```
In [1]:  M import pandas as pd
import ipykernel
    from pathlib import Path
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.backends.backend_pdf import PdfPages
from matplotlib.ticker import MultipleLocator
from matplotlib.ticker import MaxNLocator
from scipy import stats

from sklearn.linear_model import LinearRegression
```

In [2]: # "mock-data" file generated from xls spreadsheets of actual data to be used
df = pd.read_csv('rats_Garb_LU_ED_INC_POP_3.csv')
df.head()

Out[2]:

	CD	total_rats	tons_of_refuge	tons_of_MGP	tons_of_paper	tons_res_organics	tons_s
0	01 BRONX	2338	764007.8	32663.3	28960.3	141.5	
1	02 BRONX	2027	582803.0	34424.3	24529.8	NaN	
2	03 BRONX	2522	723072.8	34086.5	28189.1	NaN	
3	04 BRONX	4158	1598950.4	78164.7	61554.6	343.0	
4	05 BRONX	3365	1351276.7	88723.5	58439.9	21.3	

5 rows × 28 columns



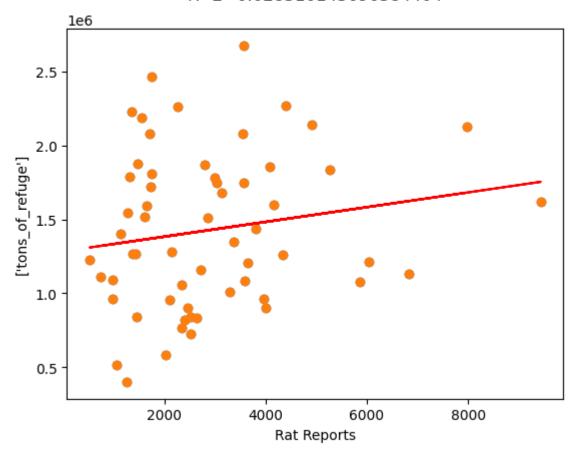
```
In [3]:
        #getting stats for rats
            data = df.total_rats
            # removing null values to avoid errors
            data.dropna(inplace = True)
            # percentile list
            perc =[.25, .40, .60, .75]
            # list of dtypes to include
            include =['object', 'float', 'int']
            # calling describe method
            desc = data.describe(percentiles = perc, include = include)
            # display
            desc
   Out[3]: count
                       59.000000
                     2897.949153
            mean
            std
                     1775.622945
                      522.000000
            min
            25%
                     1572.000000
            40%
                     2160.400000
            50%
                     2517.000000
            60%
                     2970.400000
                     3625.500000
            75%
            max
                     9437.000000
            Name: total rats, dtype: float64
        # columnList=['tons_of_MGP', 'tons_of_paper', 'tons_res_organics', 'tons_sch_
In [4]:
                           'tons_xmastrees', 'pct_LU1', 'pct_LU2', 'pct_LU3', 'pct_LU4',
                           'pct_LU8', 'pct_LU9', 'pct_LU10', 'pct_LU11','BDorHigher', 'Ass
            #
                          'LessHS', 'Adults25nOlder', 'Under18', 'eighteentoTwo']
            columnList=['tons_of_refuge', 'tons_of_MGP', 'tons_of_paper',
                        'tons_xmastrees', 'pct_LU1', 'pct_LU2', 'pct_LU3', 'pct_LU4', 'pc
                        'pct_LU8', 'pct_LU9', 'pct_LU10', 'pct_LU11','BDorHigher', 'Assoc
                         'LessHS', 'Adults25nOlder', 'Under18', '18to24']
```



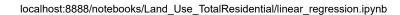
```
In [6]:
         cdf=df
                cdf.dropna(axis='columns')
                # The data in the df column must be reshaped into an array with shape (nu
                # https://stackoverflow.com/questions/18691084/what-does-1-mean-in-numpy-
                plt.scatter(cdf.total_rats, cdf[i])
                  plt.xlabel('Rat Reports')
                  plt.ylabel([i])
                #plt.show()
                X = cdf.total_rats.values.reshape(-1, 1)
                \# The shape of X is no samples, with a single feature (column)
                X.shape
                y = cdf[i]
                cdf[i].describe()
                # Create a model with scikit-learn
                model = LinearRegression()
                # Fit the data into the model
                # By convention, X is capitalized and y is lowercase
                model.fit(X, y)
                # The model creates predicted y values based on X values
                y pred = model.predict(X)
                #calculate R-squared of regression model
                r_squared = model.score(X, y)
                # Plot the results. The best fit line is red.
                plt.scatter(X, y)
                plt.title(f'Y = {round(model.coef_[0], 3)} * x + {round(model.intercept_,
                plt.xlabel('Rat Reports')
                plt.ylabel([i])
                plt.plot(X, y_pred, color='red')
                plt.show()
                # The slope
                # The y-intercept
                print((model.coef ))
                print(model.intercept )
                print(r_squared)
                filename = [i]
                plt.tight layout()
                #plt.savefig(filename, format='png')
```



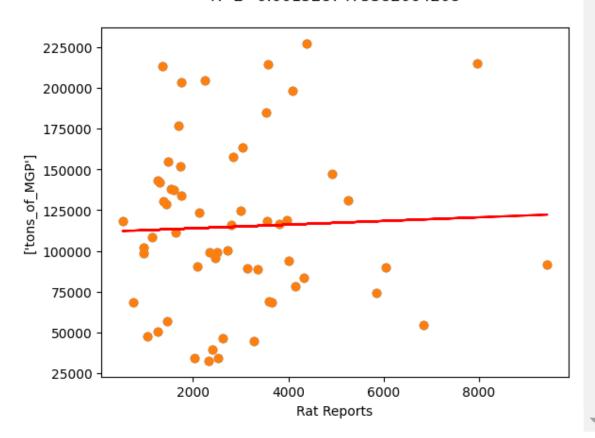
Y = 49.794 * x + 1283364.2 R^2=0.028310143696384404



[49.79434272] 1283364.2284225286 0.028310143696384404



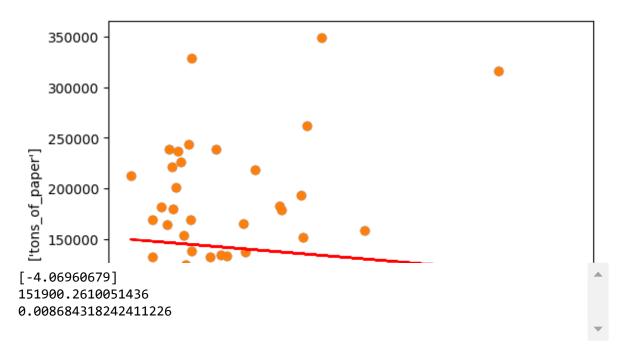
Y = 1.126 * x + 111602.9 R^2=0.0015287475382064208



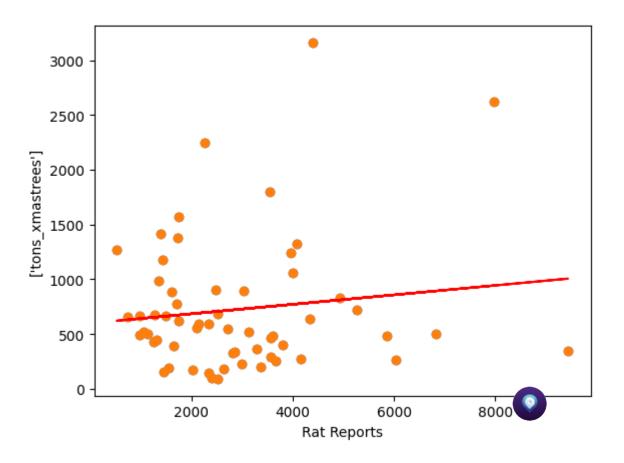
[1.12634825] 111602.86274860674 0.0015287475382064208



