In [103]: import pandas as pd from sklearn import preprocessing df=pd.read csv(r"C:\Users\91809\Desktop\train.csv") In [104]: df.head() Out[104]: Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilities ... PoolArea PoolQC Fence 0 1 60 RL 65.0 8450 Pave NaN Reg AllPub 0 NaN NaN 1 2 20 RL 80.0 9600 AllPub 0 NaN Pave NaN Reg NaN Lvl 2 3 60 RL 68.0 11250 Pave NaN IR1 Lvl AllPub 0 NaN NaN 3 70 RL 60.0 9550 IR1 AllPub 0 NaN Pave NaN Lvl NaN 5 60 RL 84.0 14260 Pave NaN IR1 Lvl AllPub 0 NaN NaN 5 rows × 81 columns In [105]: df=df.dropna(axis=1) In [106]: df.head() Out[106]: MSSubClass MSZoning LotArea Street LotShape LandContour Utilities LotConfig LandSlope ... EnclosedPorch 3SsnPorch 0 60 RL AllPub Gtl ... 0 0 - 1 8450 Pave Reg Lvl Inside 2 20 RL 9600 AllPub FR2 0 0 1 Pave Reg Lvl Gtl ... 0 0 2 3 60 RL 11250 IR1 AllPub Pave Lvl Inside Gtl 3 4 70 RL 9550 IR1 AllPub Gtl ... 272 0 Pave Lvl Corner 60 IR1 AllPub FR2 Gtl ... 0 0 5 RL 14260 Pave Lvl 5 rows × 62 columns In [107]: from sklearn.preprocessing import LabelEncoder In [108]: number=LabelEncoder() df.keys() df['LandContour'] = number.fit_transform(df['LandContour'].astype('str')) In [109]: df.head() Out[109]: MSZoning **MSSubClass** Utilities LotConfig LandSlope ld LotArea Street LotShape LandContour EnclosedPorch 3SsnPorch 0 1 60 RL 8450 Pave Reg **AllPub** Inside Gtl 0 0 2 RL FR2 0 1 20 9600 Pave Reg 3 AllPub Gtl ... 0 2 3 60 RL 11250 Pave IR1 3 **AllPub** Inside Gtl ... 0 3 IR1 272 0 4 70 RL 9550 Pave 3 AllPub Corner Gtl ... 0 5 60 RL 14260 Pave IR1 **AllPub** FR2 Gtl ... 5 rows × 62 columns In [110]: df['Utilities'] = number.fit_transform(df['Utilities'].astype('str')) df['LotConfig'] = number.fit_transform(df['LotConfig'].astype('str')) df['LandSlope'] = number.fit_transform(df['LandSlope'].astype('str')) df['Neighborhood'] = number.fit transform(df['Neighborhood'].astype('str')) df['Condition1']=number.fit_transform(df['Condition1'].astype('str')) df['Condition2']=number.fit_transform(df['Condition2'].astype('str')) df['SaleType'] = number.fit transform(df['SaleType'].astype('str')) df['SaleCondition']=number.fit_transform(df['SaleCondition'].astype('str')) df['Street'] = number.fit_transform(df['Street'].astype('str')) In [111]: df.head() Out[111]: Id MSSubClass MSZoning LotArea Street LotShape LandContour Utilities LotConfig LandSlope ... EnclosedPorch 3SsnPorch 0 ... 0 1 60 RL 8450 Reg 3 0 4 0 1 2 20 RL 3 0 2 0 ... 0 0 9600 1 Reg 2 3 60 RL 11250 IR1 3 0 4 0 ... 0 0 3 4 70 RL IR1 3 0 0 0 ... 272 0 9550 1 0 ... 5 60 RL 14260 IR1 3 0 2 0 0 5 rows × 62 columns In [112]: df.shape Out[112]: (1460, 62) In [113]: df.dtypes Out[113]: Id int.64 MSSubClass int64 MSZoning object LotArea int64 Street int32 . . . MoSold int64 YrSold int64 int32 SaleType SaleCondition int32 SalePrice Length: 62, dtype: object In [114]: x=df.columnsIn [115]: x Out[115]: Index(['Id', 'MSSubClass', 'MSZoning', 'LotArea', 'Street', 'LotShape', 'LandContour', 'Utilities', 'LotConfig', 'LandSlope', 'Neighborhood', 'Condition1', 'Condition2', 'BldgType', 'HouseStyle', 'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd', 'RoofStyle', 'RoofMatl', 'Exterior1st', 'Exterior2nd', 'ExterQual', 'ExterCond', 'Foundation', 'BsmtFinSF1', 'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating', 'HeatingQC', 'CentralAir', '1stFlrSF', '2ndFlrSF', 'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual', 'TotRmsAbvGrd', 'Functional', 'Fireplaces', 'GarageCars', 'GarageArea', 'PavedDrive', 'WoodDeckSF', 'OpenPorchSF', 'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'MiscVal', 'MoSold', 'YrSold', 'SaleType', 'SaleCondition', 'SalePrice'], dtype='object') In [116]: **for** y **in** x: df[y]=number.fit transform(df[y].astype('str')) In [117]: df.head() Out[117]: Id MSSubClass MSZoning LotArea Street LotShape LandContour Utilities LotConfig LandSlope ... EnclosedPorch 3SsnPorch 3 4 0 ... 0 0 0 9 3 859 3 0 0 0 ... **1** 572 4 3 1030 1 3 3 0 2 0 0 **2** 683 9 3 161 0 3 0 4 0 0 **3** 794 10 3 1021 1 0 3 0 0 ... 78 0 4 905 0 ... 0 0 5 rows × 62 columns In [118]: df.dtypes Out[118]: Id int32 MSSubClass int32 int32 MSZoning LotArea int32 Street int32 MoSold int32 YrSold int32 int32 SaleType SaleCondition int32 SalePrice int32 Length: 62, dtype: object In [119]: df.describe() Out[119]: Id MSSubClass LandSlo **MSZoning** LotArea Street LotShape LandContour Utilities LotConfig count 1460.000000 1460.000000 1460.000000 1460.000000 1460.000000 1460.000000 1460.000000 1460.000000 1460.000000 1460.0000 0.000685 729.500000 6.214384 3.028767 565.210959 0.995890 1.942466 2.777397 3.019178 0.0623 mean 421.610009 3.543318 0.632017 314.364525 0.063996 1.409156 0.707666 0.026171 1.622634 0.2762 std 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.0000 min 289.750000 25% 364.750000 4.000000 3.000000 1.000000 0.000000 3.000000 0.000000 2.000000 0.0000 3.000000 50% 729.500000 5.000000 3.000000 603.000000 1.000000 3.000000 0.000000 4.000000 0.0000 9.000000 0.000000 4.000000 0.0000 75% 1094.250000 3.000000 839.000000 1.000000 3.000000 3.000000 1.000000 max 1459.000000 14.000000 4.000000 1072.000000 3.000000 3.000000 1.000000 4.000000 2.0000 8 rows × 62 columns In [120]: from sklearn.linear model import LinearRegression In [121]: from sklearn.model_selection import train test split In []: In [126]: y=df['SalePrice'] x= df.drop('SalePrice', 1) In [127]: | y.head() Out[127]: 0 343 1 270 2 373 3 125 425 Name: SalePrice, dtype: int32 In [128]: y.shape Out[128]: (1460,) In [129]: x.shape Out[129]: (1460, 61) In [130]: df.shape Out[130]: (1460, 62) In [173]: | X_train, X_test, y_train, y_test = train_test_split(x, y, test_size = 0.21) In [174]: regr = LinearRegression() regr.fit(X_train, y_train) print(regr.score(X_test, y_test)) 0.36116234398592284 In [175]: y_pred = regr.predict(X_test) import matplotlib.pyplot as plt In [176]: | plt.scatter(y_test,y_pred) plt.show() 700 600 500 400 300 200 100 100 200 300 400 500 600 In [177]: | accuracy = regr.score(X_test,y_test) In [178]: accuracy Out[178]: 0.36116234398592284 In [179]: regr.coef_ Out[179]: array([-9.49476736e-03, -2.86591257e+00, 3.24391608e+00, -4.13878929e-02, 5.97360825e+01, -5.97875294e+00, -4.35526710e+00, -1.12455749e+02, 2.15020232e+00, 2.43106372e+01, 6.44526756e-01, -7.77278956e+00, -2.50963017e+01, -7.68810149e+00, 3.03380397e+00, 7.93638514e+00, -2.17227068e+01, 3.80232653e-01, -1.50105375e-02, -1.70624621e+00, 1.92062640e+01, -5.51079375e+00, 3.30069499e+00, -3.99952528e+01, -9.08754033e+00, -7.90362063e+00, -8.69064766e-02, -1.22411001e-01, -1.75057497e-02, -5.17127038e-02, 4.20144140e+01, -2.93858261e+00, -1.36501653e+02, -5.01852344e-02, -4.81164576e-02, -5.80277429e+00, 1.33369317e-01, 2.31201531e+01, 2.06756999e+01, 6.29307821e+01, 2.98838514e+01, -1.35043464e+01, -9.23393301e+01, -2.71317193e+01, -6.03065345e-02, 9.67416213e+00, 1.71734451e+01, -1.64718692e+01, 1.03891514e-01, -3.70570760e+01, 9.05944086e-02, 8.93164393e-02, 2.52089309e-01, 2.46344957e+00, -1.52769675e-01, 7.70522200e+00, 4.82552596e+00, 2.85120268e+00, 1.76237111e+00, 4.93719644e-01, -9.38429909e+00]) In [180]: regr.intercept_ Out[180]: 690.1649956406885 In [182]: import numpy as np y pred = regr.predict(X test) plt.plot(y_test, y_pred, '.') # plot a line, a perfit predict would all fall on this line x = np.linspace(0, 700, 100)y = xplt.plot(x, y) plt.show() 700 600 500 400 300 200 100 0 100 300 400 500 600 700 In []: