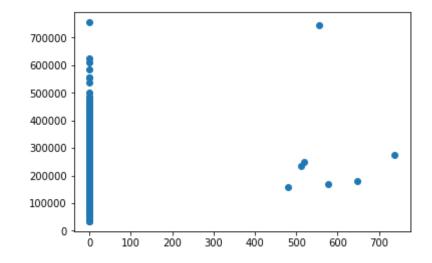
```
In [1]: import pandas as pd
In [12]: import numpy as np
In [46]: df train = pd.read csv('train.csv')
         df test = pd.read csv('test.csv')
In [87]: df train.columns
Out[87]: Index(['Id', 'MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Stree
         t',
                'Alley', 'LotShape', 'LandContour', 'Utilities', 'LotConfig',
                'LandSlope', 'Neighborhood', 'Condition1', 'Condition2', 'BldgTy
         pe',
                'HouseStyle', 'OverallOual', 'OverallCond', 'YearBuilt', 'YearRe
         modAdd'.
                'RoofStyle', 'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrTy
         pe',
                'MasVnrArea', 'ExterQual', 'ExterCond', 'Foundation', 'BsmtQua
         l',
                'BsmtCond', 'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1',
                'BsmtFinType2', 'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heati
         ng',
                'HeatingQC', 'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF',
                'LowQualFinSF', 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'Fu
         llBath',
                'HalfBath', 'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual',
                'TotRmsAbvGrd', 'Functional', 'Fireplaces', 'FireplaceQu', 'Gara
         geType',
                'GarageYrBlt', 'GarageFinish', 'GarageCars', 'GarageArea', 'Gara
         geQual',
                'GarageCond', 'PavedDrive', 'WoodDeckSF', 'OpenPorchSF',
                'EnclosedPorch', '3SsnPorch', 'ScreenPorch', 'PoolArea', 'Pool0
         С',
                'Fence', 'MiscFeature', 'MiscVal', 'MoSold', 'YrSold', 'SaleTyp
```

```
e',
                  'SaleCondition', 'SalePrice'],
                 dtype='object')
In [42]: df_test.head()
Out[42]:
               Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour U
           0 1461
                          20
                                   RH
                                             0.08
                                                    11622
                                                           Pave
                                                                 NaN
                                                                          Reg
                                                                                      Lvl
           1 1462
                                   RL
                          20
                                             81.0
                                                    14267
                                                           Pave
                                                                 NaN
                                                                          IR1
                                                                                      LvI
                                   RL
                                                                                      Lvl
           2 1463
                          60
                                             74.0
                                                    13830
                                                           Pave
                                                                NaN
                                                                          IR1
           3 1464
                          60
                                   RL
                                             78.0
                                                     9978
                                                           Pave
                                                                 NaN
                                                                          IR1
                                                                                      Lvl
           4 1465
                         120
                                                                                     HLS
                                    RL
                                             43.0
                                                     5005
                                                           Pave
                                                                NaN
                                                                          IR1
          5 rows × 80 columns
In [86]:
In [ ]:
          import matplotlib.pyplot as plt
In [19]:
          %matplotlib inline
In [32]: plt.scatter(df train['PoolArea'],df train['SalePrice'])
Out[32]: <matplotlib.collections.PathCollection at 0x1e409c324e0>
```



```
In [139]: x = df train.drop(['Id', 'SalePrice'], axis=1)
          y = df train['SalePrice']
          x.columns
Out[139]: Index(['MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street', 'A
          lley',
                 'LotShape', 'LandContour', 'Utilities', 'LotConfig', 'LandSlop
          e',
                 'Neighborhood', 'Condition1', 'Condition2', 'BldgType', 'HouseSt
          yle',
                 'OverallQual', 'OverallCond', 'YearBuilt', 'YearRemodAdd', 'Roof
          Style',
                 'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType', 'MasVnrA
          rea',
                 'ExterQual', 'ExterCond', 'Foundation', 'BsmtQual', 'BsmtCond',
                 'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1', 'BsmtFinType2',
                 'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating', 'Heating'
          С',
                 'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF', 'LowQualFinS
          F',
                 'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBa
          th',
                 'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual', 'TotRmsAbvGrd',
```

```
'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType', 'Garage
          YrBlt',
                  'GarageFinish', 'GarageCars', 'GarageArea', 'GarageQual', 'Garag
          eCond',
                 'PavedDrive', 'WoodDeckSF', 'OpenPorchSF', 'EnclosedPorch', '3Ss
          nPorch',
                 'ScreenPorch', 'PoolArea', 'PoolOC', 'Fence', 'MiscFeature', 'Mi
          scVal',
                 'MoSold', 'YrSold', 'SaleType', 'SaleCondition'],
                dtvpe='object')
In [140]: from sklearn.preprocessing import LabelEncoder
          le = LabelEncoder()
          c = ['MSSubClass', 'MSZoning', 'LotFrontage', 'LotArea', 'Street', 'All
          ey',