Y = -4.07 * x + 151900.3R^2=0.008684318242411226

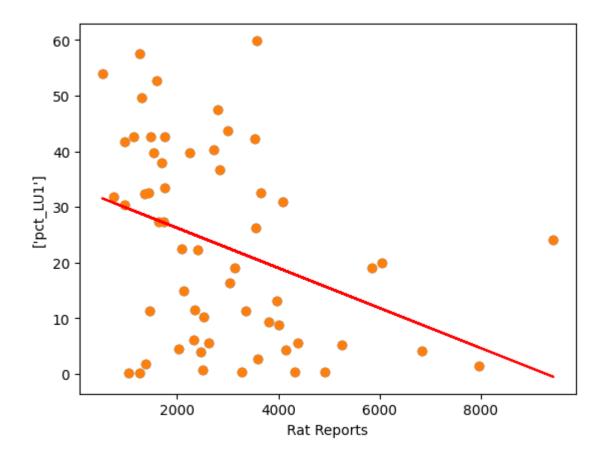


Y = 0.043 * x + 599.1R^2=0.016009384915131486



[0.04313309] 599.0702826728844 0.016009384915131486

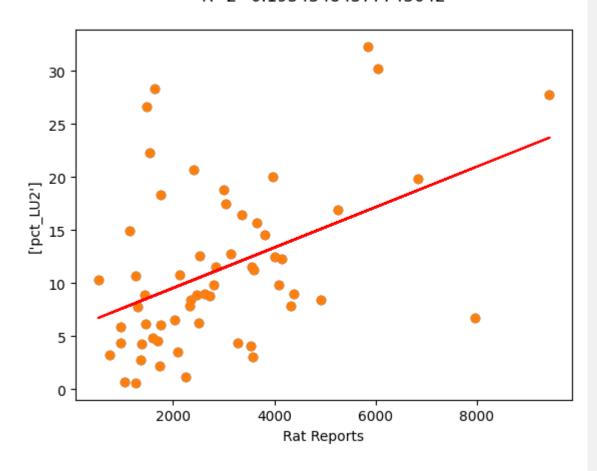
Y = -0.004 * x + 33.4R^2=0.1307572364800612





[-0.00360162] 33.42374069608063 0.1307572364800612

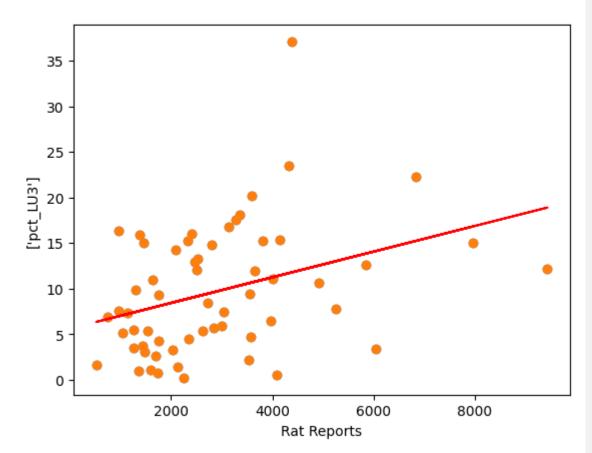
Y = 0.002 * x + 5.7 R^2=0.19543484577743642



[0.00190677]

5.720035065914443

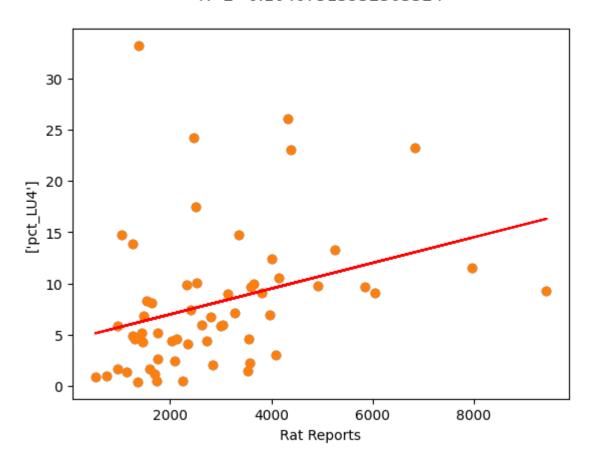
Y = 0.001 * x + 5.6R^2=0.12784596440689933





[0.00140806] 5.599177914932682 0.12784596440689933

Y = 0.001 * x + 4.5R^2=0.10407515932303324

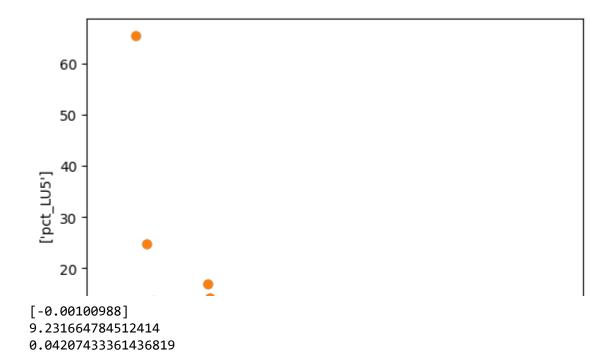


[0.00125325]

