

```
In [ ]: import pandas as pd
df = pd.read_stata("A1_kommune.dta")
print(df)
```

	nr	kommune	taxrev	taxrate	pop
0	101	Københavns Kommune	44170.335938	23.799999	528208
1	147	Frederiksberg Kommune	6682.439941	23.100000	96718
2	151	Ballerup Kommune	4598.704102	25.500000	47652
3	153	Brøndby Kommune	3121.204834	24.500000	33795
4	155	Dragør Kommune	858.017944	24.799999	13564
..
93	840	Rebild Kommune	1957.348511	25.100000	28852
94	846	Mariagerfjord Kommune	3186.054932	25.700001	42604
95	849	Jammerbugt Kommune	3241.256592	25.299999	38927
96	851	Aalborg Kommune	16330.091797	25.400000	197426
97	860	Hjørring Kommune	5061.796875	25.400000	66803

[98 rows x 5 columns]

```
In [ ]: df[['taxrev', 'taxrate', 'pop']].describe()
```

```
Out [ ]:
```

	taxrev	taxrate	pop
count	98.000000	98.000000	98.000000
mean	4477.341309	25.208162	56475.887755
std	5251.175293	0.908003	62925.301713
min	211.228409	22.799999	1969.000000
25%	2466.702271	24.799999	29997.750000
50%	3317.848633	25.299999	43475.000000
75%	4786.060913	25.700001	59733.000000
max	44170.335938	27.799999	528208.000000

```
In [ ]: df['kommune'].value_counts(normalize=True)
```

```
Out [ ]:
```

kommune	
Københavns Kommune	0.010204
Syddjurs Kommune	0.010204
Lemvig Kommune	0.010204
Holstebro Kommune	0.010204
Herning Kommune	0.010204
...	
Køge Kommune	0.010204
Greve Kommune	0.010204
Bornholms Kommune	0.010204
Gribskov Kommune	0.010204
Hjørring Kommune	0.010204

Name: proportion, Length: 98, dtype: float64

```
In [ ]: import numpy as np
import statsmodels.api as sm

y = np.log(df['taxrev'])
X = df['taxrate']
```

```
X = sm.add_constant(X)

OLS = sm.OLS(y, X).fit()
print(OLS.summary())
```

OLS Regression Results

```
=====
===
Dep. Variable:          taxrev    R-squared:
0.029
Model:                  OLS      Adj. R-squared:
0.018
Method:                Least Squares    F-statistic:
2.818
Date:                  Thu, 18 Sep 2025    Prob (F-statistic):          0.
0965
Time:                  14:05:09    Log-Likelihood:          -11
1.12
No. Observations:      98    AIC:          2
26.2
Df Residuals:          96    BIC:          2
31.4
Df Model:              1
Covariance Type:      nonrobust
=====
=====
=====
coef      std err          t      P>|t|      [0.025      0.
975]
-----
-----
const      11.6982      2.143      5.459      0.000      7.444      1
5.952
taxrate     -0.1426      0.085     -1.679      0.096     -0.311
0.026
=====
=====
=====
Omnibus:          17.794    Durbin-Watson:
1.863
Prob(Omnibus):    0.000    Jarque-Bera (JB):          4
5.446
Skew:            -0.567    Prob(JB):          1.35
e-10
Kurtosis:        6.138    Cond. No.
705.
=====
=====
```

Notes:

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

```
In [ ]: df['logtaxrev'] = np.log(df['taxrev'])

y = df['logtaxrev']
df['logpop'] = np.log(df['pop'])

X = sm.add_constant(df[['taxrate', 'logpop']])

OLS = sm.OLS(y, X).fit()
print(OLS.summary())
```

OLS Regression Results

=====						
====						
Dep. Variable:	logtaxrev		R-squared:			
0.980						
Model:	OLS		Adj. R-squared:			
0.980						
Method:	Least Squares		F-statistic:		2	
344.						
Date:	Thu, 18 Sep 2025		Prob (F-statistic):		1.42	
e-81						
Time:	14:05:09		Log-Likelihood:		7	
9.497						
No. Observations:	98		AIC:		-1	
53.0						
Df Residuals:	95		BIC:		-1	
45.2						
Df Model:	2					
Covariance Type:	nonrobust					
=====						
====						
	coef	std err	t	P> t	[0.025	0.
975]						

const	-2.8022	0.376	-7.461	0.000	-3.548	-
2.057						
taxrate	0.0226	0.012	1.816	0.072	-0.002	
0.047						
logpop	0.9711	0.014	67.471	0.000	0.943	
1.000						
=====						
====						
Omnibus:	15.730		Durbin-Watson:			
2.007						
Prob(Omnibus):	0.000		Jarque-Bera (JB):		2	
0.068						
Skew:	0.810		Prob(JB):		4.39	
e-05						
Kurtosis:	4.514		Cond. No.			
933.						
=====						
====						

