# all\_score

## December 10, 2019

```
[66]: import os
     from IPython.display import display, Image
     import pandas as pd
     import numpy as np
     import seaborn as sns
     %matplotlib inline
     import matplotlib.pyplot as plt
     from matplotlib import colors
     from matplotlib.ticker import PercentFormatter
     from scipy.stats import linregress
     import math
     from functools import reduce
     import argparse
     from Bio import SeqIO, Entrez, pairwise2
     Entrez.email = 'hongyingsun11010gmail.com'
     from Bio.SeqRecord import SeqRecord
     import re, time
     import os, sys, glob
     import random
     import uuid
     # from skbio.tree import TreeNode
     # from skbio import read
     # from skbio.stats.distance import DistanceMatrix
     # from skbio.stats.distance import DissimilarityMatrix
     from scipy import stats
     from ast import literal_eval
     import sqlite3
     import statsmodels.api as sm
     from sklearn import linear_model
     from matplotlib import pyplot
 [2]: df =pd.read_csv("pplacer_stats_score_merged.csv", index_col='unique_index')
```

d:\program files (x86)\python37-32\lib\sitepackages\IPython\core\interactiveshell.py:3049: DtypeWarning: Columns
(1,7,8,10,13,19,20,21,24,30,31,32,35,41,42,43) have mixed types. Specify dtype

```
option on import or set low_memory=False.
interactivity=interactivity, compiler=compiler, result=result)
```

```
[3]: df.columns
 [3]: Index(['Unnamed: 0', 'index_x', 'RDP_10398_adcl', 'RDP_5224_adcl',
            'RDP_1017_adcl', 'RDP_92_adcl', 'RDP_12_adcl', 'community_x',
            'sourceSeq_x', 'organism_x', 'ncbi_tax_id_x', 'multiplicity_x',
            'index_y', 'RDP_10398_edpl', 'RDP_5224_edpl', 'RDP_1017_edpl',
            'RDP_92_edpl', 'RDP_12_edpl', 'community_y', 'sourceSeq_y',
            'organism_y', 'ncbi_tax_id_y', 'multiplicity_y', 'index_x.1',
            'RDP_10398_mindistl', 'RDP_5224_mindistl', 'RDP_1017_mindistl',
            'RDP_92_mindistl', 'RDP_12_mindistl', 'community_x.1', 'sourceSeq_x.1',
            'organism_x.1', 'ncbi_tax_id_x.1', 'multiplicity_x.1', 'index_y.1',
            'RDP_10398_prichness', 'RDP_5224_prichness', 'RDP_1017_prichness',
            'RDP_92_prichness', 'RDP_12_prichness', 'community_y.1',
            'sourceSeq_y.1', 'organism_y.1', 'ncbi_tax_id_y.1', 'multiplicity_y.1',
            'RDP_10398_score', 'RDP_5224_score', 'RDP_1017_score', 'RDP_92_score',
            'RDP_12_score'],
           dtype='object')
[18]: df=df.fillna(method='ffill')
 []:
[28]: | df_rdp_10398_CC11CM0=df.loc[df["community_x"]=="CC11CM0",["RDP_10398_adcl",__
      \rightarrow "RDP_10398_edpl", \Box
      → "RDP_10398_mindistl", "RDP_10398_prichness", "RDP_10398_score", "community_x"]]
[34]: df_rdp_10398_CC11CM0.RDP_10398_score.isna()
[34]: unique_index
     CC11CMOSCR00cdec97c058446e83e0ef032e61806d
                                                    True
     CC11CMOSCR35529da454f0497fa16e04841e8e1639
                                                    True
     CC11CMOSCR1083e70ce28e4961b8298356c0d69000
                                                    True
     CC11CMOSCR8898bf89e307445282cd2fb3acb183a0
                                                    True
     CC11CMOSCRfc7f700b82e44da39405162a7e5b8b77
                                                    True
     CC11CMOSCRbb008cff8b87453b9ed228be5d008673
                                                    True
     CC11CMOSCR3e39a5fc61b6420cac1c2dd465292aec
                                                    True
     CC11CMOSCRb15eac580de84cfd8e55224087a7b676
                                                    True
     CC11CMOSCR1e06de32f41c414aaa57f33949f4905c
                                                    True
     CC11CMOSCRaf64d873b07c43e18a82b0be7ef37e3a
                                                    True
     CC11CMOSCRf7daacc404cb4eb9a49ba1de3996a898
                                                    True
                                                    True
     CC11CMOSCRb70978c24c1a4fab8659357b833938e6
     CC11CMOSCRe9f5fb9ad3e5407fb686716731557fb8
                                                    True
     CC11CMOSCRade8cd02ba724e42b455e0ced33aba32