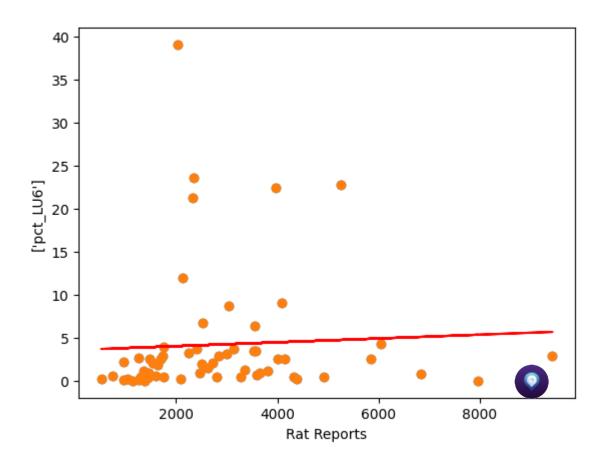
4.4766254327441315



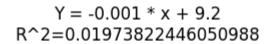
Y = -0.001 * x + 9.2R^2=0.04207433361436819

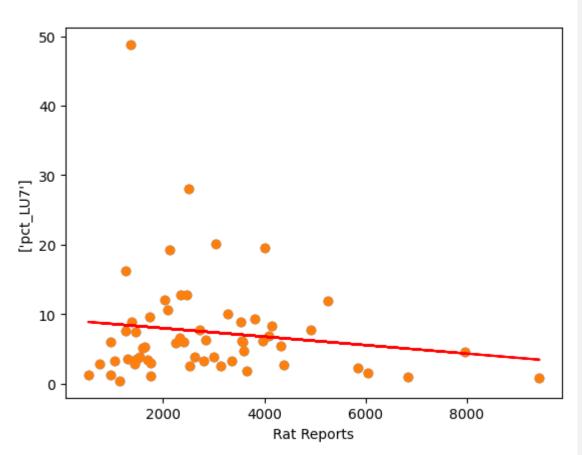


Y = 0.0 * x + 3.6 R^2=0.0029231159111855476



[0.00022246] 3.59092873347047 0.0029231159111855476



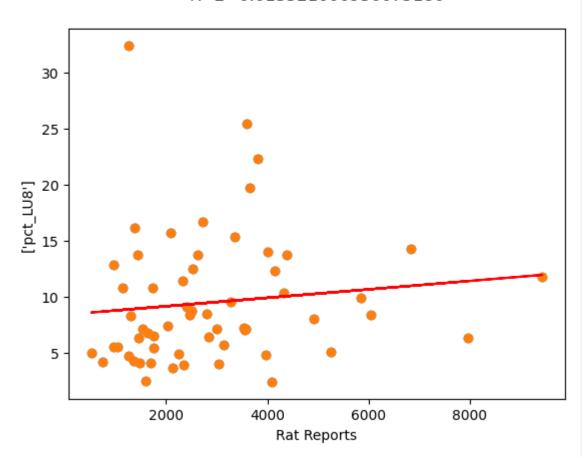




[-0.00060975]

- 9.200928299339719
- 0.01973822446050988

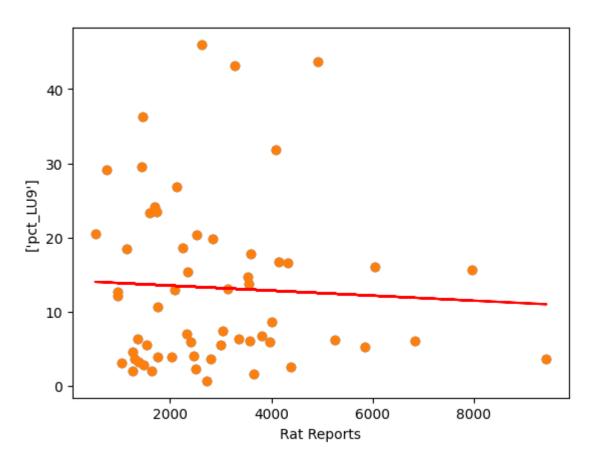
Y = 0.0 * x + 8.4 R^2=0.013521006930075186





