

PS3

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0.1 Problem set 3

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0.3 Link to your PS3 github repo: [<https://github.com/IsabellaTan/Brown-DATA1030-HW3.git>]

0.3.1 Problem 0

-2 points for every missing green OK sign.

Make sure you are in the DATA1030 environment.

```
[1]: from __future__ import print_function
from packaging.version import parse as Version
from platform import python_version

OK = '\x1b[42m[ OK ]\x1b[0m'
FAIL = "\x1b[41m[FAIL]\x1b[0m"

try:
    import importlib
except ImportError:
    print(FAIL, "Python version 3.12.10 is required,"
          " but %s is installed." % sys.version)

def import_version(pkg, min_ver, fail_msg=""):
    mod = None
    try:
        mod = importlib.import_module(pkg)
        if pkg in {'PIL'}:
            ver = mod.VERSION
        else:
            ver = mod.__version__
        if Version(ver) == Version(min_ver):
            print(OK, "%s version %s is installed."
                  % (lib, min_ver))
        else:
            print(FAIL, "%s version %s is required, but %s installed."
                  % (lib, min_ver, ver))
```

```

except ImportError:
    print(FAIL, '%s not installed. %s' % (pkg, fail_msg))
return mod

# first check the python version
pyversion = Version(python_version())

if pyversion >= Version("3.12.10"):
    print(OK, "Python version is %s" % pyversion)
elif pyversion < Version("3.12.10"):
    print(FAIL, "Python version 3.12.10 is required,"
          " but %s is installed." % pyversion)
else:
    print(FAIL, "Unknown Python version: %s" % pyversion)

print()
requirements = {'numpy': "2.2.5", 'matplotlib': "3.10.1", 'sklearn': "1.6.1",
               'pandas': "2.2.3", 'xgboost': "3.0.0", 'shap': "0.47.2",
               'polars': "1.27.1", 'seaborn': "0.13.2"}

# now the dependencies
for lib, required_version in list(requirements.items()):
    import_version(lib, required_version)

```

[OK] Python version is 3.12.10

[OK] numpy version 2.2.5 is installed.
 [OK] matplotlib version 3.10.1 is installed.
 [OK] sklearn version 1.6.1 is installed.
 [OK] pandas version 2.2.3 is installed.
 [OK] xgboost version 3.0.0 is installed.
 [OK] shap version 0.47.2 is installed.
 [OK] polars version 1.27.1 is installed.
 [OK] seaborn version 0.13.2 is installed.

0.4 Problem 1: EDA and visualizations

0.4.1 Problem 1a: EDA (5 points)

One of the datasets we will be working with this semester is the kaggle house price dataset. The goal of PS3 is to use this dataset to practice dataframe manipulations and perform EDA. The dataset, and its description, are located in the `data` folder.

Carefully read the dataset description. Whenever you work with a dataset, it is highly recommended that you prepare a similar description if it is not readily available. Specific things to note:

- each feature is described in full detail,

- the meaning of continuous features is explained and their unit is provided (e.g., lot size is measured in square feet),
- each category in a categorical or ordinal feature is spelled out and explained.

Answer the following EDA-related questions.

The sequence of questions here are typical things to ask when you perform EDA on a new dataset. First, you always want to know how many data points and features you have, and whether they are continuous, ordinal, or categorical. You should then take a closer look at the target variable. We will study the properties of the features and the relationships between the features and the target variable in 1b.

Q0 First, read the data into a data frame and display the columns of the data frame below. You might encounter error messages and other issues along the way. Please diagnose and resolve them.

```
[ ]: # your code here

import pandas as pd
df = pd.read_excel('C:/Users/DELL/Desktop/zy/OneDrive/Brown/DATA1030/
assignment3/Brown-DATA1030-HW3/data/train.xlsx', sheet_name='data')
print(df.columns)

Index(['MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street', 'Alley',
      'LotShape', 'LandContour', 'Utilities', 'LotConfig', 'LandSlope',
      'Neighborhood', 'Condition1', 'Condition2', 'BldgType', 'HouseStyle',
      'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd', 'RoofStyle',
      'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType', 'MasVnrArea',
      'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual', 'BsmtCond',
      'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1', 'BsmtFinType2',
      'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating', 'HeatingQC',
      'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF',
      'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath',
      'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual', 'TotRmsAbvGrd',
      'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType', 'GarageYrBlt',
      'GarageFinish', 'GarageCars', 'GarageArea', 'GarageQual', 'GarageCond',
      'PavedDrive', 'WoodDeckSF', 'OpenPorchSF', 'EnclosedPorch', '3SsnPorch',
      'ScreenPorch', 'PoolArea', 'PoolQC', 'Fence', 'MiscFeature', 'MiscVal',
      'MoSold', 'YrSold', 'SaleType', 'SaleCondition', 'SalePrice'],
      dtype='object')
```

Q1 How many rows and columns do we have in the dataframe?

```
[ ]: # your code here

print("The number of row is " + str(df.shape[0]))
print("The number of column is " + str(df.shape[1]))
```

The number of row is 1460

The number of column is 80

Q2 What are the data types of the columns? Make sure that the output is not truncated and you

see the type of each column.

```
[ ]: # your code here

# Create a for loop to iterate through each column and print its name and data type
for col, dtype in df.dtypes.items():
    print(f"{col}: {dtype}")
```

```
MSSubClass: int64
MSZoning: object
LotFrontage: float64
LotArea: int64
Street: object
Alley: object
LotShape: object
LandContour: object
Utilities: object
LotConfig: object
LandSlope: object
Neighborhood: object
Condition1: object
Condition2: object
BldgType: object
HouseStyle: object
OverallQual: int64
OverallCond: int64
YearBuilt: int64
YearRemodAdd: int64
RoofStyle: object
RoofMatl: object
Exterior1st: object
Exterior2nd: object
MasVnrType: object
MasVnrArea: float64
ExterQual: object
ExterCond: object
Foundation: object
BsmtQual: object
BsmtCond: object
BsmtExposure: object
BsmtFinType1: object
BsmtFinSF1: int64
BsmtFinType2: object
BsmtFinSF2: int64
BsmtUnfSF: int64
TotalBsmtSF: int64
Heating: object
```

HeatingQC: object
CentralAir: object
Electrical: object
1stFlrSF: int64
2ndFlrSF: int64
LowQualFinSF: int64
GrLivArea: int64
BsmtFullBath: int64
BsmtHalfBath: int64
FullBath: int64
HalfBath: int64
BedroomAbvGr: int64
KitchenAbvGr: int64
KitchenQual: object
TotRmsAbvGrd: int64
Functional: object
Fireplaces: int64
FireplaceQu: object
GarageType: object
GarageYrBlt: float64
GarageFinish: object
GarageCars: int64
GarageArea: int64
GarageQual: object
GarageCond: object
PavedDrive: object
WoodDeckSF: int64
OpenPorchSF: int64
EnclosedPorch: int64
3SsnPorch: int64
ScreenPorch: int64
PoolArea: int64
PoolQC: object
Fence: object
MiscFeature: object
MiscVal: int64
MoSold: int64
YrSold: int64
SaleType: object
SaleCondition: object
SalePrice: int64

Q3 The ML target variable in this dataset is the sale price. We will develop ML pipelines to predict this variable based on the other features.

Is this column continuous or categorical? Please use `.describe` or `.value_counts` to take a quick look at this feature.

```
[ ]: # your code here

print(df["SalePrice"].describe())
# 'SalePrice' is continuous variable
```

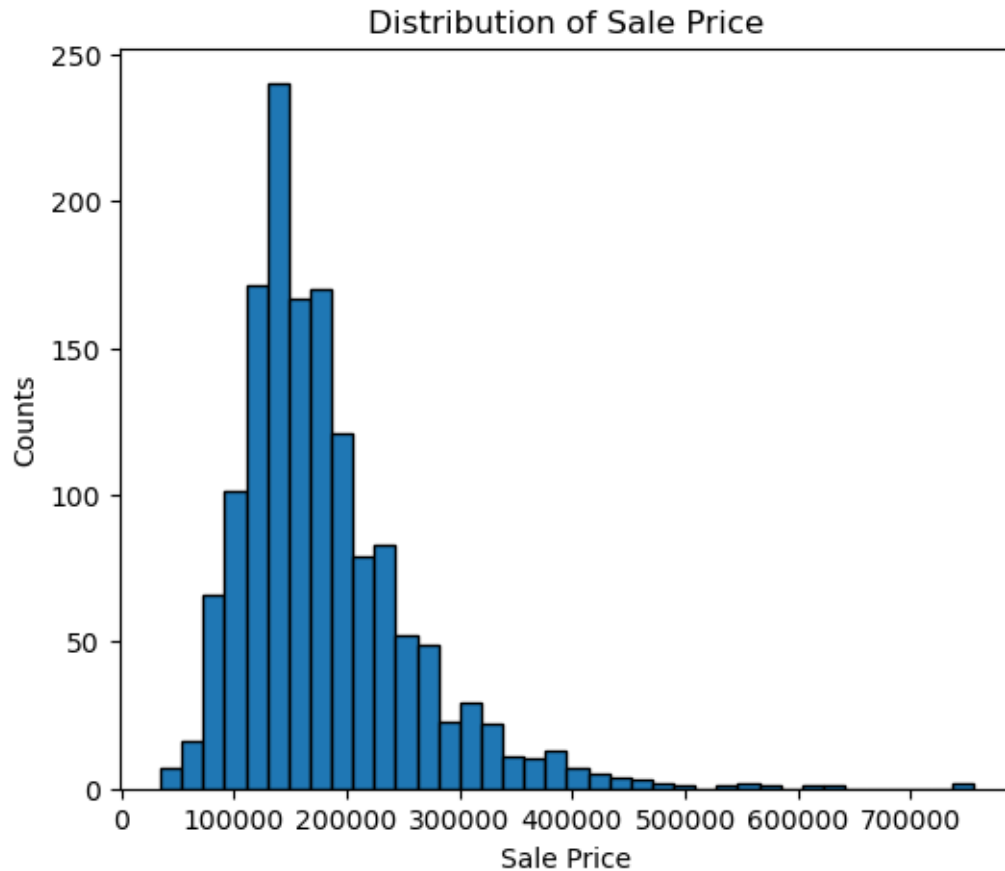
```
count      1460.000000
mean       180921.195890
std        79442.502883
min        34900.000000
25%       129975.000000
50%       163000.000000
75%       214000.000000
max        755000.000000
Name: SalePrice, dtype: float64
```

Q4 Visualize the target variable. Don't forget the axis labels and graph title. Make sure to use appropriate arguments to best display the data.

```
[ ]: # your code here

import matplotlib
from matplotlib import pylab as plt
import numpy as np
# Set the figure size
plt.figure(figsize=(6,5))
# Create the histogram
df['SalePrice'].plot.hist(
    bins = int(np.sqrt(df.shape[0])), # Let bins be the square root of the
    ↪number of rows
    edgecolor='black', # Color of the edge of the bars
    linewidth=1 # Width of the edge of the bars
)
# Add labels and title
plt.xlabel('Sale Price')
plt.ylabel('Counts')
plt.title('Distribution of Sale Price')

plt.show()
```



0.4.2 Problem 1b: visualization (10 points)

Find one continuous, one ordinal, and one categorical feature that strongly correlates with the sale price. Create figures that illustrate your selected features and the sale price.

Don't forget to add axis labels and titles, and find appropriate arguments. Write figure captions to explain what the figure shows.

If you know how to quantitatively assess correlation strengths between variables, feel free to use those techniques. Qualitative/visual assessment also works for now.

```
[ ]: # your code here
import matplotlib.pyplot as plt
import seaborn as sns

# Features that we choose:
# continuous feature : GrLivArea which is Above grade (ground) living area
#    ↳ square feet
# ordinal feature: OverallQual which is Overall material and finish of the house
# categorical feature: Alley which is Type of alley access to property
```

```

# Create a scatter plot for GrLivArea vs SalePrice
df.plot.scatter('GrLivArea', 'SalePrice', alpha=0.1, s=10, figsize=(6,5))
# Add labels and title
plt.xlabel('Above grade (ground) living area square feet')
plt.ylabel('Sale Price')
plt.title('Scatter plot of Sale Price vs Above grade (ground) living area_
↪square feet', weight='bold')
# Add a regression line
sns.regplot(x='GrLivArea', y='SalePrice', data=df, scatter_kws={'alpha':0.1,
↪'s':10}, line_kws={'color':'black'})

plt.show()
# Calculate the Pearson correlation coefficient
corr1 = round(df['GrLivArea'].corr(df['SalePrice']),2)

print('The Pearson correlation coefficient between GrLivArea and SalePrice is_
↪'+ str(corr1))
print('Based on the scatter plot and the correlation coefficient, there is a_
↪strong positive linear relationship between GrLivArea and SalePrice. As the_
↪above grade living area increases, the sale price tends to increase as well.
↪')

# Create a dictionary to map the ordinal values to their corresponding labels
cond_labels = {
    1: 'Very Poor',
    2: 'Poor',
    3: 'Fair',
    4: 'Below Average',
    5: 'Average',
    6: 'Above Average',
    7: 'Good',
    8: 'Very Good',
    9: 'Excellent',
    10: 'Very Excellent'}
# Create a box plot for OverallQual vs SalePrice
df[['OverallQual', 'SalePrice']].boxplot(by='OverallQual', figsize=(5,5))
# Add labels and title
plt.ylabel('Sale Price')
plt.xlabel('Overall material and finish of the house')

```



```

plt.title('Box plot of Sale Price vs Overall material and finish of the house',
         weight='bold')
# Delete the automatic 'Boxplot grouped by group_by_column' title
plt.suptitle('')
# Add the mapping of ordinal values to their corresponding labels on the right
side of the plot
textstr = '\n'.join([f"{k}: {v}" for k,v in sorted(cond_labels.items())])
plt.gcf().text(0.95, 0.5, textstr, fontsize=10, va='center',
             bbox=dict(facecolor='white', alpha=0.5))

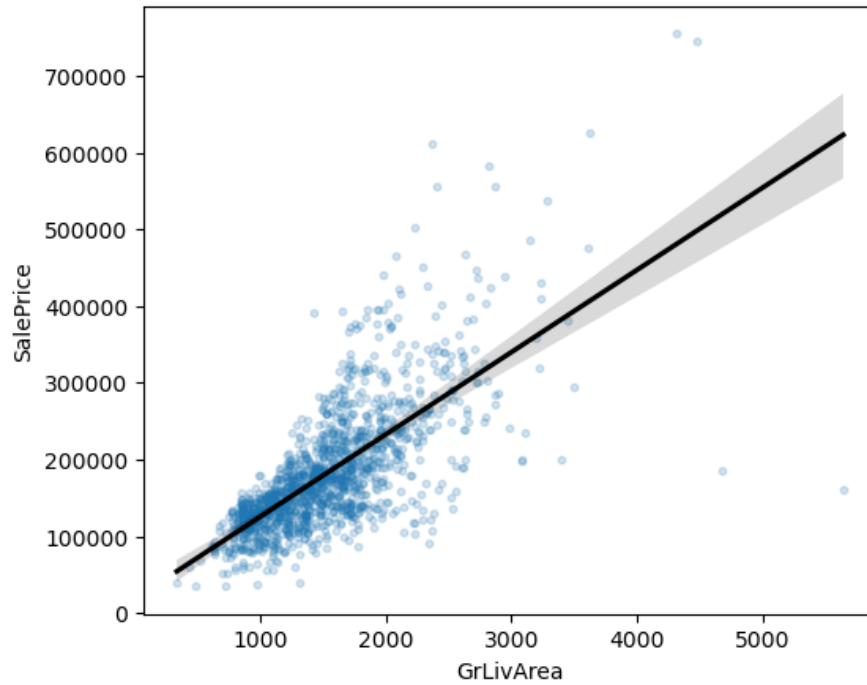
plt.show()
# Calculate the Spearman correlation
corr2 = round(df['OverallQual'].corr(df['SalePrice'], method='spearman'),2)

print('The Spearman correlation coefficient between OverallQual and SalePrice
is ' + str(corr2))
print('Based on the box plot and the correlation coefficient, there is a strong
positive monotonic relationship between OverallQual and SalePrice. As the
overall quality of the house increases, the sale price tends to increase as
well.')

# Create a dictionary to map the categorical values to their corresponding
labels
alley_labels = {
    'Grvl': 'Gravel',
    'Pave': 'Paved'}
# Create a violin plot for Alley vs SalePrice
plt.figure(figsize=(5,5))
sns.violinplot(x='Alley', y='SalePrice', data=df, color='lightblue')
# Add labels and title
plt.ylabel('Sale Price')
plt.xlabel('Type of alley access to property')
plt.title('Violin plot of Sale Price vs Type of alley access to property',
         weight='bold')
# Add the mapping of categorical values to their corresponding labels on the
right side of the plot
textstr2 = '\n'.join([f"{k}: {v}" for k,v in sorted(alley_labels.items())])
plt.gcf().text(0.95, 0.5, textstr2, fontsize=10, va='center',
             bbox=dict(facecolor='white', alpha=0.5))
plt.show()
print('Based on the violin plot, houses with paved alley access tend to have
higher sale prices compared to those with gravel alley access. The type of
alley access appears to have an impact on the sale price of the house.')

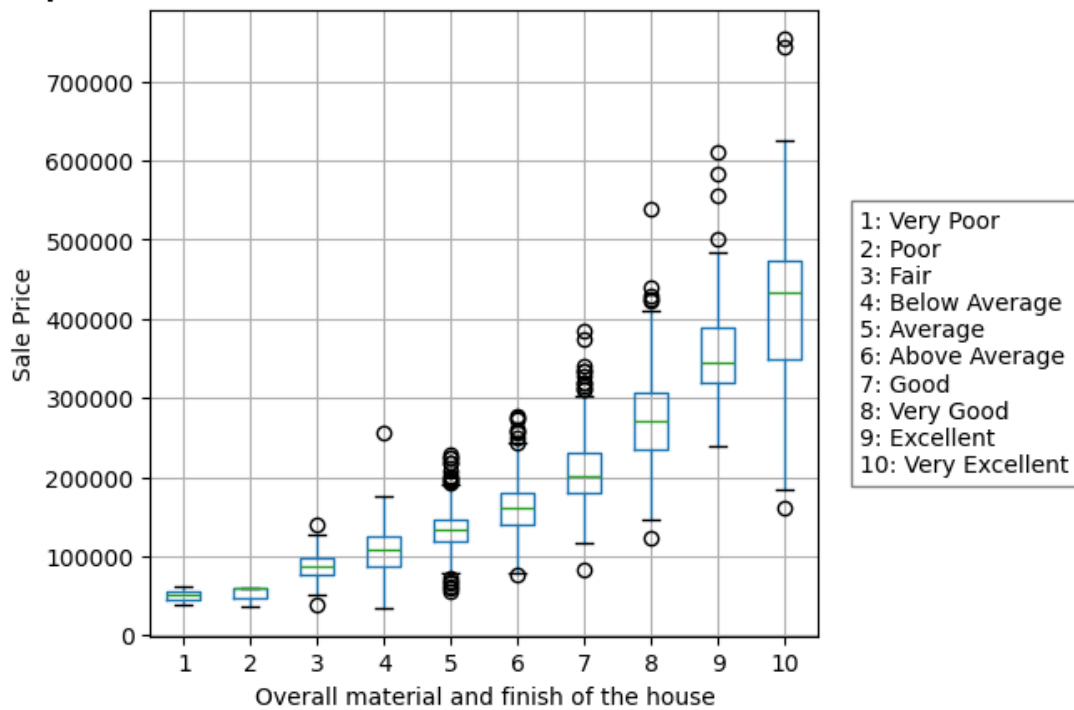
```