                  'LotShape', 'LandContour', 'Utilities', 'LotConfig', 'LandSlope'
                  'Neighborhood', 'Condition1', 'Condition2', 'BldaType', 'HouseSt
          yle',
                  'OverallOual', 'OverallCond', 'YearBuilt', 'YearRemodAdd', 'Roof
          Style',
                  'RoofMatl', 'Exterior1st', 'Exterior2nd', 'MasVnrType', 'MasVnrA
          rea',
                  'ExterOual', 'ExterCond', 'Foundation', 'BsmtOual', 'BsmtCond',
                  'BsmtExposure', 'BsmtFinType1', 'BsmtFinSF1', 'BsmtFinType2',
                  'BsmtFinSF2', 'BsmtUnfSF', 'TotalBsmtSF', 'Heating', 'HeatingQC'
                  'CentralAir', 'Electrical', '1stFlrSF', '2ndFlrSF', 'LowQualFinS
          F',
                  'GrLivArea', 'BsmtFullBath', 'BsmtHalfBath', 'FullBath', 'HalfBa
          th',
                  'BedroomAbvGr', 'KitchenAbvGr', 'KitchenQual', 'TotRmsAbvGrd',
                  'Functional', 'Fireplaces', 'FireplaceQu', 'GarageType', 'Garage
          YrBlt',
                  'GarageFinish', 'GarageCars', 'GarageArea', 'GarageOual', 'Garag
          eCond',
                  'PavedDrive', 'WoodDeckSF', 'OpenPorchSF', 'EnclosedPorch', '3Ss
          nPorch',
                  'ScreenPorch', 'PoolArea', 'PoolOC', 'Fence', 'MiscFeature', 'Mi
```

```
scVal',
                   'MoSold', 'YrSold', 'SaleType', 'SaleCondition']
           for i in c:
               s = le.fit_transform(x[i].astype(str))
               x[i] = s
           x.head()
           #from sklearn.preprocessing import OneHotEncoder
           #neHotEncoder().fit transform(x)
Out[140]:
              MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilities
              MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilities
            0
                       9
                                3
                                          75
                                                 859
                                                        1
                                                              2
                                                                       3
                                                                                          0
                                3
                                                              2
                                                                       3
                                                                                  3
            1
                       4
                                          90
                                                1030
                                                        1
                                                                                          0
                                                              2
                                          78
                                                 161
                                                                                          0
                                                        1
            3
                      10
                                3
                                                              2
                                                                      0
                                                                                  3
                                                                                          0
                                          70
                                                1021
                       9
                                                 386
                                                        1
                                                              2
                                                                                  3
                                                                                          0
           5 rows × 79 columns
In [141]: from sklearn.model selection import train test split
           x train,x test,y train,y test = train test split(x,y,test size=0.2, ran
           dom state=42)
In [143]: x train.head()
Out[143]:
                 MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilit
             254
                         4
                                   3
                                            80
                                                   852
                                                                2
                                                                         3
                                                                                     3
                                   3
                                                                         0
            1066
                         9
                                            69
                                                   789
                                                                2
                                                                                     3
```

```
MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape LandContour Utilit
                         5
                                           77
                                                  906
                                  3
                                                              2
                                                                       3
            638
                                                                                  3
            799
                         8
                                  3
                                           70
                                                  734
                                                              2
                                                                       3
                                                  620
                                                                       3
            380
                         8
                                  3
                                           60
           5 rows × 79 columns
  In [ ]:
In [144]: from sklearn.linear model import LinearRegression
In [145]: clf = LinearRegression()
In [147]: clf.fit(x_train, y_train)
Out[147]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None,
                    normalize=False)
In [148]: clf.score(x test,y test)
Out[148]: 0.7857307099868154
  In [ ]:
```