

6주차 Basic 발표

김찬원

df.isnull().sum()

```
In [11]: df.isnull().sum() #널 값이 포함되어 있는 요소들을 구한다
```

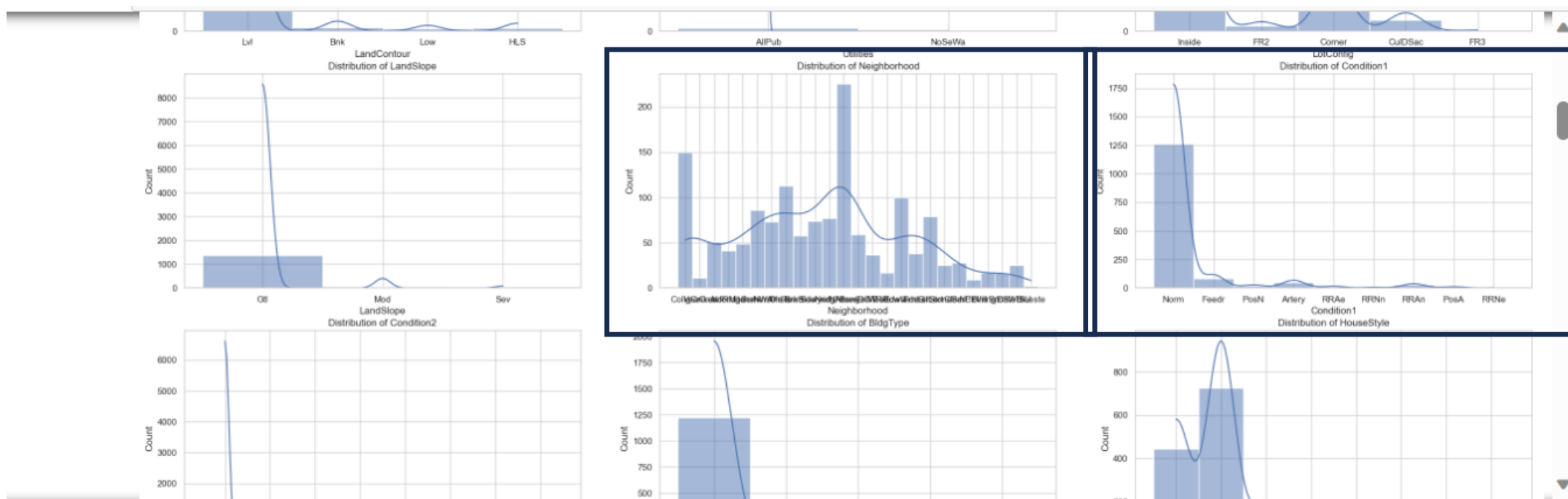
```
Out[11]: Id                0  
         MSSubClass        0  
         MSZoning          0  
         LotFrontage      259  
         LotArea           0  
         ...  
         MoSold            0  
         YrSold            0  
         SaleType          0  
         SaleCondition     0  
         SalePrice         0  
         Length: 81, dtype: int64
```

```
In [12]: df = df.dropna(axis=1, how='any') #dropna 함수를 통해 이것들을 제거한다 axis=1이므로 column을 제거한다는 뜻
```

```
In [13]: df.isnull().sum()
```

```
Out[13]: Id                0  
         MSSubClass        0  
         MSZoning          0  
         LotArea           0  
         Street            0  
         ..  
         MoSold            0  
         YrSold            0  
         SaleType          0  
         SaleCondition     0  
         SalePrice         0  
         Length: 62, dtype: int64
```

Training에 필요 없는 column 제거



we can clearly see that there are outlier present in our dataset

```
In [20]: #여기서 특정 항에만 너무 많은 값들이 모여 training에 도움이 되지 않는 column들을 구해 이들을 추후에 빼버린다.
column_are_not_helpful_in_prediction=ptr_df[['Street', 'LandContour', 'Utilities', 'LandSlope', 'Condition1', 'Condition2', 'RoofMatl',
'ExterCond', 'BsmtFinSF2', 'Heating', 'CentralAir', 'LowQualFinSF', 'BsmtHalfBath',
'KitchenAbvGr', 'Functional', 'PavedDrive', 'EnclosedPorch', '3SsnPorch', 'ScreenPorch',
'PoolArea', 'MiscVal', 'SaleType', 'SaleCondition', 'BldgType']]

column_are_not_helpful_in_prediction
```

