

```
In [295]: import numpy as np
import pandas as pd
import matplotlib as mpl
import matplotlib.pyplot as plt
import mglearn
%matplotlib inline
import seaborn as sns
import platform
from matplotlib import font_manager , rc

if platform.system() == 'Darwin':
    rc('font' , family = 'AppleGothic')
elif platform.system() == 'Windows':
    path = 'C:/Windows/Fonts/malgun.ttf'
    font_name = font_manager.FontProperties(fname = path).get_name()
    rc('font' , family = font_name)
else:
    print('모름')
plt.rcParams['axes.unicode_minus'] = False
import warnings
warnings.filterwarnings('ignore')
from sklearn.metrics import accuracy_score , precision_score , recall_score , r
```



executed in 21ms, finished 16:47:26 2023-11-01

```
In [296]: path = 'C:/k_digital/machine/source/house-prices-advanced-regression-techniques'
```

executed in 14ms, finished 16:47:26 2023-11-01

```
In [297]: house = pd.read_csv(path + '/train.csv')
test = pd.read_csv(path + '/test.csv')
pred = pd.read_csv(path + '/sample_submission.csv')
```

executed in 45ms, finished 16:47:26 2023-11-01

In [298]:

house

executed in 29ms, finished 16:47:26 2023-11-01

Out[298]:

	Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandC
0	1	60	RL	65.0	8450	Pave	NaN	Reg	
1	2	20	RL	80.0	9600	Pave	NaN	Reg	
2	3	60	RL	68.0	11250	Pave	NaN	IR1	
3	4	70	RL	60.0	9550	Pave	NaN	IR1	
4	5	60	RL	84.0	14260	Pave	NaN	IR1	
...	
1455	1456	60	RL	62.0	7917	Pave	NaN	Reg	
1456	1457	20	RL	85.0	13175	Pave	NaN	Reg	
1457	1458	70	RL	66.0	9042	Pave	NaN	Reg	
1458	1459	20	RL	68.0	9717	Pave	NaN	Reg	
1459	1460	20	RL	75.0	9937	Pave	NaN	Reg	

1460 rows × 81 columns

In [299]:

```
house.info()
```

executed in 30ms, finished 16:47:26 2023-11-01

```
<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	1452 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     770 non-null    object
58 GarageType      1379 non-null    object
59 GarageYrBlt     1379 non-null    float64
60 GarageFinish    1379 non-null    object
61 GarageCars      1460 non-null    int64
62 GarageArea      1460 non-null    int64
63 GarageQual      1379 non-null    object
64 GarageCond      1379 non-null    object
65 PavedDriveway   1460 non-null    object
66 WoodDeckSF      1460 non-null    int64
67 OpenPorchSF     1460 non-null    int64
68 EnclosedPorch   1460 non-null    int64
69 3SsnPorch       1460 non-null    int64
70 ScreenPorch     1460 non-null    int64
71 PoolArea        1460 non-null    int64
72 PoolQC          7 non-null      object
73 Fence           281 non-null    object
74 MiscFeature     54 non-null     object
75 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

```

In [300]: `house.MSZoning.unique()`

executed in 13ms, finished 16:47:26 2023-11-01

Out[300]: `array(['RL', 'RM', 'C (all)', 'FV', 'RH'], dtype=object)`

In [301]: `house.corr()[house.corr()['SalePrice']>0.3]['SalePrice'].index`

executed in 30ms, finished 16:47:26 2023-11-01

Out[301]: `Index(['LotFrontage', 'OverallQual', 'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'GrLivArea', 'FullBath', 'TotRmsAbvGrd', 'Fireplaces', 'GarageYrBlt', 'GarageCars', 'GarageArea', 'WoodDeckSF', 'OpenPorchSF', 'SalePrice'], dtype='object')`

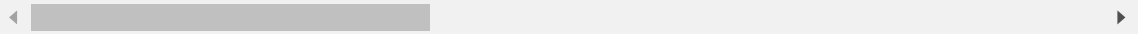
```
In [302]: house[['LotFrontage', 'OverallQual', 'YearBuilt', 'YearRemodAdd', 'MasVnrArea',
                'BsmtFinSF1', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'GrLivArea',
                'FullBath', 'TotRmsAbvGrd', 'Fireplaces', 'GarageYrBlt', 'GarageCars',
                'GarageArea', 'WoodDeckSF', 'OpenPorchSF', 'SalePrice']]
```

executed in 33ms, finished 16:47:26 2023-11-01

Out[302]:

	LotFrontage	OverallQual	YearBuilt	YearRemodAdd	MasVnrArea	BsmtFinSF1	Total
0	65.0	7	2003	2003	196.0	706	
1	80.0	6	1976	1976	0.0	978	
2	68.0	7	2001	2002	162.0	486	
3	60.0	7	1915	1970	0.0	216	
4	84.0	8	2000	2000	350.0	655	
...	
1455	62.0	6	1999	2000	0.0	0	
1456	85.0	6	1978	1988	119.0	790	
1457	66.0	7	1941	2006	0.0	275	
1458	68.0	5	1950	1996	0.0	49	
1459	75.0	5	1965	1965	0.0	830	

1460 rows × 19 columns



```
In [303]: house.drop('Id' , axis = 1 , inplace = True)
```

executed in 11ms, finished 16:47:26 2023-11-01

```
In [304]: house.describe().columns
```

executed in 56ms, finished 16:47:26 2023-11-01

```
Out[304]: Index(['MSSubClass', 'LotFrontage', 'LotArea', 'OverallQual', 'OverallCond',
                'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1', 'BsmtFinSF2',
                'BsmtUnfSF', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF',
                'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath',
                'BedroomAbvGr', 'KitchenAbvGr', 'TotRmsAbvGrd', 'Fireplaces',
                'GarageYrBlt', 'GarageCars', 'GarageArea', 'WoodDeckSF', 'OpenPorchSF',
                'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'MiscVal',
                'MoSold', 'YrSold', 'SalePrice'],
                dtype='object')
```

LowQualFinSF -> 저품질 마감된 구역이니까 , 이걸 음수로 바꾸자

```
In [305]: from sklearn.preprocessing import StandardScaler, MinMaxScaler
```

executed in 13ms, finished 16:47:26 2023-11-01

```
In [306]: mm = MinMaxScaler()
```

executed in 14ms, finished 16:47:26 2023-11-01

```
In [307]: house1 = house[['MSSubClass', 'LotFrontage', 'LotArea', 'OverallQual', 'OverallCond',
                        'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1', 'BsmtFinSF2',
                        'BsmtUnfSF', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF',
                        'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath',
                        'BedroomAbvGr', 'KitchenAbvGr', 'TotRmsAbvGrd', 'Fireplaces',
                        'GarageYrBlt', 'GarageCars', 'GarageArea', 'WoodDeckSF', 'OpenPorchSF',
                        'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'MiscVal',
                        'MoSold', 'YrSold', 'SalePrice']]
```

executed in 14ms, finished 16:47:26 2023-11-01

```
In [308]: house1.isna().sum()
```

executed in 14ms, finished 16:47:26 2023-11-01

```
Out[308]: MSSubClass      0
LotFrontage    259
LotArea        0
OverallQual    0
OverallCond    0
YearBuilt      0
YearRemodAdd   0
MasVnrArea     8
BsmtFinSF1     0
BsmtFinSF2     0
BsmtUnfSF      0
TotalBsmtSF    0
1stFlrSF       0
2ndFlrSF       0
LowQualFinSF   0
GrLivArea      0
BsmtFullBath   0
BsmtHalfBath   0
FullBath       0
HalfBath       0
BedroomAbvGr   0
KitchenAbvGr   0
TotRmsAbvGrd   0
Fireplaces     0
GarageYrBlt    81
GarageCars     0
GarageArea     0
WoodDeckSF     0
OpenPorchSF    0
EnclosedPorch  0
3SsnPorch      0
ScreenPorch    0
PoolArea       0
MiscVal        0
MoSold         0
YrSold         0
SalePrice      0
dtype: int64
```

```
In [309]: house1.LotFrontage.mean()
test.LotFrontage.mean()
```

executed in 15ms, finished 16:47:26 2023-11-01

```
Out[309]: 68.58035714285714
```

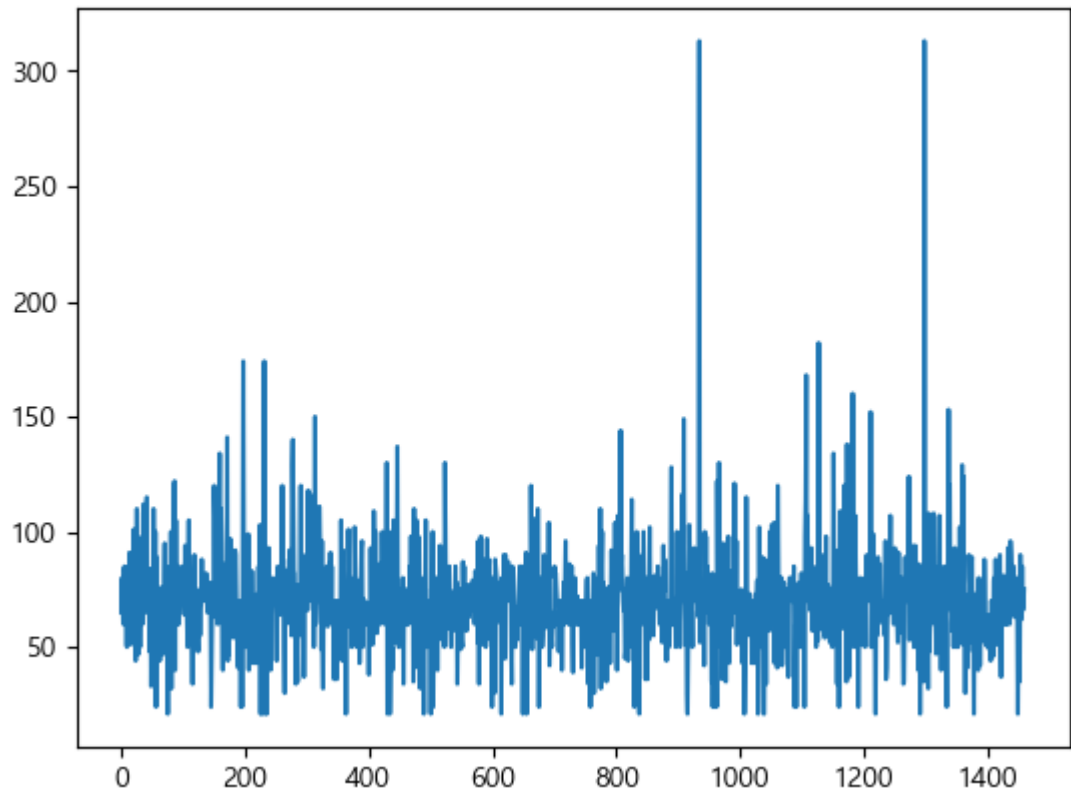
```
In [310]: house1.LotFrontage = house1.LotFrontage.fillna(70)
test.LotFrontage = test.LotFrontage.fillna(69)
```

executed in 14ms, finished 16:47:26 2023-11-01

```
In [311]: plt.plot(house1.LotFrontage)
```

executed in 140ms, finished 16:47:26 2023-11-01

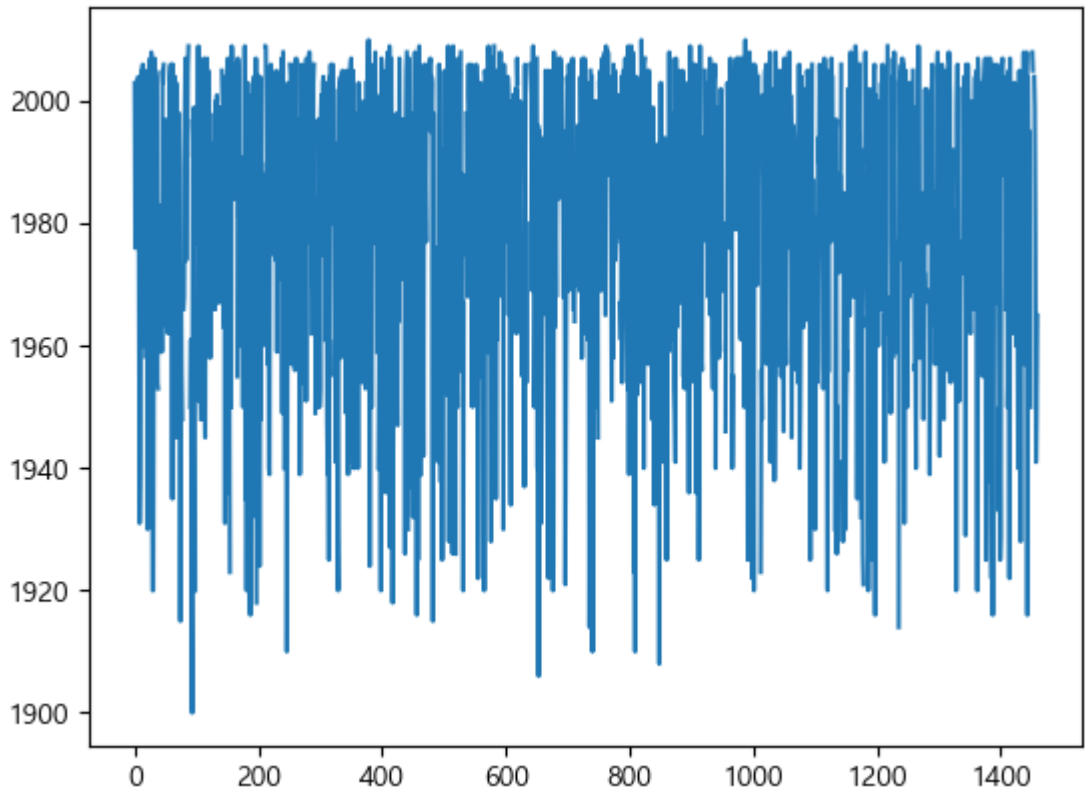
Out[311]: [<matplotlib.lines.Line2D at 0x252a461b6a0>]



In [312]: `plt.plot(house1.GarageYrBlt)`

executed in 153ms, finished 16:47:26 2023-11-01

Out[312]: [`<matplotlib.lines.Line2D at 0x252a48eb250>`]



In [313]: `test = test[['MSSubClass', 'LotFrontage', 'LotArea', 'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1', 'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'TotRmsAbvGrd', 'Fireplaces', 'GarageYrBlt', 'GarageCars', 'GarageArea', 'WoodDeckSF', 'OpenPorchSF', 'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'MiscVal', 'MoSold', 'YrSold']]`

executed in 15ms, finished 16:47:26 2023-11-01

In [314]: `house1.GarageYrBlt = house1.GarageYrBlt.fillna(house1.GarageYrBlt.mean())`
`test.GarageYrBlt = test.GarageYrBlt.fillna(test.GarageYrBlt.mean())`

executed in 13ms, finished 16:47:26 2023-11-01

In [315]: `house1.MasVnrArea = house1.MasVnrArea.fillna(0)`
`test.MasVnrArea = test.MasVnrArea.fillna(0)`

executed in 14ms, finished 16:47:26 2023-11-01

In [316]: `test = test.fillna(0)`

executed in 13ms, finished 16:47:26 2023-11-01

In []:

In []:

In [317]: target

executed in 15ms, finished 16:47:26 2023-11-01

Out[317]:

0	13.247694
1	13.109011
2	13.317167
3	12.849398
4	13.429216

...

1455	13.072541
1456	13.254863
1457	13.493130
1458	12.864462
1459	12.901583

Name: SalePrice, Length: 1460, dtype: float64

In [349]: house1.LowQualFinSF = -house1.LowQualFinSF

executed in 13ms, finished 16:49:27 2023-11-01

In [350]: data = house1.iloc[:, :-1]
target = house1.iloc[:, -1]

executed in 12ms, finished 16:49:29 2023-11-01

In [351]: data.iloc[:, 14] = -data.iloc[:, 14]

executed in 5ms, finished 16:49:29 2023-11-01

In [352]: mm.fit(data)

executed in 24ms, finished 16:49:29 2023-11-01

Out[352]:

▼ MinMaxScaler

MinMaxScaler()

In []:

In [353]: data_mm = mm.transform(data)

executed in 16ms, finished 16:49:30 2023-11-01

In [354]: data_mm

executed in 17ms, finished 16:49:30 2023-11-01

```
Out[354]: array([[0.23529412, 0.15068493, 0.0334198 , ..., 0.          , 0.09090909,
        0.5          ],
       [0.          , 0.20205479, 0.03879502, ..., 0.          , 0.36363636,
        0.25          ],
       [0.23529412, 0.1609589 , 0.04650728, ..., 0.          , 0.72727273,
        0.5          ],
       ...,
       [0.29411765, 0.15410959, 0.03618687, ..., 0.16129032, 0.36363636,
        1.          ],
       [0.          , 0.1609589 , 0.03934189, ..., 0.          , 0.27272727,
        1.          ],
       [0.          , 0.18493151, 0.04037019, ..., 0.          , 0.45454545,
        0.5          ]])
```