Notes:

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

```
In [ ]: import pandas as pd

corr = df[["taxrate", "logpop"]].corr()

cov = df[["taxrate", "logpop"]].cov()

print("Korrelation:")
print(corr, "\n")

print("Kovarians:")
print(cov)
```

Korrelation:

	taxrate	logpop
taxrate	1.000000	-0.196649
logpop	-0.196649	1.000000

Kovarians:

	taxrate	logpop
taxrate	0.824469	-0.140289
logpop	-0.140289	0.617290

```
In [ ]: X = sm.add_constant(df['logpop'])
        y = df['taxrate']

        results = sm.OLS(y, X).fit()
        df['res1'] = results.resid

        print(results.summary())
```

OLS Regression Results

```

=====
====
Dep. Variable:          taxrate    R-squared:
0.039
Model:                  OLS        Adj. R-squared:
0.029
Method:                 Least Squares    F-statistic:
3.862
Date:                   Thu, 18 Sep 2025    Prob (F-statistic):      0.
0523
Time:                   14:05:57    Log-Likelihood:        -12
7.16
No. Observations:      98    AIC:                        2
58.3
Df Residuals:          96    BIC:                        2
63.5
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.
975]						
const	27.6268	1.234	22.386	0.000	25.177	3
logpop	-0.2273	0.116	-1.965	0.052	-0.457	

```

=====
====
Omnibus:                3.367    Durbin-Watson:
1.570
Prob(Omnibus):          0.186    Jarque-Bera (JB):
2.699
Skew:                   -0.328    Prob(JB):
0.259
Kurtosis:               3.480    Cond. No.
147.
=====

```

Notes:

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

OLS Regression Results

```

=====
====
Dep. Variable:          logtaxrev    R-squared:
0.001
Model:                  OLS        Adj. R-squared:      -
0.010
Method:                 Least Squares    F-statistic:          0.0
6626
Date:                   Thu, 18 Sep 2025    Prob (F-statistic):
0.797
Time:                   14:05:57    Log-Likelihood:        -11
2.50
No. Observations:      98    AIC:                        2
29.0

```

```

Df Residuals:          96    BIC:          2
34.2
Df Model:              1
Covariance Type:      nonrobust
=====
=====
              coef    std err          t      P>|t|      [0.025      0.
975]
-----
const         8.1031     0.078    104.100     0.000     7.949
8.258
res1          0.0226     0.088     0.257     0.797    -0.152
0.197
=====
=====
Omnibus:          15.581  Durbin-Watson:
1.763
Prob(Omnibus):    0.000  Jarque-Bera (JB):      4
6.400
Skew:            -0.383  Prob(JB):          8.40
e-11
Kurtosis:        6.283  Cond. No.
1.13
=====
=====

```

Notes:

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

```

In [ ]: y1 = df['logtaxrev']
        X1 = sm.add_constant(df['res1'])

        results2 = sm.OLS(y1, X1).fit()
        print(results2.summary())

```

```

=====
                                OLS Regression Results
=====
====
Dep. Variable:                logtaxrev    R-squared:
0.001
Model:                        OLS          Adj. R-squared:    -
0.010
Method:                       Least Squares    F-statistic:        0.0
6626
Date:                         Thu, 18 Sep 2025    Prob (F-statistic):
0.797
Time:                         14:08:13          Log-Likelihood:     -11
2.50
No. Observations:             98              AIC:                2
29.0
Df Residuals:                 96              BIC:                2
34.2
Df Model:                     1
Covariance Type:              nonrobust
=====
====
                                coef      std err          t      P>|t|      [0.025      0.
975]
-----
const          8.1031      0.078      104.100      0.000      7.949
8.258
res1           0.0226      0.088       0.257      0.797     -0.152
0.197
=====
====
Omnibus:                15.581    Durbin-Watson:
1.763
Prob(Omnibus):          0.000    Jarque-Bera (JB):    4
6.400
Skew:                  -0.383    Prob(JB):            8.40
e-11
Kurtosis:              6.283    Cond. No.
1.13
=====
=====

```

Notes:

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

In []: