.drop([column], axis=1, inplace=True)
          return df
      # Funksjon for å gjøre prediksjoner
      def predict_price(input_file):
          # Les inn data fra CSV-filen
          input_df = pd.read_csv(input_file.name) # Tilgang til filnavnet fra_
       \hookrightarrow filobjektet
          # Bruk modellen til å gjøre prediksjon
          prediction = best_model.predict(input_df)
          return prediction[0]
      # Opprett Gradio-grensesnittet med en inndatakomponent for filopplasting
      iface = gr.Interface(
          fn=predict_price,
          inputs=gr.File(label="Last opp CSV-fil", type="file"), # Liste over_
       ⇒gyldige filtyper
          outputs="text"  # Du kan endre dette basert på prediksjonens output-format
      # Start Gradio-grensesnittet
      iface.launch()
     Running on local URL: http://127.0.0.1:7870
     To create a public link, set `share=True` in `launch()`.
     <IPython.core.display.HTML object>
[57]:
 []:
```