                                                    True
     CC11CMOSCRd526be427f7d40d99ca20d6b9ffc434d
                                                    True
     CC11CMOSCR4fb0d8fe4a4f42d598229f7496ec6c09
                                                    True
     CC11CMOSCR374d88a64290415ba45a37a3e99e9405
                                                    True
     CC11CMOSCR2ae719dea83c49169b6cb67e532d1fbf
                                                    True
```

```
CC11CMOSCRabb7b9b009944a9ba1d1db4c868741af
                                                  True
   CC11CMOSCR5c24420214de48e2bae6693c53d442af
                                                  True
   CC11CMOSCRd32581f0a84a430985ca339a20df58ff
                                                  True
   CC11CMOSCRda69a360140c4f6da522f9245a9f5db3
                                                  True
   CC11CMOSCR295d55115ed64b7b87a6e1d5d0858b11
                                                  True
                                                  True
   CC11CMOSCRfc1acb97fe7e4e9aa7ed3305a7b7b39a
   CC11CMOSCR6d4aac78b0d242779e124ddd2ab22394
                                                  True
   CC11CMOSCR49c33904b91a42b28626c772a112d3ff
                                                  True
   CC11CMOSCR2398a7ab2bf14cd9bf4f493cd92dca7f
                                                  True
   CC11CMOSCR5cedb8a7e1bf473d99e12728f04a173c
                                                  True
   CC11CMOSCR3823cad73fba408b8b4a7cda1eb5e493
                                                  True
   CC11CMOSCR0a33595fd49e47acb027d6c82aa45efd
                                                  True
   CC11CMOSCR5ff6206ff4d94c8f86932306fb9b4535
                                                  True
   CC11CMOSCRd38f3e00f04b4a2ebf6626b1899a38fa
                                                  True
   CC11CMOSCR6793c1fe87f940a28c62ca45b275741f
                                                  True
   CC11CMOSCRcd7237d670bc46b29f6289e56b4b6c9d
                                                  True
   CC11CMOSCR828e2a45138c4a598abd19e72801818c
                                                  True
   CC11CMOSCR1cfc4024617f454881035fb1bfd55cbc
                                                  True
   CC11CMOSCRf4609d1b137d4abfa11af2713e075214
                                                  True
   CC11CMOSCR06770c5a7fe2475f8059ee919d70eeb5
                                                  True
   CC11CMOSCRa805e58fa65d440fbb0d08558326dff6
                                                  True
   CC11CMOSCR38d38820115f4fbba6d14b09a6cba6a2
                                                  True
   CC11CMOSCR8e7653535a7e4378994c3519fe151150
                                                  True
   CC11CMOSCR8b2096c800db4548bed42a6d71bfe922
                                                  True
   CC11CMOSCR7590e26c76214a0091249bb2e74d3ea6
                                                  True
   CC11CMOSCR6f14135fa1ba41f49b480f6973684fb2
                                                  True
   CC11CMOSCR4b00f7bde5c34317a74ff4bd6e5c594f
                                                  True
   CC11CMOSCRec63631bcd1c474083a6d4bc0ba5ce7c
                                                  True
   CC11CMOSCRc168f76390fb4e7c9f4809f8d0100c39
                                                  True
   CC11CMOSCR660eae0fe6f94d0b95e1bff8fc7a6ff5
                                                  True
   CC11CMOSCRf811279ee40c406e9795ebc76e29a3d0
                                                  True
   CC11CMOSCR00a5f19edcba4a52affb746001888dcb
                                                  True
   Name: RDP_10398_score, dtype: bool
[5]: df_rdp_10398=df[["RDP_10398_adcl", "RDP_10398_edpl", |
    → "RDP_10398_mindistl", "RDP_10398_prichness", "RDP_10398_score", "community_x"]]
   df_rdp_10398=df_rdp_10398.dropna(how='any')
   x_rdp_10398=df_rdp_10398[["RDP_10398_adcl", "RDP_10398_edpl",__
    →"RDP_10398_mindistl","RDP_10398_prichness"]]
   y rdp 10398=df rdp 10398[["RDP 10398 score"]]
   model_rdp_10398=sm.OLS(y_rdp_10398,x_rdp_10398).fit()
   pridictions_rdp_10398=model_rdp_10398.predict(x_rdp_10398)
```