Scatter plot of Sale Price vs Above grade (ground) living area square feet



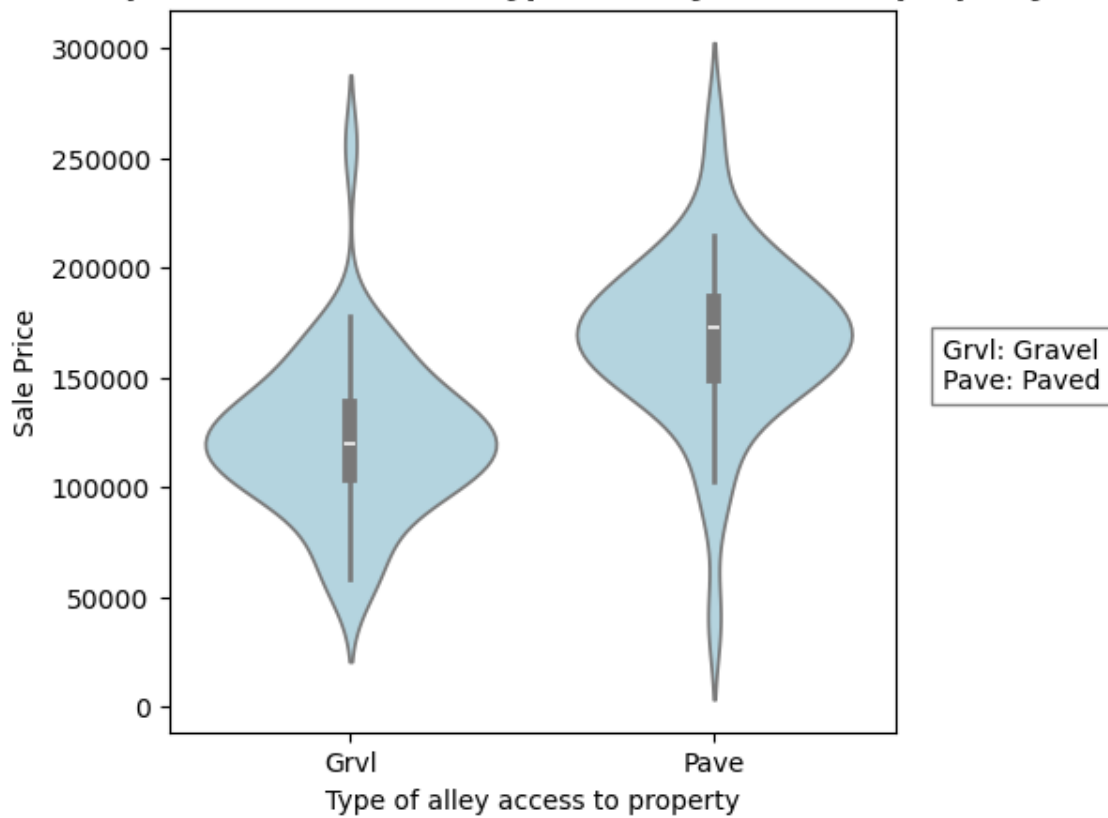
The Pearson correlation coefficient between GrLivArea and SalePrice is 0.71. Based on the scatter plot and the correlation coefficient, there is a strong positive linear relationship between GrLivArea and SalePrice. As the above grade living area increases, the sale price tends to increase as well.

Box plot of Sale Price vs Overall material and finish of the house



The Spearman correlation coefficient between OverallQual and SalePrice is 0.81. Based on the box plot and the correlation coefficient, there is a strong positive monotonic relationship between OverallQual and SalePrice. As the overall quality of the house increases, the sale price tends to increase as well.

Violin plot of Sale Price vs Type of alley access to property



Based on the violin plot, houses with paved alley access tend to have higher sale prices compared to those with gravel alley access. The type of alley access appears to have an impact on the sale price of the house.

0.5 Problem 2: basic splitting strategy (15 points)

Write a general function that performs basic splitting on a dataset, while also conducting integrity tests on both its inputs and outputs. The function is called `basic_split`, it is outlined in the cell below. It takes the following arguments as inputs: feature matrix (`X`), a target variable (`y`), `train_size`, `val_size`, `test_size`, and `random_state`. The output of the function should be: `X_train`, `y_train`, `X_val`, `y_val`, `X_test`, `y_test`, the three sets split according to the train, val, test sizes.

This function is general purpose, you'll be able to reuse it for any project if you want to perform basic split on your data.

```
[91]: def basic_split(feature_matrix, target_variable, train_size = 0.6, val_size=0.
      ↪ 2, test_size=0.2, random_state=42):
      """
      Split dataframes (feature matrix X and target variable y) into random
      ↪ train, validation and test sets
```

Parameters:

feature_matrix: a dataframe that contains your feature matrix
target_variable: a series that contains your target variable
train_size: a float between 0.0 and 1.0, it represents the proportion of the dataset to include in the training set
val_size: a float between 0.0 and 1.0, it represents the proportion of the dataset to include in the validation set
test_size: a float between 0.0 and 1.0, it represents the proportion of the dataset to include in the test set
random_state: an int, it controls the shuffling applied to the data before applying the split

NOTE: train_size+val_size+test_size must be equal to 1.

Returns:

a tuple containing the train, validation, and test sets

Example:

```
>>> import numpy as np
>>> from sklearn.model_selection import train_test_split
>>>
>>> X, y = np.arange(10).reshape((5, 2)), range(5)
>>>
>>> X_train, y_train, X_val, y_val, X_test, y_test = basic_split(X,y)

'''

# *****
# TODO: test the inputs first
# *****
# write an if statement to perform each of these checks
# Important: raise a ValueError with a descriptive error message if
↳ something is off
# call basic_split with incorrect arguments to make sure all of the tests
↳ work as intended!
import pandas as pd
import polars as pl

# test 1: if feature_matrix is not a dataframe (pandas or polars), raise
↳ ValueError
if not isinstance(feature_matrix, (pd.DataFrame, pl.DataFrame)):
    raise ValueError("Feature_matrix is not a dataframe (pandas or polars)")
```

```

# test 2: if the target-variable is not a series (pandas or polars), raise
↳ ValueError
if not isinstance(target_variable, (pd.Series, pl.Series)):
    raise ValueError("Target_variable is not a series (pandas or polars)")
# test 3: if the number of rows in feature_matrix is not equal to the
↳ length of target_variable, raise a ValueError
if feature_matrix.shape[0] != len(target_variable):
    raise ValueError("The number of rows in feature_matrix is not equal to
↳ the length of target_variable")
# test 4: if train_size is less than 0.0 or larger than 1.0, raise
↳ ValueError
if train_size < 0.0 or train_size > 1.0:
    raise ValueError("train_size is less than 0.0 or larger than 1.0")
# test 5: if val_size is less than 0.0 or larger than 1.0, raise ValueError
if val_size < 0.0 or val_size > 1.0:
    raise ValueError("val_size is less than 0.0 or larger than 1.0")
# test 6: if test_size is less than 0.0 or larger than 1.0, raise ValueError
if test_size < 0.0 or test_size > 1.0:
    raise ValueError("test_size is less than 0.0 or larger than 1.0")
# test 7: if train_size+val_size+test_size is not equal to 1.0, raise
↳ ValueError
if train_size + val_size + test_size != 1.0:
    raise ValueError("train_size+val_size+test_size is not equal to 1.0")
# test 8: if random state is not an integer, raise ValueError
if not isinstance(random_state, int):
    raise ValueError("random_state is not an integer")

# *****
# TODO: implement the splitting strategy
# *****
# as we discussed in class, use sklearn's train_test_split twice
from sklearn.model_selection import train_test_split
X_train, X_other, y_train, y_other = train_test_split(feature_matrix,
↳ target_variable, train_size=train_size, random_state=random_state)
X_val, X_test, y_val, y_test = train_test_split(X_other, y_other,
↳ test_size=test_size/(test_size + val_size), random_state=random_state)

# *****
# TODO: test the outputs
# *****
# same as above, write an if statement to perform each of these checks
# raise a ValueError with a descriptive error message if something is off.

