df = pd.read_csv('ames_clean.csv') df Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilities ... PoolArea PoolQC Fence MiscFeature MiscVal MoSold Yrs Out[4]: AllPub ... 0 1 60 RL65.0 8450 Pave NaN NaN NaN NaN 0 2 2 Reg 1 2 RL 80.0 9600 NaN AllPub ... NaN 0 20 Pave Reg 0 NaN NaN 5 2 3 AllPub ... 0 60 RL68.0 11250 Pave NaN IR1 Lvl 0 NaN NaN 9 NaN 2 3 70 RL60.0 NaN IR1 AllPub ... 0 9550 Pave NaN NaN NaN 4 5 60 RL84.0 14260 Pave NaN IR1 Lvl AllPub ... 0 NaN NaN NaN 0 12 2 • • • ••• ••• • • • ••• • • • ••• • • • ••• 62.0 0 **1455** 1456 60 RL7917 AllPub ... 8 2 Pave NaN Reg 0 NaN NaN NaN **1456** 1457 20 RL 0 2 85.0 13175 Pave NaN AllPub ... 0 NaN MnPrv NaN Reg Lvl **1457** 1458 70 RL66.0 9042 Pave NaN AllPub ... 0 NaN GdPrv Shed 2500 5 2 Reg RL 0 **1458** 1459 20 68.0 9717 Pave AllPub ... NaN 4 2 NaN Reg Lvl 0 NaN NaN **1459** 1460 6 20 RL75.0 9937 AllPub ... NaN 0 2 Pave NaN Lvl 0 NaN NaN Reg 1460 rows × 81 columns

In [12]: target = 'SalePrice' df[target].value_counts()

In [4]: **import** pandas **as** pd

import numpy as np

import seaborn as sns

import matplotlib.pyplot as plt

Out[12]: SalePrice 140000 20 135000 17 155000 14 145000 14 190000 13 202665 1 164900 208300 1 181500 1 147500

Name: count, Length: 663, dtype: int64

In [16]: features = df.drop(columns='SalePrice')

features

Out[16]: Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilities ... ScreenPorch PoolArea PoolQC Fence MiscFeature MiscVal 1 60 RLNaN AllPub ... 0 0 65.0 8450 Pave Lvl 0 NaN NaN NaN C Reg 2 AllPub ... 1 20 RL 0.08 9600 Pave NaN Reg Lvl 0 0 NaN NaN NaN C 3 RL 2 AllPub ... 0 60 68.0 11250 NaN IR1 Lvl 0 NaN C Pave NaN NaN AllPub ... 3 4 RL 9550 IR1 0 0 C 70 60.0 Pave NaN NaN NaN NaN RL4 5 60 84.0 14260 NaN IR1 Lvl AllPub ... 0 0 NaN NaN C Pave NaN ••• ••• ••• ••• RL62.0 0 0 C **1455** 1456 60 7917 Pave NaN AllPub ... NaN Reg Lvl NaN NaN **1456** 1457 20 RL 13175 AllPub ... 0 C 85.0 Pave NaN Lvl 0 NaN MnPrv NaN Reg **1457** 1458 70 RL9042 NaN AllPub ... 0 0 NaN GdPrv 66.0 Pave Reg Lvl Shed 250C **1458** 1459 AllPub ... 0 20 RL68.0 9717 Pave NaN Reg Lvl 0 NaN NaN NaN C

Reg

AllPub ...

Lvl

0

0

NaN

NaN

C

NaN

1460 rows × 80 columns

1459 1460

categorical = features.drop(columns=numeric).columns.values

print(f'''

RL

numeric = features.select_dtypes(include=np.number).columns.values

There are {features.shape[0]} observations and {features.shape[1]} features. {numeric.size} = Numeric Features Count

{categorical.size} = Categorical Features Count

75.0

9937

NaN

Pave

Numeric: {', '.join(numeric)}.