[0.00037634] 8.39073217194562 0.013521006930075186

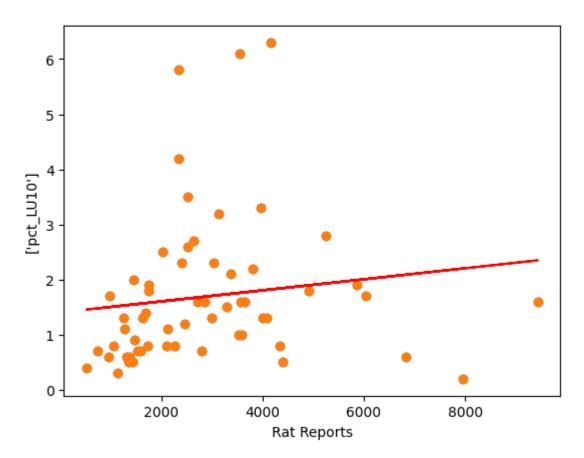
Y = -0.0 * x + 14.2 R^2=0.002822091158191209





[-0.00033876] 14.212203751538457 0.002822091158191209

Y = 0.0 * x + 1.4R^2=0.017757409072682306



[0.00010015]

- 1.4046856247328312
- 0.017757409072682306



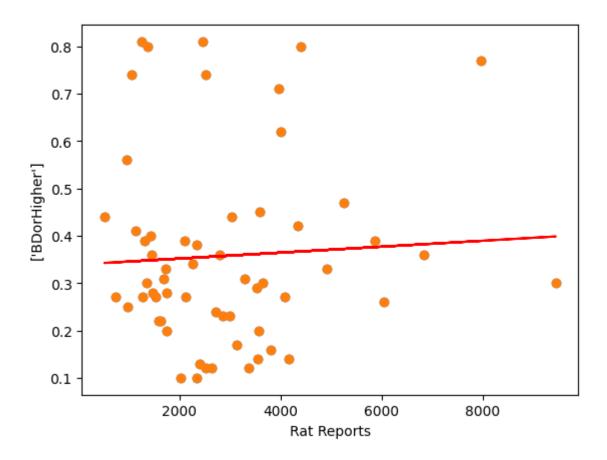
$$Y = 0.0 * x + 4.8$$

R^2=0.005631553889120067



[0.00029175] 4.769775715695724 0.005631553889120067

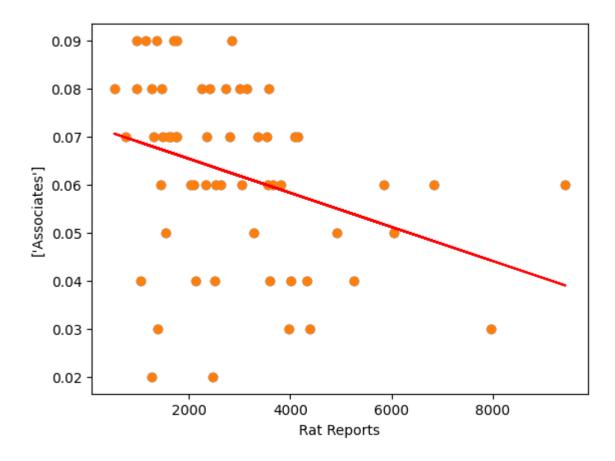
Y = 0.0 * x + 0.3R^2=0.003140210110514796





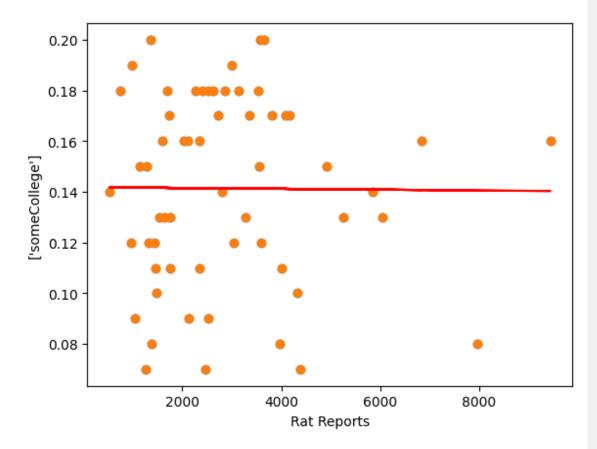
[6.27547797e-06]
0.3392716110473312
0.003140210110514796

