

## ▼ 5-16-2021-Panel and DiD in Python

- Name: Jikhan Jeong
- Ch 13. Using Python for Introductory Econometrics
- Ref: <http://www.upfie.net/downloads.html>
- Ref code: <http://www.upfie.net/downloads13.html>
- statsmodels package: <https://www.statsmodels.org/stable/index.html> (for DiD)
- linearmodels package: <https://bashtage.github.io/linearmodels/> (for Panel dataset)
- Using **statsmodels**

```
# (if not installed) pip install wooldridge
import wooldridge as woo
```

```
import pandas as pd
import statsmodels.formula.api as smf
```

```
/usr/local/lib/python3.7/dist-packages/statsmodels/tools/_testing.py:19: FutureWarning: pandas.u
import pandas.util.testing as tm
```

Before running code, please running user-defiend functions in the bottom of this code.

## ▼ 13.1 Pooled OLS

```
data = woo.dataWoo('cps78_85')
```

```
data_summary(data)
```

```
(1084, 15)
Index(['educ', 'south', 'nonwhite', 'female', 'married', 'exper', 'expersq',
reg = smf.ols('lwage ~y85*(educ+female) + exper + l((exper**2)/100) + union', data )
results = reg.fit()
results.summary()
```

```

OLS Regression Results

Dep. Variable:   lwage                R-squared:    0.426
Model:          OLS                  Adj. R-squared: 0.422
Method:         Least Squares        F-statistic:   99.80
Date:           Mon, 17 May 2021      Prob (F-statistic): 4.46e-124
Time:           01:20:24              Log-Likelihood: -574.24
No. Observations: 1084                AIC:          1166.
Df Residuals:    1075                  BIC:          1211.
Df Model:         8
Covariance Type: nonrobust

               coef  std err      t    P>|t| [0.025 0.975]
Intercept      0.4589  0.093    4.911  0.000  0.276  0.642
y85             0.1178  0.124    0.952  0.341 -0.125  0.361
educ            0.0747  0.007   11.192  0.000  0.062  0.088
female         -0.3167  0.037   -8.648  0.000 -0.389 -0.245
y85:educ        0.0185  0.009    1.974  0.049  0.000  0.037
y85:female      0.0851  0.051    1.658  0.098 -0.016  0.186
exper           0.0296  0.004    8.293  0.000  0.023  0.037
l((exper ** 2) / 100) -0.0399  0.008   -5.151  0.000 -0.055 -0.025
union           0.2021  0.030    6.672  0.000  0.143  0.262

Omnibus:      83.747  Durbin-Watson:   1.918
Prob(Omnibus): 0.000  Jarque-Bera (JB): 317.985
Skew:         -0.271  Prob(JB):      8.92e-70
Kurtosis:     5.597   Cond. No.      296.
```

Warnings:

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

```
stat_ols('lwage ~y85*(educ+female) + exper + l((exper**2)/100) + union', data)
```

```

OLS Regression Results

=====
Dep. Variable:   lwage                R-squared:    0.426
Model:          OLS                  Adj. R-squared: 0.422
Method:         Least Squares        F-statistic:   99.80
Date:           Mon, 17 May 2021      Prob (F-statistic): 4.46e-124
Time:           01:35:56              Log-Likelihood: -574.24
No. Observations: 1084                AIC:          1166.
Df Residuals:    1075                  BIC:          1211.
Df Model:         8
Covariance Type: nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.4589	0.093	4.911	0.000	0.276	0.642
y85	0.1178	0.124	0.952	0.341	-0.125	0.361
educ	0.0747	0.007	11.192	0.000	0.062	0.088
female	-0.3167	0.037	-8.648	0.000	-0.389	-0.245
y85:educ	0.0185	0.009	1.974	0.049	0.000	0.037
y85:female	0.0851	0.051	1.658	0.098	-0.016	0.186
exper	0.0296	0.004	8.293	0.000	0.023	0.037
l((exper ** 2) / 100)	-0.0399	0.008	-5.151	0.000	-0.055	-0.025
union	0.2021	0.030	6.672	0.000	0.143	0.262
=====						
Omnibus:	83.747	Durbin-Watson:		1.918		
Prob(Omnibus):	0.000	Jarque-Bera (JB):		317.985		
Skew:	-0.271	Prob(JB):		8.92e-70		
Kurtosis:	5.597	Cond. No.		296.		
=====						

## Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.  
 <statsmodels.regression.linear\_model.RegressionResultsWrapper at 0x7f3dcb8dc750>