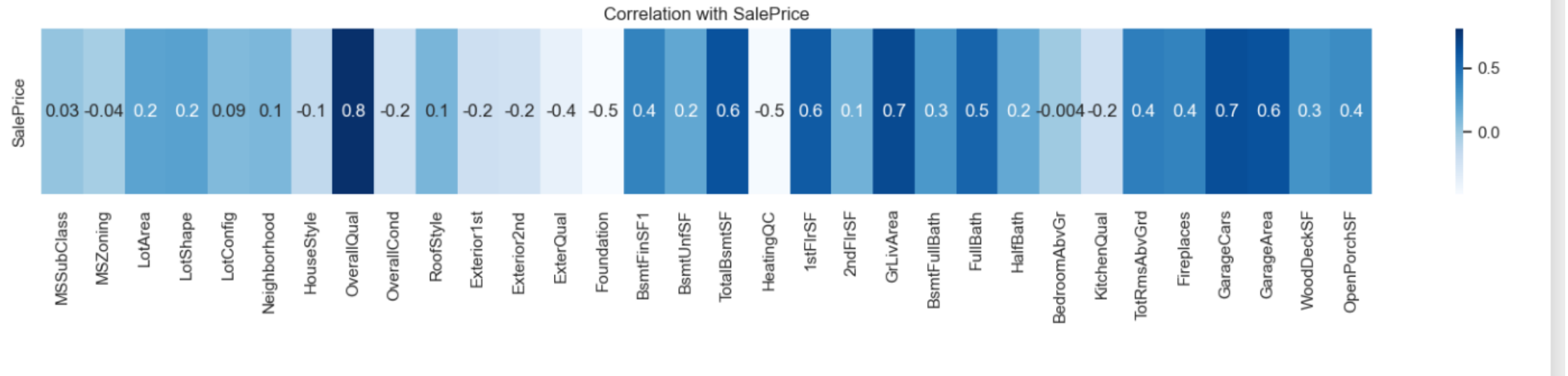
factorize_categorical_columns(column)

```
In [26]: def factorize_categorical_columns(column): #값들이 숫자가 아니라 문자로 된 값들을 숫자로 바꿔준다
          if column.dtype == 'object':
              column_encoded, _ = pd.factorize(column)
              return column_encoded
          return column

# Apply factorize only to categorical columns
df_encoded = removed_outlier.apply(factorize_categorical_columns)

# 'df_encoded' now contains the encoded values for categorical columns
```

correlation



Baseline 코드 성능

```
In [46]: r2=reg.score(X,y)  
r2
```

```
Out[46]: 0.8460515711954664
```

```
In [47]: X.shape
```

```
Out[47]: (1014, 20)
```

```
In [48]: r2=reg.score(X, y)  
n=X.shape[0]  
p=X.shape[1]  
adj_r2=1-(1-r2)*(n-1)/(n-p-1)  
adj_r2
```

```
Out[48]: 0.8429508979063519
```

성능개선1.

Training에 불필요하다가 생각하여 지운 column들 중 data분포가 너무 치우치지 않은 column 몇 개를 살림

```
In [78]: r2=reg.score(X,y)
         r2
```

```
Out[78]: 0.8567979091155806
```

```
In [79]: X.shape
```

```
Out[79]: (1014, 24)
```

```
In [80]: r2=reg.score(X, y)
         n=X.shape[0]
         p=X.shape[1]
         adj_r2=1-(1-r2)*(n-1)/(n-p-1)
         adj_r2
```

```
Out[80]: 0.853322833098163
```

성능개선2.

DescisionTreeClassifier 사용

In [195]:

```
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import VotingClassifier
from sklearn.metrics import mean_squared_error, r2_score

reg = DecisionTreeClassifier()
reg.fit(X,y)
```

Out[195]:

```
▼ DecisionTreeClassifier
DecisionTreeClassifier()
```

In [210]:

```
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import VotingClassifier
from sklearn.metrics import mean_squared_error, r2_score

model = DecisionTreeClassifier()
model.fit(X,y)
```

Out[210]:

```
▼ DecisionTreeClassifier
DecisionTreeClassifier()
```


최종결과



submission.csv

Complete · 2m ago

0.2615



submission.csv

Complete · 4m ago · 6

0.2811



submission.csv

Complete · 6m ago · 4

0.28463



submission.csv

Complete · 9m ago · 3

0.60819



submission.csv

Complete · 25m ago · 2

0.53875



submission.csv

Complete · 26m ago · 1

0.60819