20

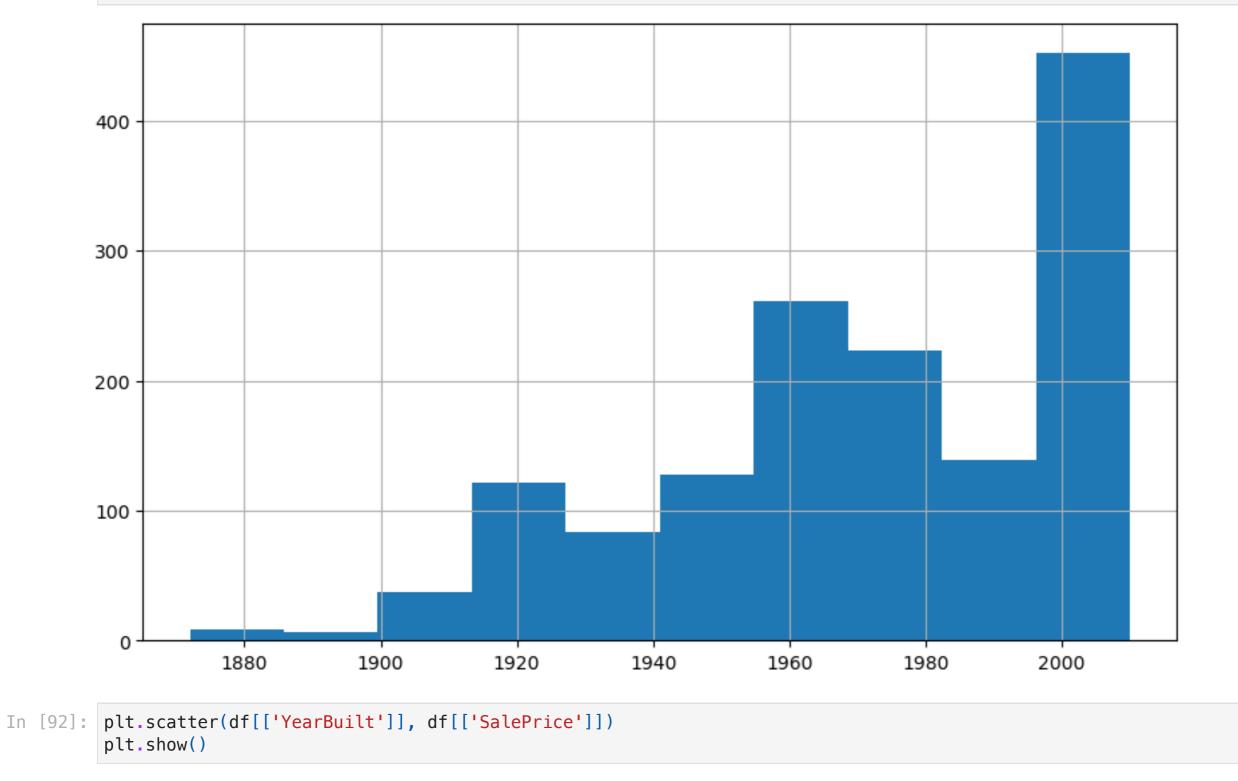
Categorical: {', '.join(categorical)}.

There are 1460 observations and 80 features. 37 = Numeric Features Count

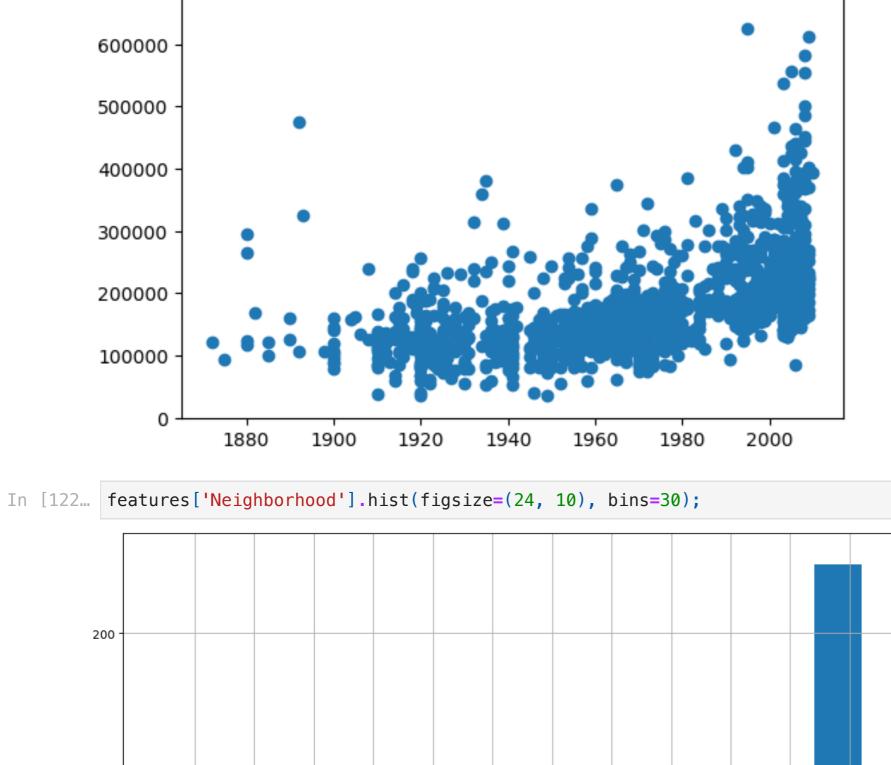
43 = Categorical Features Count Numeric: Id, MSSubClass, LotFrontage, LotArea, OverallQual, OverallCond, YearBuilt, YearRemodAdd, MasVnrArea, BsmtFinSF1, BsmtFinSF2, BsmtUnfSF, TotalBsmt