In [355]: house1.columns

executed in 12ms, finished 16:49:34 2023-11-01

```
Out[355]: Index(['MSSubClass', 'LotFrontage', 'LotArea', 'OverallQual', 'OverallCond',
        'YearBuilt', 'YearRemodAdd', 'MasVnrArea', 'BsmtFinSF1', 'BsmtFinSF2',
        'BsmtUnfSF', 'TotalBsmtSF', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF',
        'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath',
        'BedroomAbvGr', 'KitchenAbvGr', 'TotRmsAbvGrd', 'Fireplaces',
        'GarageYrBlt', 'GarageCars', 'GarageArea', 'WoodDeckSF', 'OpenPorchSF',
        'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'MiscVal',
        'MoSold', 'YrSold', 'SalePrice'],
        dtype='object')
```

In [356]: data_mm = pd.DataFrame(data = data_mm , columns = house1.columns[:-1])

executed in 16ms, finished 16:49:35 2023-11-01

```
In [357]: from sklearn.linear_model import LogisticRegression
        lr = LogisticRegression()
```

executed in 18ms, finished 16:49:35 2023-11-01

In [358]: lr.fit(data_mm , target)

executed in 1.84s, finished 16:49:37 2023-11-01

```
Out[358]: ▾ LogisticRegression
        LogisticRegression()
```

In [359]: pred.SalePrice = lr.predict(test_mm)

executed in 14ms, finished 16:49:37 2023-11-01

In [360]: target.shape

executed in 13ms, finished 16:49:37 2023-11-01

Out[360]: (1460,)

In [361]: data.shape

executed in 13ms, finished 16:49:37 2023-11-01

Out[361]: (1460, 36)

In [362]: from sklearn.ensemble import RandomForestClassifier
rf = RandomForestClassifier()

executed in 8ms, finished 16:49:43 2023-11-01

In [363]: target.shape

executed in 5ms, finished 16:49:43 2023-11-01

Out[363]: (1460,)

In [364]: target.shape

executed in 19ms, finished 16:49:43 2023-11-01

Out[364]: (1460,)

In [365]: data = house1.iloc[:, :-1]
target = house1.iloc[:, -1]

executed in 7ms, finished 16:49:43 2023-11-01

In [366]: rf.fit(data, target)

executed in 6.04s, finished 16:49:50 2023-11-01

Out[366]:
▼ RandomForestClassifier
RandomForestClassifier()

In [367]: data_mm

executed in 29ms, finished 16:49:50 2023-11-01

Out[367]:

	MSSubClass	LotFrontage	LotArea	OverallQual	OverallCond	YearBuilt	YearRemod
0	0.235294	0.150685	0.033420	0.666667	0.500	0.949275	0.88%
1	0.000000	0.202055	0.038795	0.555556	0.875	0.753623	0.43%
2	0.235294	0.160959	0.046507	0.666667	0.500	0.934783	0.86%
3	0.294118	0.133562	0.038561	0.666667	0.500	0.311594	0.33%
4	0.235294	0.215753	0.060576	0.777778	0.500	0.927536	0.83%
...
1455	0.235294	0.140411	0.030929	0.555556	0.500	0.920290	0.83%
1456	0.000000	0.219178	0.055505	0.555556	0.625	0.768116	0.63%
1457	0.294118	0.154110	0.036187	0.666667	1.000	0.500000	0.93%
1458	0.000000	0.160959	0.039342	0.444444	0.625	0.565217	0.76%
1459	0.000000	0.184932	0.040370	0.444444	0.625	0.673913	0.25%

1460 rows × 36 columns



```
In [368]: house1.shape , test.shape
```

```
executed in 14ms, finished 16:49:50 2023-11-01
```

```
Out[368]: ((1460, 37), (1459, 36))
```

```
In [369]: data.shape
```

```
executed in 13ms, finished 16:49:50 2023-11-01
```

```
Out[369]: (1460, 36)
```

```
In [370]: test.shape
```

```
executed in 14ms, finished 16:49:50 2023-11-01
```

```
Out[370]: (1459, 36)
```

```
In [371]: test_scaled = mm.transform(test)
```

```
executed in 9ms, finished 16:49:52 2023-11-01
```

```
In [372]: test_mm = pd.DataFrame(data = test_scaled , columns = house1.columns[:-1])
```

```
executed in 15ms, finished 16:49:53 2023-11-01
```

```
In [373]: pred.to_csv('house.csv' , index = False)
```

```
executed in 9ms, finished 16:49:53 2023-11-01
```

```
In [ ]:
```

```
In [342]: # 문자열을 수치형 데이터로 바꿔주는 함수  
from sklearn.preprocessing import LabelEncoder  
le = LabelEncoder()
```

```
executed in 14ms, finished 16:47:35 2023-11-01
```

```
In [ ]:
```