```

```

    # test 1: the number of rows in X_train divided by the number of rows in X
    ↪should be close to train_size
    # think why we sometimes cannot achieve exact equality
    # and how you should express this as a condition in the if statement
    if not abs((X_train.shape[0] / feature_matrix.shape[0]) - train_size) < 0.
    ↪05:
        raise ValueError("The number of rows in X_train divided by the number
    ↪of rows in X is not close to train_size")
    # test 2: the number of rows in X_val divided by the number of rows in X
    ↪should be close to val_size
    if not abs((X_val.shape[0] / feature_matrix.shape[0]) - val_size) < 0.05:
        raise ValueError("The number of rows in X_val divided by the number of
    ↪rows in X is not close to val_size")
    # test 3: the number of rows in X_test divided by the number of rows in X
    ↪should be close to test_size
    if not abs((X_test.shape[0] / feature_matrix.shape[0]) - test_size) < 0.05:
        raise ValueError("The number of rows in X_test divided by the number of
    ↪rows in X is not close to test_size")
    # test 4: make sure that the length of y_train, y_val, y_test is equal to
    # the number of rows in X_train, X_val, X_test, respectively
    if not (len(y_train) == X_train.shape[0] and len(y_val) == X_val.shape[0]
    ↪and len(y_test) == X_test.shape[0]):
        raise ValueError("The length of y_train, y_val, y_test is not equal to
    ↪the number of rows in X_train, X_val, X_test, respectively")

    return X_train, y_train, X_val, y_val, X_test, y_test

# Call the function and perform tests here
# test 1: Apply the function to the house price dataset from problem 1 with
    ↪train_size = 0.6, val_size = 0.2, and test_size = 0.2.
X = df.drop(columns=['SalePrice'])
y = df['SalePrice']
X_train, y_train, X_val, y_val, X_test, y_test = basic_split(X,y)
# test 2: Print out the head of X_train, X_val, and X_test.
print("Train/Val/Test sizes:", len(X_train), len(X_val), len(X_test))
print("X_train head:\n", X_train.head())
print("X_val head:\n", X_val.head())
print("X_test head:\n", X_test.head())
# test 3: Make sure that you get the same points in each set every time you
    ↪rerun the cell (a.k.a., test for reproducibility).
X_train1, y_train1, X_val1, y_val1, X_test1, y_test1 = basic_split(X, y)
X_train2, y_train2, X_val2, y_val2, X_test2, y_test2 = basic_split(X, y)
print('Checking reproducibility:')
print('If X_train is the same:', X_train1.equals(X_train2))
print('If y_train is the same:', y_train1.equals(y_train2))

```

```

print('If X_val is the same:', X_val1.equals(X_val2))
print('If y_val is the same:', y_val1.equals(y_val2))
print('If X_test is the same:', X_test1.equals(X_test2))
print('If y_test is the same:', y_test1.equals(y_test2))
# test 4: Try a couple of other train, val, test sizes here. make sure to test
↳ the possible extreme values!
X_train_extrem, y_train_extrem, X_val_extrem, y_val_extrem, X_test_extrem,
↳ y_test_extrem = basic_split(X, y, train_size = 0.8, val_size=0.1, test_size=0.
↳ 1)
X_train_extrem2, y_train_extrem2, X_val_extrem2, y_val_extrem2, X_test_extrem2,
↳ y_test_extrem2 = basic_split(X, y, train_size = 0.9, val_size=0.
↳ 05, test_size=0.05)

# notice how most of the lines in basic_split are about testing the inputs and
↳ outputs and
# testing the inputs ensures that the user correctly calls the function. if
↳ they do not, a descriptive error message is returned.
# testing the outputs ensures that your code correctly performs the intended
↳ operation anticipating edge cases and potential issues.
# this is pretty typical in software engineering and this is the key to writing
↳ reusable code.

```

Train/Val/Test sizes: 876 292 292

X_train head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
314	70	RM	60.0	9600	Pave	Grvl	Reg	
442	50	RM	52.0	6240	Pave	NaN	Reg	
319	80	RL	NaN	14115	Pave	NaN	Reg	
767	50	RL	75.0	12508	Pave	NaN	IR1	
756	60	RL	68.0	10769	Pave	NaN	IR1	

	LandContour	Utilities	LotConfig	...	ScreenPorch	PoolArea	PoolQC	Fence	\
314	Lvl	AllPub	Inside	...	0	0	NaN	NaN	
442	Lvl	AllPub	Inside	...	0	0	NaN	NaN	
319	Lvl	AllPub	Inside	...	0	0	NaN	NaN	
767	Lvl	AllPub	Inside	...	0	0	NaN	NaN	
756	Lvl	AllPub	Inside	...	0	0	NaN	NaN	

	MiscFeature	MiscVal	MoSold	YrSold	SaleType	SaleCondition
314	NaN	0	8	2006	WD	Normal
442	NaN	0	6	2008	WD	Normal
319	NaN	0	6	2009	WD	Normal
767	Shed	1300	7	2008	WD	Normal
756	NaN	0	4	2009	WD	Normal

[5 rows x 79 columns]

X_val head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
1336	90	RL	87.0	9246	Pave	NaN	IR1	
178	20	RL	63.0	17423	Pave	NaN	IR1	
619	60	RL	85.0	12244	Pave	NaN	Reg	
548	20	RM	49.0	8235	Pave	NaN	IR1	
1046	60	RL	85.0	16056	Pave	NaN	IR1	

	LandContour	Utilities	LotConfig	...	ScreenPorch	PoolArea	PoolQC	Fence	\
1336	Lvl	AllPub	Inside	...	0	0	NaN	NaN	
178	Lvl	AllPub	CulDSac	...	0	0	NaN	NaN	
619	Lvl	AllPub	Inside	...	0	0	NaN	NaN	
548	HLS	AllPub	Inside	...	0	0	NaN	MnPrv	
1046	Lvl	AllPub	Inside	...	0	0	NaN	NaN	

	MiscFeature	MiscVal	MoSold	YrSold	SaleType	SaleCondition
1336	NaN	0	11	2008	WD	Normal
178	NaN	0	7	2009	New	Partial
619	NaN	0	8	2008	WD	Normal
548	NaN	0	6	2008	WD	Normal
1046	NaN	0	7	2006	New	Partial

[5 rows x 79 columns]

X_test head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
1290	80	RL	NaN	14112	Pave	NaN	IR1	
282	120	RL	34.0	5063	Pave	NaN	Reg	
836	30	RM	90.0	8100	Pave	Pave	Reg	
594	20	RL	88.0	7990	Pave	NaN	IR1	
544	60	RL	58.0	17104	Pave	NaN	IR1	

	LandContour	Utilities	LotConfig	...	ScreenPorch	PoolArea	PoolQC	Fence	\
1290	Lvl	AllPub	Corner	...	0	0	NaN	NaN	
282	Lvl	AllPub	Inside	...	0	0	NaN	NaN	
836	Lvl	AllPub	Inside	...	0	0	NaN	GdWo	
594	Lvl	AllPub	Inside	...	0	0	NaN	MnPrv	
544	Lvl	AllPub	Inside	...	0	0	NaN	NaN	

	MiscFeature	MiscVal	MoSold	YrSold	SaleType	SaleCondition
1290	NaN	0	4	2010	WD	Normal
282	NaN	0	4	2009	ConLw	Normal
836	NaN	0	6	2007	COD	Normal
594	NaN	0	4	2008	WD	Normal
544	NaN	0	9	2006	New	Partial

[5 rows x 79 columns]

Checking reproducibility:

If X_train is the same: True

If y_train is the same: True

If X_val is the same: True
If y_val is the same: True
If X_test is the same: True
If y_test is the same: True

1 Problem 3: kfold splitting (20 points)

Write a function that performs kfold splitting. We provided the input arguments and the header of the function. You need test the inputs, implement the algorithm, and test the outputs.

```
[ ]: def basic_kfold(feature_matrix,target_variable, k = 5, shuffle = True,
    ↪random_state = 42):
    """
    Split dataframes (feature matrix X and target variable y) into `k` number
    ↪of equal size folds.
    One fold is used as the test set.
    Iterate over the remaining k-1 folds. Fold i is used as the validation set,
    the remaining folds are used as the training set.

    Parameters:
    -----

    feature_matrix: a dataframe that contains your feature matrix
    target_variable: a series that contains your target variable
    k: an int, the number of folds
    shuffle: boolean variable. If True, the feature matrix and the target
    ↪variable are shuffled
    before the folds are created to randomize the sets
    random_state: an int, it controls the shuffling applied to the data

    Returns:
    -----

    one test set and a list with `k-1` elements containing tuples of the
    ↪train and validation sets

    >>> import numpy as np
    >>> from sklearn.model_selection import train_test_split
    >>>
    >>> X, y = np.arange(10).reshape((5, 2)), range(5)
    >>>
    >>> X_test, y_test, train_val_sets = basic_kfold(X,y)

    """
    import pandas as pd
    import polars as pl
    import numpy as np
```

```

# one of the outputs, it will contain the train and validation sets
# k-1 train and val -> tuple1(train, val), tuple2(train, val), .....
# train_val_sets = [(train1, val1) = ((x_train, y_train), (x_val, y_val))
#                     .....
#                     (traink-1, valk-1)]
#
train_val_sets = []

# *****
# TODO: test the inputs!
# *****
# test each of the input arguments. consider their types and what values
→are possible as you come up with the tests
# come up with at least 8 tests
# among other things, consider what the smallest and largest k we can have

# test 1: if feature_matrix is not a dataframe (pandas or polars), raise
→ValueError
if not isinstance(feature_matrix, (pd.DataFrame, pl.DataFrame)):
    raise ValueError("Feature_matrix is not a dataframe (pandas or polars)")
# test 2: if the target-variable is not a series (pandas or polars), raise
→ValueError
if not isinstance(target_variable, (pd.Series, pl.Series)):
    raise ValueError("Target_variable is not a series (pandas or polars)")
# test 3: if the number of rows in feature_matrix is not equal to the
→length of target_variable, raise a ValueError
if feature_matrix.shape[0] != len(target_variable):
    raise ValueError("The number of rows in feature_matrix is not equal to
→the length of target_variable")
# test 4: if k is not an integer, raise ValueError
if not isinstance(k, int):
    raise ValueError("k is not an integer")
# test 5: if shuffle is not a boolean, raise ValueError
if not isinstance(shuffle, bool):
    raise ValueError("shuffle is not a boolean")
# test 6: if random_state is not an integer, raise ValueError
if not isinstance(random_state, int):
    raise ValueError("random_state is not an integer")
# test 7: if k is less than 3 or larger than the number of rows in
→feature_matrix, raise ValueError
if k < 3 or k > feature_matrix.shape[0]:
    raise ValueError("k is less than 2 or larger than the number of rows in
→feature_matrix")
# test 8: if feature_matrix is empty, raise ValueError
if feature_matrix.shape[0] == 0 or feature_matrix.shape[1] == 0:
    raise ValueError("feature_matrix is empty")

```

```

# test 9: if target_variable is empty, raise ValueError
if len(target_variable) == 0:
    raise ValueError("target_variable is empty")

# *****
# TODO: implement the splitting strategy
# *****
# you can use numpy, pandas or polars. do not use sklearn here!

n_samples = len(feature_matrix)

# Shuffle indices if needed
indices = np.arange(n_samples)
if shuffle:
    rng = np.random.default_rng(seed=random_state)
    rng.shuffle(indices)

# Split indices into k folds
fold_sizes = [n_samples // k] * k
for i in range(n_samples % k):
    fold_sizes[i] += 1 # distribute remainder

folds = []
start = 0
for size in fold_sizes:
    folds.append(indices[start:start + size])
    start += size

# Pick first fold as test set
test_idx = folds[0]
X_test = feature_matrix.iloc[test_idx]
y_test = target_variable.iloc[test_idx]

# Remaining folds for train/validation
train_val_folds = folds[1:]

# Prepare train_val_sets
train_val_sets = []
for i, val_idx in enumerate(train_val_folds):
    # validation set
    X_val = feature_matrix.iloc[val_idx]
    y_val = target_variable.iloc[val_idx]
    # training set = all other folds
    other_idx = np.hstack([train_val_folds[j] for j in
↪range(len(train_val_folds)) if j != i])

```

```

X_train = feature_matrix.iloc[other_idx]
y_train = target_variable.iloc[other_idx]
# append tuple ((X_train, y_train), (X_val, y_val))
train_val_sets.append(((X_train, y_train), (X_val, y_val)))

# *****
# TODO: test the outputs!
# *****
# test 1: check whether each point is in exactly one set (no point is
↳ duplicated and no point is left out)
# All original sample indices
all_indices = set(range(len(feature_matrix)))
# Indices in the test set
test_indices = set(X_test.index)
# Indices in all train and validation sets
train_val_indices = set()
for train_set, val_set in train_val_sets:
    X_train, y_train = train_set # get train features and labels
    X_val, y_val = val_set # get validation features and labels
    train_val_indices.update(X_train.index) # add train indices
    train_val_indices.update(X_val.index) # add validation indices

# Check the intersection between test and train/val
if test_indices & train_val_indices:
    raise ValueError("Some points are duplicated between test and train/
↳ validation sets!")

# Check that all original samples are included
if all_indices != test_indices.union(train_val_indices):
    raise ValueError("Some points are missing from the output sets!")

# test 2: check whether you preserve the row-wise alignment between the
↳ feature matrix and the target variable in each set
# hint: for row-wise alignment, it may be useful to note that row indices
↳ do not change when subsetting dfs.
# i.e. if a row is index id 10 in dataframe 1, it is still given the index
↳ id of 10 in a subsetting df.
# Loop over each train/validation fold
for i, (train_set, val_set) in enumerate(train_val_sets):
    X_train, y_train = train_set # get features and labels for train set
    X_val, y_val = val_set # # get features and labels for validation set
    # Check that row indices match between features and labels in train set
    if not X_train.index.equals(y_train.index):
        raise ValueError(f"Train set row alignment broken in fold {i+1}")

```

```

        # Check that row indices match between features and labels in validation set
        if not X_val.index.equals(y_val.index):
            raise ValueError(f"Validation set row alignment broken in fold_{i+1}")

        # Check that row indices match between features and labels in test set
        if not X_test.index.equals(y_test.index):
            raise ValueError("Test set row alignment broken!")

# test 3: check whether the order of the columns is the same in each set.
# We don't check y because y only has 1 column
# Check train and validation
# Loop over each train/validation fold
for i, (train_set, val_set) in enumerate(train_val_sets):
    X_train, _ = train_set # get feature matrix from train set
    X_val, _ = val_set # get feature matrix from validation set
    # Check if train columns match original feature matrix
    if not X_train.columns.equals(feature_matrix.columns):
        raise ValueError(f"Train set column order broken in fold {i+1}")
    # Check if validation columns match original feature matrix
    if not X_val.columns.equals(feature_matrix.columns):
        raise ValueError(f"Validation set column order broken in fold_{i+1}")

# Check Test set
if not X_test.columns.equals(feature_matrix.columns):
    raise ValueError("Test set column order broken!")

# test 4: perform output tests similar to those in basic_split
# Check if X_test is Dataframe
if not isinstance(X_test, (pd.DataFrame, pl.DataFrame)):
    raise ValueError('The output X_test is not a Dataframe')
# Check if y-test is Series
if not isinstance(y_test, (pd.Series, pl.Series)):
    raise ValueError('The output y_test is not a Series')
# Check if train_val_sets is List
if not isinstance(train_val_sets, list):
    raise ValueError('The output train_val_sets is not a List')

return X_test, y_test, train_val_sets

```

```
#####
# TODO: call the function and perform tests here
#####
# test 1: Apply the function to the house price dataset from problem 1 with 4
↳ folds.
X = df.drop(columns=['SalePrice'])
y = df['SalePrice']
X_test, y_test, train_val_sets = basic_kfold(X,y,k = 4)

# test 2: Print out the head of the sets.
# print the head of test sets
print('Head of X_test:')
print(X_test.head())
print('Head of y_test')
print(y_test.head())
# make for loop to print the head of training and validation sets for every fold
for fold_idx, (train_set, val_set) in enumerate(train_val_sets):
    X_train, y_train = train_set
    X_val, y_val = val_set

    print(f"=== Fold {fold_idx+1} ===")

    print("X_train head:")
    print(X_train.head())

    print("y_train head:")
    print(y_train.head())

    print("X_val head:")
    print(X_val.head())

    print("y_val head:")
    print(y_val.head())

    print("\n")

# test 3: Make sure that you get the same points in each set every time you
↳ rerun the cell (a.k.a., test for reproducibility).
X_test1, y_test1, train_val_sets1 = basic_kfold(X, y)
X_test2, y_test2, train_val_sets2 = basic_kfold(X, y)

print('Checking reproducibility:')
print('If X_test is the same:', X_test1.equals(X_test2))
print('If y_test is the same:', y_test1.equals(y_test2))
```

```

# Loop over each fold in two separate runs of k-fold to check reproducibility
for i, ((train1, val1), (train2, val2)) in enumerate(zip(train_val_sets1,
↳train_val_sets2)):
    X_train1, y_train1 = train1
    X_val1, y_val1 = val1
    X_train2, y_train2 = train2
    X_val2, y_val2 = val2
    if not X_train1.index.equals(X_train2.index) or not y_train1.index.
↳equals(y_train2.index):
        raise AssertionError(f"Fold {i+1} train indices differ!")
    if not X_val1.index.equals(X_val2.index) or not y_val1.index.equals(y_val2.
↳index):
        raise AssertionError(f"Fold {i+1} validation indices differ!")
print("All train/validation fold indices are same.")

# test 4: Try a couple of other `k` values. test the extreme values!
X_test_extreme, y_test_extreme, train_val_sets_extreme = basic_kfold(X,y,k = 3)
X_test_extreme2, y_test_extreme2, train_val_sets_extreme2 = basic_kfold(X,y,k =
↳len(y))

```

Head of X_test:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
1257	30	RL	56.0	4060	Pave	NaN	Reg	
612	60	RL	NaN	11885	Pave	NaN	Reg	
1270	40	RL	NaN	23595	Pave	NaN	Reg	
1061	30	C (all)	120.0	18000	Grvl	NaN	Reg	
184	50	RL	92.0	7438	Pave	NaN	IR1	

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1	\
1257	Lvl	AllPub	Corner	Gtl	Edwards	Feedr	
612	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	
1270	Low	AllPub	Inside	Sev	ClearCr	Norm	
1061	Low	AllPub	Inside	Gtl	IDOTRR	Norm	
184	Lvl	AllPub	Inside	Gtl	BrkSide	RRAn	

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	\
1257	Norm	1Fam	1Story	5	8	1922	
612	Norm	1Fam	2Story	8	5	2001	
1270	Norm	1Fam	1Story	7	6	1979	
1061	Norm	1Fam	1Story	3	4	1935	
184	Feedr	1Fam	1.5Fin	5	8	1908	