True

True

True

True

True

CC11CMOSCR4220806246c9445f9abbf3606a5a56a1

CC11CMOSCR5f7af4e63f104f02a310f226bcb3eba4

CC11CMOSCR89ad905da6c34fb087c878aa4d409a2c

CC11CMOSCR1beb97d3e3bb418b9843d9afdc14e7bf

CC11CMOSCR5d0910eba67847e79b3cb42c12631ebc

### model\_rdp\_10398.summary()

d:\program files (x86)\python37-32\lib\sitepackages\statsmodels\regression\linear\_model.py:1657: RuntimeWarning: divide by
zero encountered in double\_scalars
 return self.ess/self.df\_model

[5]: <class 'statsmodels.iolib.summary.Summary'>

### OLS Regression Results

=======================================	========					
Dep. Variable:	RDP_10398	_score	R-sq	uared:		0.000
Model:		OLS	Adj.	R-squared:		0.000
Method:	Least So	quares	F-st	atistic:		inf
Date:	Tue, 10 Dec	2019	Prob	(F-statisti	Lc):	nan
Time:	08	:57:40	Log-	Likelihood:		-5.4743e+05
No. Observations:	-	182991	AIC:			1.095e+06
Df Residuals:	-	182990	BIC:			1.095e+06
Df Model:		0				
Covariance Type:	non	robust				
	=======	======				
======						
	coef	std e	err	t	P> t	[0.025
0.975]						
RDP_10398_adcl	4.627e-06	1.97e-	-08	234.673	0.000	4.59e-06
4.67e-06						
RDP_10398_edpl	2.347e-05	1e-	-07	234.673	0.000	2.33e-05
2.37e-05						
RDP_10398_mindistl	4.551e-06	1.94e-	-08	234.673	0.000	4.51e-06
4.59e-06						
RDP_10398_prichness	1.3219	0.0	006	234.673	0.000	1.311
1.333						
					:======	
Omnibus:	21036	66.952	Durb	in-Watson:		0.019
<pre>Prob(Omnibus):</pre>		0.000	Jarq	ue-Bera (JB)	:	13970473.803
Skew:		6.331	Prob	(JB):		0.00
Kurtosis:	4	43.890	Cond	. No.		4.29e+40

#### Warnings:

<sup>[1]</sup> Standard Errors assume that the covariance matrix of the errors is correctly specified.

<sup>[2]</sup> The smallest eigenvalue is 3.98e-76. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

[6]: <class 'statsmodels.iolib.summary.Summary'>

### OLS Regression Results

============		=======		.=======	==========
Dep. Variable: Model: Method: Date: Time: No. Observations: Df Residuals: Df Model: Covariance Type:	Least S Tue, 10 De	OLS Squares	R-squared: Adj. R-squar F-statistic: Prob (F-stat Log-Likeliho AIC: BIC:	cistic):	-0.000 -0.000 -inf nan -6.3882e+05 1.278e+06 1.278e+06
=======================================					
0.975]	coef	std er	r t	P> t	[0.025
RDP_5224_adcl 8.62e-07	8.55e-07	3.71e-0	9 230.286	0.000	8.48e-07
RDP_5224_edpl 1.12e-05	1.113e-05	4.83e-0	8 230.286	0.000	1.1e-05
	8.682e-07	3.77e-0	9 230.286	0.000	8.61e-07
RDP_5224_prichness 0.862	0.8550	0.00		0.000	0.848
Omnibus: Prob(Omnibus): Skew: Kurtosis:	1312	297.941 0.000 3.575 14.605	Durbin-Watso Jarque-Bera Prob(JB): Cond. No.	(JB):	0.019 1416540.181 0.00 6.90e+41

### Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The smallest eigenvalue is 9.6e-78. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

```
[7]: df_rdp_1017=df[["RDP_1017_adcl", "RDP_1017_edpl", |
    →"RDP_1017_mindistl", "RDP_1017_prichness", "RDP_1017_score", "community_x"]]
    df_rdp_1017=df_rdp_1017.dropna(how='any')
    x_rdp_1017=df_rdp_1017[["RDP_1017_adcl", "RDP_1017_edpl", __
    →"RDP_1017_mindistl","RDP_1017_prichness"]]
    y_rdp_1017=df_rdp_1017[["RDP_1017_score"]]
    model_rdp_1017=sm.OLS(y_rdp_1017,x_rdp_1017).fit()
    pridictions_rdp_1017=model_rdp_1017.predict(x_rdp_1017)
   model_rdp_1017.summary()
```