Y = -0.0 * x + 0.1R^2=0.11259221404677078



[-3.54526581e-06] 0.07247738987195385

Y = -0.0 * x + 0.1R^2=5.8562156140973265e-05

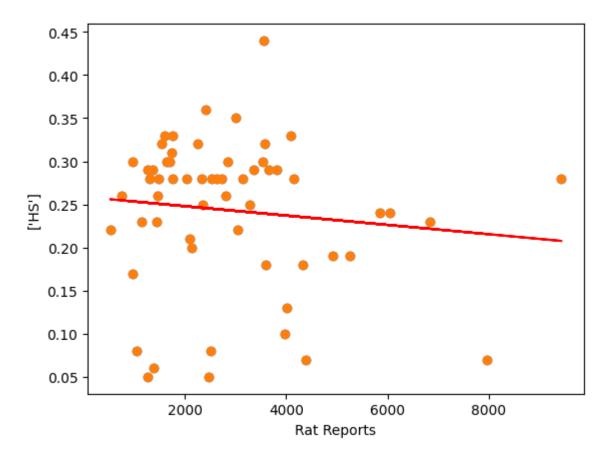


[-1.60752504e-07]

0.14182178478632154

5.8562156140973265e-05

Y = -0.0 * x + 0.3R^2=0.01248818032322574

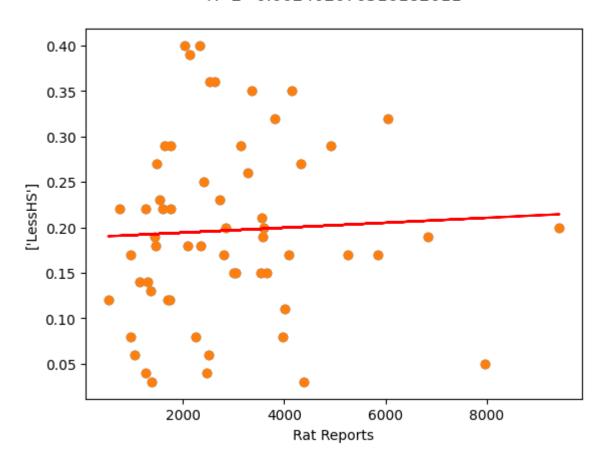


[-5.41805899e-06]

0.2587521069104024



Y = 0.0 * x + 0.2R^2=0.002402076316182611



- [2.71588925e-06]
- 0.1887396605303406
- 0.002402076316182611

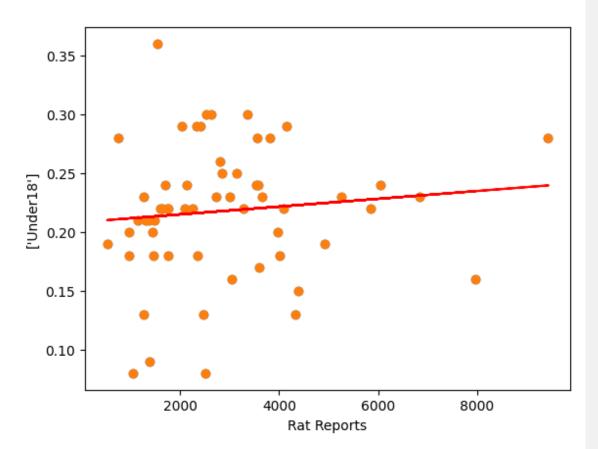


Y = -0.0 * x + 0.7R^2=0.013270998016166824



[-4.31386105e-06] 0.6984335533711107 0.013270998016166824

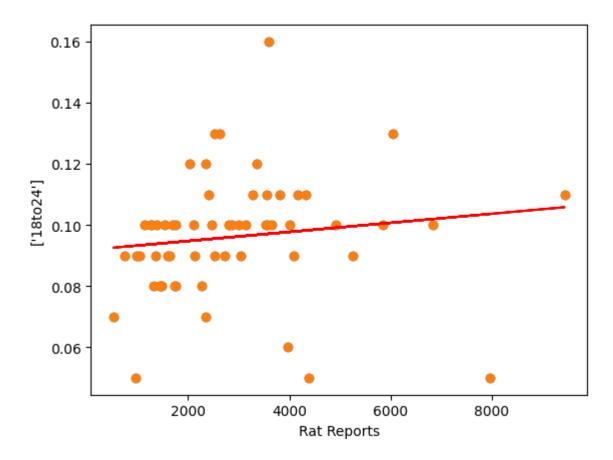
Y = 0.0 * x + 0.2R^2=0.0108176252572596





[3.30777038e-06] 0.2083803513290559 0.0108176252572596

Y = 0.0 * x + 0.1R^2=0.01817996800458832



[1.48573534e-06] 0.09179610944171435 0.01817996800458832

<Figure size 640x480 with 0 Axes>



In []:	M	<pre>#plt.savefig('NYCrats_linear_regression.png')</pre>
In []:	H	
In []:	M	