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	\
1257	1950	Gable	CompShg	Wd Sdng	Wd Sdng	NaN	
612	2001	Gable	CompShg	VinylSd	VinylSd	BrkFace	

1270	1979	Shed	WdShake	Plywood	Plywood	NaN
1061	1950	Gable	CompShg	MetalSd	MetalSd	NaN
184	1991	Gable	CompShg	AsbShng	Plywood	NaN

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	\
1257	0.0	TA	TA	PConc	Fa	TA	
612	108.0	Gd	TA	PConc	Gd	TA	
1270	0.0	Gd	TA	PConc	Gd	TA	
1061	0.0	Fa	TA	CBlock	TA	TA	
184	0.0	TA	TA	PConc	Fa	TA	

	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	\
1257	No	Unf	0	Unf	0	
612	Av	GLQ	990	Unf	0	
1270	Gd	GLQ	1258	Unf	0	
1061	No	Unf	0	Unf	0	
184	No	Unf	0	Unf	0	

	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	\
1257	864	864	GasA	Ex	Y	SBrkr	
612	309	1299	GasA	Ex	Y	SBrkr	
1270	74	1332	GasA	TA	Y	SBrkr	
1061	894	894	GasA	TA	Y	SBrkr	
184	504	504	GasA	Gd	Y	SBrkr	

	1stFlrSF	2ndFlrSF	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	\
1257	864	0	0	864	0	0	
612	1299	573	0	1872	1	0	
1270	1332	192	0	1524	2	0	
1061	894	0	0	894	0	0	
184	936	316	0	1252	0	0	

	FullBath	HalfBath	BedroomAbvGr	KitchenAbvGr	KitchenQual	\
1257	1	0	2	1	TA	
612	2	1	3	1	Ex	
1270	0	1	0	1	Gd	
1061	1	0	2	1	TA	
184	1	0	3	1	TA	

	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	\
1257	4	Typ	0	NaN	NaN	NaN	
612	7	Typ	1	TA	BuiltIn	2001.0	
1270	4	Typ	1	TA	Attchd	1979.0	
1061	6	Typ	0	NaN	Detchd	1994.0	
184	5	Typ	0	NaN	Attchd	1986.0	

	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive	\
1257	NaN	0	0	NaN	NaN	Y	

612	RFn	2	531	TA	TA	Y
1270	Fin	2	586	TA	TA	Y
1061	RFn	3	1248	TA	TA	Y
184	Unf	2	576	TA	TA	Y

	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	\
1257	0	96	0	0	0	
612	160	122	0	0	0	
1270	268	0	0	0	0	
1061	0	20	0	0	0	
184	104	0	0	0	0	

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	\
1257	0	NaN	NaN	NaN	0	7	2009	WD	
612	0	NaN	NaN	NaN	0	11	2009	WD	
1270	0	NaN	NaN	NaN	0	4	2010	WD	
1061	0	NaN	NaN	Shed	560	8	2008	ConLD	
184	0	NaN	MnPrv	NaN	0	6	2006	WD	

	SaleCondition
1257	Normal
612	Normal
1270	Normal
1061	Normal
184	Normal

Head of y_test

1257	99900
612	261500
1270	260000
1061	81000
184	127000

Name: SalePrice, dtype: int64

=== Fold 1 ===

X_train head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
541	60	RL	NaN	11000	Pave	NaN	Reg	
341	20	RH	60.0	8400	Pave	NaN	Reg	
1074	20	RL	74.0	8556	Pave	NaN	Reg	
521	20	RL	90.0	11988	Pave	NaN	IR1	
850	120	RM	36.0	4435	Pave	NaN	Reg	

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1	\
541	Lvl	AllPub	FR2	Gtl	NoRidge	Norm	
341	Lvl	AllPub	Inside	Gtl	SawyerW	Feedr	
1074	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	
521	Lvl	AllPub	Corner	Gtl	Names	Feedr	
850	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	\
541	Norm	1Fam	2Story	8	5	2000	
341	Norm	1Fam	1Story	4	4	1950	
1074	Norm	1Fam	1Story	7	5	2006	
521	Norm	1Fam	1Story	6	6	1957	
850	Norm	TwnhsE	1Story	6	5	2003	

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	\
541	2000	Gable	CompShg	VinylSd	VinylSd	BrkFace	
341	1950	Gable	CompShg	Wd Sdng	AsbShng	NaN	
1074	2006	Gable	CompShg	VinylSd	VinylSd	NaN	
521	1957	Gable	CompShg	VinylSd	VinylSd	NaN	
850	2003	Gable	CompShg	VinylSd	VinylSd	BrkFace	

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	\
541	72.0	Gd	TA	PConc	Gd	TA	
341	0.0	Fa	Fa	CBlock	TA	Fa	
1074	0.0	Gd	TA	PConc	Gd	TA	
521	0.0	TA	TA	CBlock	TA	TA	
850	170.0	Gd	TA	PConc	Gd	TA	

	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	\
541	No	Unf	0	Unf	0	
341	No	Unf	0	Unf	0	
1074	Av	Unf	0	Unf	0	
521	No	Rec	777	Unf	0	
850	Av	GLQ	659	Unf	0	

	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	\
541	969	969	GasA	Ex	Y	SBrkr	
341	721	721	GasA	Gd	Y	SBrkr	
1074	1240	1240	GasA	Ex	Y	SBrkr	
521	467	1244	GasA	Ex	Y	FuseA	
850	189	848	GasA	Ex	Y	SBrkr	

	1stFlrSF	2ndFlrSF	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	\
541	997	1288	0	2285	0	0	
341	841	0	0	841	0	0	
1074	1240	0	0	1240	0	0	
521	1244	0	0	1244	0	0	
850	848	0	0	848	1	0	

	FullBath	HalfBath	BedroomAbvGr	KitchenAbvGr	KitchenQual	\
541	2	1	4	1	Gd	
341	1	0	2	1	TA	
1074	2	0	2	1	Gd	
521	1	1	3	1	TA	
850	1	0	1	1	Gd	

	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	\
541	8	Typ	1	TA	BuiltIn	2000.0	
341	4	Typ	0	NaN	CarPort	1950.0	
1074	5	Typ	0	NaN	Attchd	2006.0	
521	6	Typ	2	Gd	Attchd	1957.0	
850	3	Typ	0	NaN	Attchd	2003.0	

	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive	\
541	Fin	3	648	TA	TA	Y	
341	Unf	1	294	TA	TA	N	
1074	RFn	3	826	TA	TA	Y	
521	Unf	1	336	TA	TA	Y	
850	Fin	2	420	TA	TA	Y	

	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	\
541	0	56	0	0	0	
341	250	0	24	0	0	
1074	140	93	0	0	0	
521	0	40	0	0	0	
850	140	0	0	0	0	

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	\
541	0	NaN	NaN	NaN	0	6	2007	WD	
341	0	NaN	NaN	NaN	0	9	2009	WD	
1074	0	NaN	NaN	NaN	0	5	2007	WD	
521	0	NaN	NaN	NaN	0	5	2007	WD	
850	0	NaN	NaN	NaN	0	11	2007	WD	

	SaleCondition
541	Normal
341	Normal
1074	Normal
521	Normal
850	Normal

y_train head:

541	248000
341	82000
1074	194000
521	150000
850	131500

Name: SalePrice, dtype: int64

X_val head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
895	60	RL	71.0	7056	Pave	NaN	Reg	
1414	50	RL	64.0	13053	Pave	Pave	Reg	
25	20	RL	110.0	14230	Pave	NaN	Reg	
0	60	RL	65.0	8450	Pave	NaN	Reg	

959	160	FV	24.0	2572	Pave	NaN	Reg
-----	-----	----	------	------	------	-----	-----

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1	\
895	Lvl	AllPub	Inside	Gtl	Names	Norm	
1414	Bnk	AllPub	Inside	Gtl	BrkSide	Norm	
25	Lvl	AllPub	Corner	Gtl	NridgHt	Norm	
0	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	
959	Lvl	AllPub	FR2	Gtl	Somerst	Norm	

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	\
895	Norm	1Fam	2Story	6	5	1963	
1414	Norm	1Fam	1.5Fin	6	7	1923	
25	Norm	1Fam	1Story	8	5	2007	
0	Norm	1Fam	2Story	7	5	2003	
959	Norm	Twnhs	2Story	7	5	1999	

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	\
895	1963	Hip	CompShg	HdBoard	HdBoard	BrkFace	
1414	2000	Gambrel	CompShg	Wd Sdng	Wd Sdng	NaN	
25	2007	Gable	CompShg	VinylSd	VinylSd	Stone	
0	2003	Gable	CompShg	VinylSd	VinylSd	BrkFace	
959	1999	Hip	CompShg	MetalSd	MetalSd	NaN	

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	\
895	415.0	TA	TA	CBlock	TA	TA	
1414	0.0	TA	TA	BrkTil	TA	TA	
25	640.0	Gd	TA	PConc	Gd	TA	
0	196.0	Gd	TA	PConc	Gd	TA	
959	0.0	Gd	TA	PConc	Gd	TA	

	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	\
895	No	BLQ	400	Unf	0	
1414	No	Unf	0	Unf	0	
25	No	Unf	0	Unf	0	
0	No	GLQ	706	Unf	0	
959	No	ALQ	604	Unf	0	

	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	\
895	380	780	GasA	TA	Y	SBrkr	
1414	833	833	GasA	Gd	Y	SBrkr	
25	1566	1566	GasA	Ex	Y	SBrkr	
0	150	856	GasA	Ex	Y	SBrkr	
959	92	696	GasA	Ex	Y	SBrkr	

	1stFlrSF	2ndFlrSF	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	\
895	983	813	0	1796	1	0	
1414	1053	795	0	1848	0	0	
25	1600	0	0	1600	0	0	

0	856	854	0	1710	1	0
959	696	720	0	1416	1	0

	FullBath	HalfBath	BedroomAbvGr	KitchenAbvGr	KitchenQual	\
895	1	1	4	1	TA	
1414	1	1	4	1	Gd	
25	2	0	3	1	Gd	
0	2	1	3	1	Gd	
959	2	1	3	1	Gd	

	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	\
895	8	Typ	1	TA	Attchd	1963.0	
1414	8	Typ	1	Gd	Detchd	1922.0	
25	7	Typ	1	Gd	Attchd	2007.0	
0	8	Typ	0	NaN	Attchd	2003.0	
959	6	Typ	0	NaN	Detchd	1999.0	

	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive	\
895	RFn	2	483	TA	TA	Y	
1414	Unf	2	370	TA	TA	N	
25	RFn	3	890	TA	TA	Y	
0	RFn	2	548	TA	TA	Y	
959	Unf	2	484	TA	TA	Y	

	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	\
895	0	50	0	0	0	
1414	0	0	0	0	220	
25	0	56	0	0	0	
0	0	61	0	0	0	
959	0	44	0	0	0	

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	\
895	0	NaN	NaN	NaN	0	10	2008	WD	
1414	0	NaN	NaN	NaN	0	6	2008	WD	
25	0	NaN	NaN	NaN	0	7	2009	WD	
0	0	NaN	NaN	NaN	0	2	2008	WD	
959	0	NaN	NaN	NaN	0	5	2010	WD	

	SaleCondition
895	Normal
1414	Normal
25	Normal
0	Normal
959	Normal

y_val head:

895	140000
1414	207000
25	256300

```

0      208500
959    155000
Name: SalePrice, dtype: int64

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=== Fold 2 ===

X_train head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
895	60	RL	71.0	7056	Pave	NaN	Reg	
1414	50	RL	64.0	13053	Pave	Pave	Reg	
25	20	RL	110.0	14230	Pave	NaN	Reg	
0	60	RL	65.0	8450	Pave	NaN	Reg	
959	160	FV	24.0	2572	Pave	NaN	Reg	

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1	\
895	Lvl	AllPub	Inside	Gtl	Names	Norm	
1414	Bnk	AllPub	Inside	Gtl	BrkSide	Norm	
25	Lvl	AllPub	Corner	Gtl	NridgHt	Norm	
0	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	
959	Lvl	AllPub	FR2	Gtl	Somerst	Norm	

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	\
895	Norm	1Fam	2Story	6	5	1963	
1414	Norm	1Fam	1.5Fin	6	7	1923	
25	Norm	1Fam	1Story	8	5	2007	
0	Norm	1Fam	2Story	7	5	2003	
959	Norm	Twnhs	2Story	7	5	1999	

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	\
895	1963	Hip	CompShg	HdBoard	HdBoard	BrkFace	
1414	2000	Gambrel	CompShg	Wd Sdng	Wd Sdng	NaN	
25	2007	Gable	CompShg	VinylSd	VinylSd	Stone	
0	2003	Gable	CompShg	VinylSd	VinylSd	BrkFace	
959	1999	Hip	CompShg	MetalSd	MetalSd	NaN	

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	\
895	415.0	TA	TA	CBlock	TA	TA	
1414	0.0	TA	TA	BrkTil	TA	TA	
25	640.0	Gd	TA	PConc	Gd	TA	
0	196.0	Gd	TA	PConc	Gd	TA	
959	0.0	Gd	TA	PConc	Gd	TA	

	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	\
895	No	BLQ	400	Unf	0	
1414	No	Unf	0	Unf	0	
25	No	Unf	0	Unf	0	
0	No	GLQ	706	Unf	0	
959	No	ALQ	604	Unf	0	

	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	\
895	380	780	GasA	TA	Y	SBrkr	
1414	833	833	GasA	Gd	Y	SBrkr	
25	1566	1566	GasA	Ex	Y	SBrkr	
0	150	856	GasA	Ex	Y	SBrkr	
959	92	696	GasA	Ex	Y	SBrkr	

	1stFlrSF	2ndFlrSF	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	\
895	983	813	0	1796	1	0	
1414	1053	795	0	1848	0	0	
25	1600	0	0	1600	0	0	
0	856	854	0	1710	1	0	
959	696	720	0	1416	1	0	

	FullBath	HalfBath	BedroomAbvGr	KitchenAbvGr	KitchenQual	\
895	1	1	4	1	TA	
1414	1	1	4	1	Gd	
25	2	0	3	1	Gd	
0	2	1	3	1	Gd	
959	2	1	3	1	Gd	

	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	\
895	8	Typ	1	TA	Attchd	1963.0	
1414	8	Typ	1	Gd	Detchd	1922.0	
25	7	Typ	1	Gd	Attchd	2007.0	
0	8	Typ	0	NaN	Attchd	2003.0	
959	6	Typ	0	NaN	Detchd	1999.0	

	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive	\
895	RFn	2	483	TA	TA	Y	
1414	Unf	2	370	TA	TA	N	
25	RFn	3	890	TA	TA	Y	
0	RFn	2	548	TA	TA	Y	
959	Unf	2	484	TA	TA	Y	

	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	\
895	0	50	0	0	0	
1414	0	0	0	0	220	
25	0	56	0	0	0	
0	0	61	0	0	0	
959	0	44	0	0	0	

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	\
895	0	NaN	NaN	NaN	0	10	2008	WD	
1414	0	NaN	NaN	NaN	0	6	2008	WD	
25	0	NaN	NaN	NaN	0	7	2009	WD	
0	0	NaN	NaN	NaN	0	2	2008	WD	

959	0	NaN	NaN	NaN	0	5	2010	WD
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SaleCondition

895	Normal
1414	Normal
25	Normal
0	Normal
959	Normal

y_train head:

895	140000
1414	207000
25	256300
0	208500
959	155000

Name: SalePrice, dtype: int64

X_val head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
541	60	RL	NaN	11000	Pave	NaN	Reg	
341	20	RH	60.0	8400	Pave	NaN	Reg	
1074	20	RL	74.0	8556	Pave	NaN	Reg	
521	20	RL	90.0	11988	Pave	NaN	IR1	
850	120	RM	36.0	4435	Pave	NaN	Reg	

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1	\
541	Lvl	AllPub	FR2	Gtl	NoRidge	Norm	
341	Lvl	AllPub	Inside	Gtl	SawyerW	Feedr	
1074	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	
521	Lvl	AllPub	Corner	Gtl	NAMES	Feedr	
850	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	\
541	Norm	1Fam	2Story	8	5	2000	
341	Norm	1Fam	1Story	4	4	1950	
1074	Norm	1Fam	1Story	7	5	2006	
521	Norm	1Fam	1Story	6	6	1957	
850	Norm	TwnhsE	1Story	6	5	2003	

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	\
541	2000	Gable	CompShg	VinylSd	VinylSd	BrkFace	
341	1950	Gable	CompShg	Wd Sdng	AsbShng	NaN	
1074	2006	Gable	CompShg	VinylSd	VinylSd	NaN	
521	1957	Gable	CompShg	VinylSd	VinylSd	NaN	
850	2003	Gable	CompShg	VinylSd	VinylSd	BrkFace	

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	\
541	72.0	Gd	TA	PConc	Gd	TA	
341	0.0	Fa	Fa	CBlock	TA	Fa	
1074	0.0	Gd	TA	PConc	Gd	TA	

521	0.0	TA	TA	CBlock	TA	TA
850	170.0	Gd	TA	PConc	Gd	TA

	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	\
541	No	Unf	0	Unf	0	
341	No	Unf	0	Unf	0	
1074	Av	Unf	0	Unf	0	
521	No	Rec	777	Unf	0	
850	Av	GLQ	659	Unf	0	

	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	\
541	969	969	GasA	Ex	Y	SBrkr	
341	721	721	GasA	Gd	Y	SBrkr	
1074	1240	1240	GasA	Ex	Y	SBrkr	
521	467	1244	GasA	Ex	Y	FuseA	
850	189	848	GasA	Ex	Y	SBrkr	

	1stFlrSF	2ndFlrSF	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	\
541	997	1288	0	2285	0	0	
341	841	0	0	841	0	0	
1074	1240	0	0	1240	0	0	
521	1244	0	0	1244	0	0	
850	848	0	0	848	1	0	

	FullBath	HalfBath	BedroomAbvGr	KitchenAbvGr	KitchenQual	\
541	2	1	4	1	Gd	
341	1	0	2	1	TA	
1074	2	0	2	1	Gd	
521	1	1	3	1	TA	
850	1	0	1	1	Gd	

	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	\
541	8	Typ	1	TA	BuiltIn	2000.0	
341	4	Typ	0	NaN	CarPort	1950.0	
1074	5	Typ	0	NaN	Attchd	2006.0	
521	6	Typ	2	Gd	Attchd	1957.0	
850	3	Typ	0	NaN	Attchd	2003.0	

	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive	\
541	Fin	3	648	TA	TA	Y	
341	Unf	1	294	TA	TA	N	
1074	RFn	3	826	TA	TA	Y	
521	Unf	1	336	TA	TA	Y	
850	Fin	2	420	TA	TA	Y	

	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	\
541	0	56	0	0	0	
341	250	0	24	0	0	

1074	140	93	0	0	0
521	0	40	0	0	0
850	140	0	0	0	0

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	\
541	0	NaN	NaN	NaN	0	6	2007	WD	
341	0	NaN	NaN	NaN	0	9	2009	WD	
1074	0	NaN	NaN	NaN	0	5	2007	WD	
521	0	NaN	NaN	NaN	0	5	2007	WD	
850	0	NaN	NaN	NaN	0	11	2007	WD	

	SaleCondition
541	Normal
341	Normal
1074	Normal
521	Normal
850	Normal

y_val head:

541	248000
341	82000
1074	194000
521	150000
850	131500

Name: SalePrice, dtype: int64

=== Fold 3 ===

X_train head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
895	60	RL	71.0	7056	Pave	NaN	Reg	
1414	50	RL	64.0	13053	Pave	Pave	Reg	
25	20	RL	110.0	14230	Pave	NaN	Reg	
0	60	RL	65.0	8450	Pave	NaN	Reg	
959	160	FV	24.0	2572	Pave	NaN	Reg	

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1	\
895	Lvl	AllPub	Inside	Gtl	NAMES	Norm	
1414	Bnk	AllPub	Inside	Gtl	BrkSide	Norm	
25	Lvl	AllPub	Corner	Gtl	NridgHt	Norm	
0	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	
959	Lvl	AllPub	FR2	Gtl	Somerst	Norm	

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	\
895	Norm	1Fam	2Story	6	5	1963	
1414	Norm	1Fam	1.5Fin	6	7	1923	
25	Norm	1Fam	1Story	8	5	2007	
0	Norm	1Fam	2Story	7	5	2003	
959	Norm	Twnhs	2Story	7	5	1999	

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	\
895	1963	Hip	CompShg	HdBoard	HdBoard	BrkFace	
1414	2000	Gambrel	CompShg	Wd Sdng	Wd Sdng	NaN	
25	2007	Gable	CompShg	VinylSd	VinylSd	Stone	
0	2003	Gable	CompShg	VinylSd	VinylSd	BrkFace	
959	1999	Hip	CompShg	MetalSd	MetalSd	NaN	

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	\
895	415.0	TA	TA	CBlock	TA	TA	
1414	0.0	TA	TA	BrkTil	TA	TA	
25	640.0	Gd	TA	PConc	Gd	TA	
0	196.0	Gd	TA	PConc	Gd	TA	
959	0.0	Gd	TA	PConc	Gd	TA	

	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	\
895	No	BLQ	400	Unf	0	
1414	No	Unf	0	Unf	0	
25	No	Unf	0	Unf	0	
0	No	GLQ	706	Unf	0	
959	No	ALQ	604	Unf	0	

	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	\
895	380	780	GasA	TA	Y	SBrkr	
1414	833	833	GasA	Gd	Y	SBrkr	
25	1566	1566	GasA	Ex	Y	SBrkr	
0	150	856	GasA	Ex	Y	SBrkr	
959	92	696	GasA	Ex	Y	SBrkr	

	1stFlrSF	2ndFlrSF	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	\
895	983	813	0	1796	1	0	
1414	1053	795	0	1848	0	0	
25	1600	0	0	1600	0	0	
0	856	854	0	1710	1	0	
959	696	720	0	1416	1	0	

	FullBath	HalfBath	BedroomAbvGr	KitchenAbvGr	KitchenQual	\
895	1	1	4	1	TA	
1414	1	1	4	1	Gd	
25	2	0	3	1	Gd	
0	2	1	3	1	Gd	
959	2	1	3	1	Gd	

	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	\
895	8	Typ	1	TA	Attchd	1963.0	
1414	8	Typ	1	Gd	Detchd	1922.0	
25	7	Typ	1	Gd	Attchd	2007.0	
0	8	Typ	0	NaN	Attchd	2003.0	

959	6	Typ	0	NaN	Detchd	1999.0
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	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive	\
895	RFn	2	483	TA	TA	Y	
1414	Unf	2	370	TA	TA	N	
25	RFn	3	890	TA	TA	Y	
0	RFn	2	548	TA	TA	Y	
959	Unf	2	484	TA	TA	Y	

	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	\
895	0	50	0	0	0	
1414	0	0	0	0	220	
25	0	56	0	0	0	
0	0	61	0	0	0	
959	0	44	0	0	0	

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	\
895	0	NaN	NaN	NaN	0	10	2008	WD	
1414	0	NaN	NaN	NaN	0	6	2008	WD	
25	0	NaN	NaN	NaN	0	7	2009	WD	
0	0	NaN	NaN	NaN	0	2	2008	WD	
959	0	NaN	NaN	NaN	0	5	2010	WD	

	SaleCondition
895	Normal
1414	Normal
25	Normal
0	Normal
959	Normal

y_train head:

895	140000
1414	207000
25	256300
0	208500
959	155000

Name: SalePrice, dtype: int64

X_val head:

	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	\
1441	120	RM	NaN	4426	Pave	NaN	Reg	
1217	20	FV	72.0	8640	Pave	NaN	Reg	
869	60	RL	80.0	9938	Pave	NaN	Reg	
855	20	RL	NaN	6897	Pave	NaN	IR1	
178	20	RL	63.0	17423	Pave	NaN	IR1	

	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	Condition1	\
1441	Lvl	AllPub	Inside	Gtl	CollgCr	Norm	
1217	Lvl	AllPub	Inside	Gtl	Somerst	Norm	
869	Lvl	AllPub	Inside	Gtl	SawyerW	Norm	

855	Lvl	AllPub	Corner	Gtl	Sawyer	Norm
178	Lvl	AllPub	CulDSac	Gtl	StoneBr	Norm

	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	\
1441	Norm	TwnhsE	1Story	6	5	2004	
1217	Norm	1Fam	1Story	8	5	2009	
869	Norm	1Fam	2Story	7	5	1993	
855	Norm	1Fam	1Story	5	8	1962	
178	Norm	1Fam	1Story	9	5	2008	

	YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	\
1441	2004	Gable	CompShg	VinylSd	VinylSd	BrkFace	
1217	2009	Gable	CompShg	CemntBd	CmentBd	Stone	
869	1994	Gable	CompShg	MetalSd	MetalSd	BrkFace	
855	2010	Gable	CompShg	HdBoard	HdBoard	NaN	
178	2009	Hip	CompShg	VinylSd	VinylSd	Stone	

	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	\
1441	147.0	Gd	TA	PConc	Gd	TA	
1217	72.0	Gd	TA	PConc	Gd	TA	
869	246.0	Gd	TA	PConc	Gd	TA	
855	0.0	TA	Gd	CBlock	TA	TA	
178	748.0	Ex	TA	PConc	Ex	TA	

	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	\
1441	Av	GLQ	697	Unf	0	
1217	Mn	GLQ	936	Unf	0	
869	No	GLQ	750	Unf	0	
855	No	ALQ	659	Unf	0	
178	No	GLQ	1904	Unf	0	

	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	\
1441	151	848	GasA	Ex	Y	SBrkr	
1217	364	1300	GasA	Ex	Y	SBrkr	
869	300	1050	GasA	Ex	Y	SBrkr	
855	381	1040	GasA	Ex	Y	SBrkr	
178	312	2216	GasA	Ex	Y	SBrkr	

	1stFlrSF	2ndFlrSF	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	\
1441	848	0	0	848	1	0	
1217	1314	0	0	1314	1	0	
869	1062	887	0	1949	1	0	
855	1040	0	0	1040	1	0	
178	2234	0	0	2234	1	0	

	FullBath	HalfBath	BedroomAbvGr	KitchenAbvGr	KitchenQual	\
1441	1	0	1	1	Gd	
1217	2	0	3	1	Gd	

869	2	1	3	1	Gd
855	1	1	3	1	TA
178	2	0	1	1	Ex

	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	\
1441	3	Typ	1	TA	Attchd	2004.0	
1217	6	Typ	0	NaN	Attchd	2009.0	
869	8	Typ	1	TA	Attchd	1993.0	
855	6	Typ	0	NaN	Detchd	1962.0	
178	9	Typ	1	Gd	Attchd	2009.0	

	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond	PavedDrive	\
1441	RFn	2	420	TA	TA	Y	
1217	RFn	2	552	TA	TA	Y	
869	Fin	2	574	TA	TA	Y	
855	Unf	1	260	TA	TA	Y	
178	Fin	3	1166	TA	TA	Y	

	WoodDeckSF	OpenPorchSF	EnclosedPorch	3SsnPorch	ScreenPorch	\
1441	149	0	0	0	0	
1217	135	112	0	0	0	
869	156	90	0	0	0	
855	0	104	0	0	0	
178	0	60	0	0	0	

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType	\
1441	0	NaN	NaN	NaN	0	5	2008	WD	
1217	0	NaN	NaN	NaN	0	9	2009	New	
869	0	NaN	GdPrv	NaN	0	6	2010	WD	
855	0	NaN	NaN	NaN	0	4	2010	WD	
178	0	NaN	NaN	NaN	0	7	2009	New	

	SaleCondition
1441	Normal
1217	Partial
869	Normal
855	Normal
178	Partial

y_val head:

1441	149300
1217	229456
869	236000
855	127000
178	501837

Name: SalePrice, dtype: int64

Checking reproducibility:

If X_test is the same: True
If y_test is the same: True
All train/validation fold indices are same.