[7]:

		_	ion Results			
Dep. Variable:	RDP_101		R-squared:	=======	0.000	
Model:	OLS		-	d:	0.000	
Method:	Least :	Squares	F-statistic:		inf	
Date:	Tue, 10 Dec 2019 Prob (F-statistic): 08:57:40 Log-Likelihood:		Prob (F-stati	stic):	nan	
Time:			d:	-7.7595e+05		
No. Observations:		182991	AIC:		1.552e+06	
Df Residuals:		182990	BIC:		1.552e+06	
Df Model:		0				
Covariance Type:	noi	nrobust				
======================================	=======	======		=======		
	coef	std er	r t	P> t	[0.025	
0.975]						
RDP_1017_adcl 6.49e-06	6.44e-06	2.62e-0	8 245.943	0.000	6.39e-06	
RDP_1017_edpl 0.000	0.0003	1.16e-0	6 245.943	0.000	0.000	
RDP_1017_mindistl 6.1e-06	6.054e-06	2.46e-0	8 245.943	0.000	6.01e-06	
RDP_1017_prichness	3.2199	0.01	3 245.943	0.000	3.194	
======================================	910	====== 636.426	Durbin-Watson	:	0.019	
Prob(Omnibus):		0.000	Jarque-Bera (	JB):	458124.874	
Skew:		2.510	Prob(JB):		0.00	
Kurtosis:		8.907	Cond. No.		3.76e+41	

### Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The smallest eigenvalue is 1.17e-77. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

[8]: <class 'statsmodels.iolib.summary.Summary'>

11 11 11		-	-		
		OLS Regress	ion Results		
Dep. Variable:	RDI	 P_92_score	R-squared:		-0.000
Model:		OLS	Adj. R-squar	red:	-0.000
Method:	Leas	st Squares	F-statistic:	:	-0.07012
Date:	Tue, 10	Dec 2019	Prob (F-stat	tistic):	1.00
Time:		08:57:40	Log-Likeliho	ood:	-8.2034e+05
No. Observations	:	182991	AIC:		1.641e+06
Df Residuals:		182989	BIC:		1.641e+06
Df Model:		1			
Covariance Type:		nonrobust			
====					
	coef	std err	t	P> t	[0.025
0.975]					
RDP_92_adcl 2.01e+11	-3.367e+10	1.2e+11	-0.281	0.779	-2.68e+11
RDP_92_edpl 3.81e+11	-6.384e+10	2.27e+11	-0.281	0.779	-5.09e+11
RDP_92_mindistl	-1.182e+06	4.21e+06	-0.281	0.779	-9.43e+06
RDP_92_prichness 5.23e+09	6.562e+08	2.33e+09	0.281	0.779	-3.92e+09
Omnibus: Prob(Omnibus): Skew:		19407.154 0.000 0.916	Durbin-Watso Jarque-Bera Prob(JB):		0.019 25955.988 0.00

 Kurtosis:
 2.778
 Cond. No.
 2.04e+42

#### Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The smallest eigenvalue is 1.42e-77. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.

```
[9]: df_rdp_12=df[["RDP_12_adcl", "RDP_12_edpl", \( \)

\[
\times\] "RDP_12_mindistl", "RDP_12_prichness", "RDP_12_score", "community_x"]]

\[
\text{df_rdp_12=df_rdp_12.dropna(how='any')} \]

\[
\times\] \times\]

\[
\times\] "RDP_12_mindistl", "RDP_12_adcl", "RDP_12_edpl", \( \)

\[
\times\] "RDP_12_mindistl", "RDP_12_prichness"]]

\[
\times\] \[
\times\] y_rdp_12=df_rdp_12[["RDP_12_score"]]

\[
\times\] model_rdp_12=sm.OLS(y_rdp_12,x_rdp_12).fit()

\[
\times\] pridictions_rdp_12=model_rdp_12.predict(x_rdp_12)

\[
\times\] model_rdp_12.summary()
```

[9]: <class 'statsmodels.iolib.summary.Summary'>

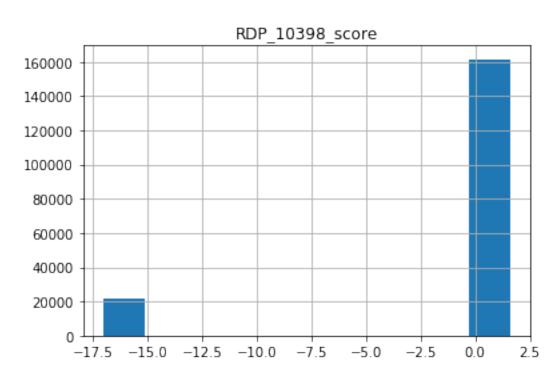
#### OLS Regression Results

		ols negress	TON RESULTS			===
Dep. Variable:	RDP	_12_score OLS	R-squared: Adj. R-squar	ed:	-0. -0.	
Method:	Leas	t Squares	F-statistic:		-19	
Date:		-	Prob (F-stat	istic):		.00
Time:		08:57:40	Log-Likeliho	od:	-8.3694e	+05
No. Observations:	:	182991	AIC:		1.674e	+06
Df Residuals:		182989	BIC:		1.674e	+06
Df Model:		1				
Covariance Type:	:	nonrobust				
0.975]	coef	std err	t	P> t	[0.025	
RDP_12_adcl 2.2e+11	1.109e+11	5.59e+10	1.983	0.047	1.31e+09	
RDP_12_edpl 1.67e+12	8.424e+11	4.25e+11	1.983	0.047	9.95e+09	
RDP_12_mindist1 -2.16e+05	-1.828e+07	9.22e+06	-1.983	0.047	-3.64e+07	
RDP_12_prichness -4.1e+07	-3.47e+09	1.75e+09	-1.983	0.047	-6.9e+09	

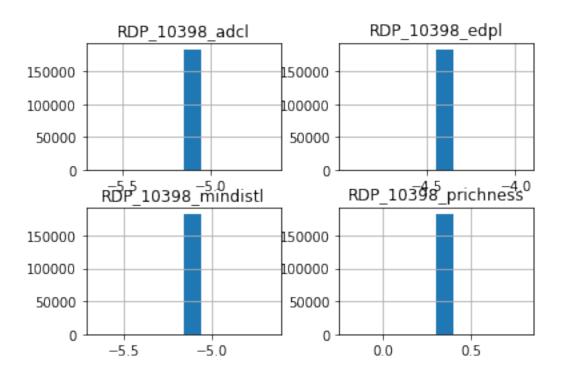
29156.440	Durbin-Watson:	0.020
0.000	Jarque-Bera (JB):	39609.033
1.109	Prob(JB):	0.00
2.479	Cond. No.	3.42e+43
	0.000	0.000 Jarque-Bera (JB): 1.109 Prob(JB):

#### Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The smallest eigenvalue is 2.65e-80. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.



```
[56]:
    y_rdp_10398.describe()
[56]:
            RDP_10398_score
              182991.000000
     count
    mean
                   2.643835
                   4.819320
    std
    min
                   0.00000
    25%
                   2.000000
    50%
                   2.000000
     75%
                   2.000000
                  38.000000
    max
 []:
[51]: x_rdp_10398_log = np.log10(x_rdp_10398+0.00000000000000001)
     x_rdp_10398_log.describe()
[51]:
            RDP_10398_adcl
                             RDP_10398_edpl
                                             RDP_10398_mindistl
                                                                  RDP_10398_prichness
                                                                          182991.00000
     count
              1.829910e+05
                               1.829910e+05
                                                    1.829910e+05
             -5.154902e+00
                              -4.449680e+00
                                                   -5.162034e+00
                                                                               0.30103
    mean
                               1.776362e-15
                                                                               0.00000
     std
              1.776362e-15
                                                    2.664543e-15
             -5.154902e+00
                              -4.449680e+00
                                                   -5.162034e+00
                                                                               0.30103
    min
     25%
             -5.154902e+00
                              -4.449680e+00
                                                   -5.162034e+00
                                                                               0.30103
     50%
             -5.154902e+00
                              -4.449680e+00
                                                   -5.162034e+00
                                                                               0.30103
     75%
             -5.154902e+00
                              -4.449680e+00
                                                   -5.162034e+00
                                                                               0.30103
             -5.154902e+00
                              -4.449680e+00
                                                   -5.162034e+00
    max
                                                                               0.30103
[53]: x_rdp_10398_log.hist()
[53]: array([[<matplotlib.axes.subplots.AxesSubplot object at 0x015F2AD0>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x2BB6FF10>],
            [<matplotlib.axes._subplots.AxesSubplot object at 0x2BBA0FF0>,
             <matplotlib.axes._subplots.AxesSubplot object at 0x2BBBFB30>]],
           dtype=object)
```



[54]: model\_rdp\_10398\_log=sm.OLS(y\_rdp\_10398\_log,x\_rdp\_10398\_log).fit() pridictions\_rdp\_10398\_log=model\_rdp\_10398\_log.predict(x\_rdp\_10398\_log) model\_rdp\_10398\_log.summary()

[54]: <class 'statsmodels.iolib.summary.Summary'>

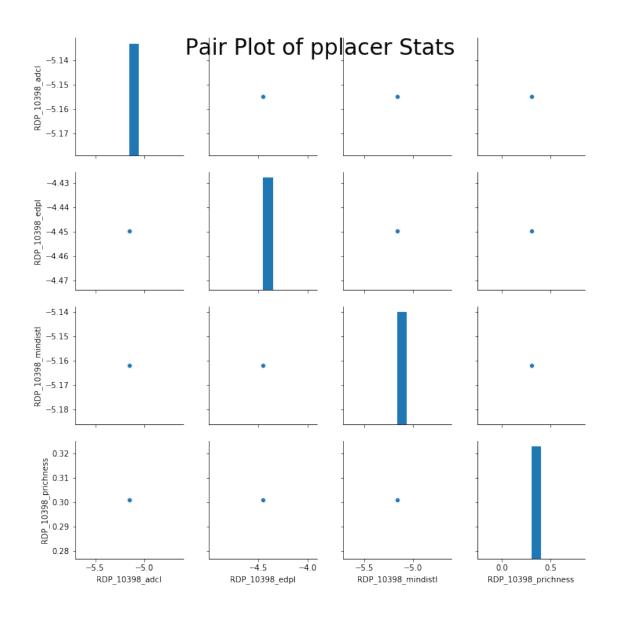
# OLS Regression Results

Dep. Variable:	RDP_10398_	score	R-sqı	uared:		-0.000
Model:		OLS	Adj.	R-squared:		-0.000
Method:	Least So	luares	F-sta	atistic:		-0.5303
Date:	Tue, 10 Dec	2019	Prob	(F-statistic	:):	1.00
Time:	09:	37:59	Log-I	Likelihood:		-5.7433e+05
No. Observations:	1	.82991	AIC:			1.149e+06
Df Residuals:	1	.82989	BIC:			1.149e+06
Df Model:		1				
Covariance Type:	nonr	obust				
======						
	coef	std e	err	t	P> t	[0.025
0.975]						
RDP_10398_adcl	3.275e+08	7.46e	-08	0.439	0.661	-1.13e+09
1.79e+09						

Omnibus: Prob(Omnibus):	806	0.000		n-watson: le-Bera (JB		274672.779
Omnibus:	======== 806	====== 41.233	 Durbi	n-Watson:		0.020
RDP_10398_prichness 2.21e+08	4.043e+07	9.21e	<b>-</b> 07	0.439	0.661	-1.4e+08
RDP_10398_mindistl 2.37e+08	4.346e+07	9.9e	<del>-</del> 07	0.439	0.661	-1.5e+08
RDP_10398_edpl 1.48e+09	-4.271e+08	9.72e	+08	-0.439	0.661	-2.33e+09

# Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The smallest eigenvalue is 1.48e-66. This might indicate that there are strong multicollinearity problems or that the design matrix is singular.



:	RDP_10398_adcl	RDP_10398_edpl	RDP_10398_mindistl	RDP_10398_prichness
count	1.829910e+05	1.829910e+05	1.829910e+05	182991.00000
mean	-5.154902e+00	-4.449680e+00	-5.162034e+00	0.30103
std	1.776362e-15	1.776362e-15	2.664543e-15	0.00000
min	-5.154902e+00	-4.449680e+00	-5.162034e+00	0.30103
25%	-5.154902e+00	-4.449680e+00	-5.162034e+00	0.30103
50%	-5.154902e+00	-4.449680e+00	-5.162034e+00	0.30103
75%	-5.154902e+00	-4.449680e+00	-5.162034e+00	0.30103
max	-5.154902e+00	-4.449680e+00	-5.162034e+00	0.30103