SF, 1stFlrSF, 2ndFlrSF, LowQualFinSF, GrLivArea, BsmtFullBath, BsmtHalfBath, FullBath, HalfBath, BedroomAbvGr, KitchenAbvGr, TotRmsAbvGrd, Fireplaces, Gar ageYrBlt, GarageCars, GarageArea, WoodDeckSF, OpenPorchSF, EnclosedPorch, 3SsnPorch, ScreenPorch, PoolArea, MiscVal, MoSold, YrSold. Categorical: MSZoning, Street, Alley, LotShape, LandContour, Utilities, LotConfig, LandSlope, Neighborhood, Condition1, Condition2, BldgType, HouseStyle,

RoofStyle, RoofMatl, Exterior1st, Exterior2nd, MasVnrType, ExterQual, ExterCond, Foundation, BsmtQual, BsmtCond, BsmtExposure, BsmtFinType1, BsmtFinType2, Heating, HeatingQC, CentralAir, Electrical, KitchenQual, Functional, FireplaceQu, GarageType, GarageFinish, GarageQual, GarageCond, PavedDrive, PoolQC, Fe nce, MiscFeature, SaleType, SaleCondition. In [90]: features['YearBuilt'].hist(figsize=(10, 6));

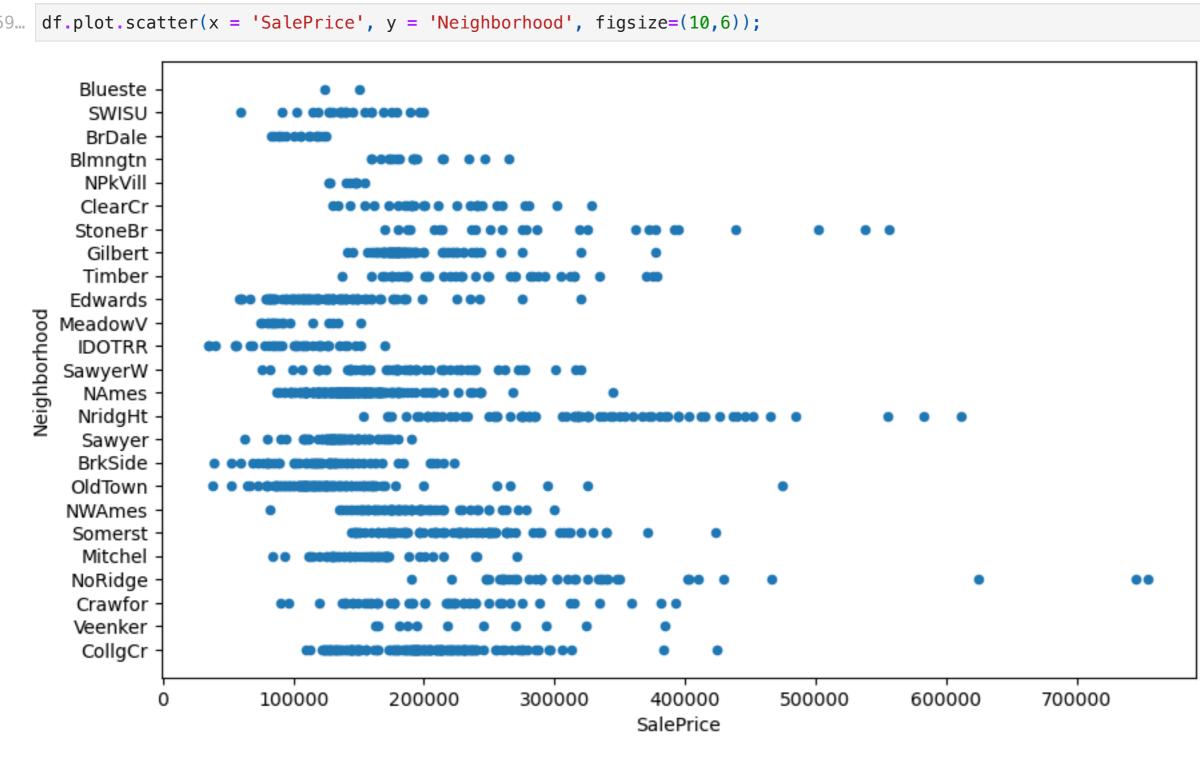


700000



CollgCr Veenker Crawfor NoRidge Mitchel Somerst NWAmes OldTown BrkSide Sawyer NridgHt NAmes SawyerW IDOTRR MeadowV Edwards Timber Gilbert StoneBr ClearCr NPkVill Blmngtn BrDale

SWISU BrDale



150