```
dir(results)
```

```
'centered_tss',
'compare_f_test',
'compare_lm_test',
'compare_lr_test',
'condition_number',
'conf_int',
'conf_int_el',
'cov_HC0',
'cov_HC1',
'cov_HC2',
'cov_HC3',
'cov_kwds',
'cov_params',
'cov_type',
'df_model',
'df_resid',
'diagn',
'eigenvals',
'el_test',
'ess',
'f_pvalue',
'f_test',
'fittedvalues',
'fvalue',
'get_influence',
'get_prediction',
'get_robustcov_results',
'het_scale',
'initialize',
'k_constant',
'llf',
'load',
'model',
...
```

```

'mse_model',
'mse_resid',
'mse_total',
'nobs',
'normalized_cov_params',
'outlier_test',
'params',
'predict',
'pvalues',
'remove_data',
'resid',
'resid_pearson',
'rsquared',
'rsquared_adj',
'save',
'scale',
'ssr',

'summary',
'summary2',
't_test',
't_test_pairwise',
'tvalues',
'uncentered_tss',
'use_t',
'wald_test',
'wald_test_terms',
.....:)))

```

```

table = pd.DataFrame({'coefficient': round(results.params, 4),
                      'se': round(results.bse, 4),
                      't': round(results.tvalues, 4),
                      'pval': round(results.pvalues, 4)})
print(f'table: \n{table}\n')

```

```

table:

```

	coefficient	se	t	pval
Intercept	0.4589	0.0934	4.9111	0.0000
y85	0.1178	0.1238	0.9517	0.3415
educ	0.0747	0.0067	11.1917	0.0000
female	-0.3167	0.0366	-8.6482	0.0000
y85:educ	0.0185	0.0094	1.9735	0.0487
y85:female	0.0851	0.0513	1.6576	0.0977
exper	0.0296	0.0036	8.2932	0.0000
l((exper ** 2) / 100)	-0.0399	0.0078	-5.1513	0.0000
union	0.2021	0.0303	6.6722	0.0000

## ▼ difference in difference (DiD)

```

data2= woo.dataWoo('kielmc')
data_summary(data2)

```

```
(321, 25)
Index(['year', 'age', 'agesq', 'nbh', 'cbd', 'intst', 'lintst', 'price',
      'rooms', 'area', 'land', 'baths', 'dist', 'ldist', 'wind', 'lprice',
      'y81', 'larea', 'lland', 'y81ldist', 'lintstsq', 'nearinc', 'y81nrinc',
      'rprice', 'lrprice'],
      dtype='object')
```

	year	age	agesq	nbh	cbd	intst	lintst	price	rooms	area	land	baths	
0	1978	48	2304.0	4	3000.0	1000.0	6.9078	60000.0	7	1660	4578.0	1	10
1	1978	83	6889.0	4	4000.0	1000.0	6.9078	40000.0	6	2612	8370.0	2	11
2	1978	58	3364.0	4	4000.0	1000.0	6.9078	34000.0	6	1144	5000.0	1	11

```
y78 = (data2['year'] == 1978)
print(type(y78))
y78
```

```
<class 'pandas.core.series.Series'>
0      True
1      True
2      True
3      True
4      True
...
316   False
317   False
318   False
319   False
320   False
Name: year, Length: 321, dtype: bool
```

## ▶ separate regressions for 1978 and 1981:

```
print('year 1978')
y78 = (data2['year'] == 1978) # subset logic array
results78 = stat_ols('rprice ~ nearinc', data2, subset=y78)
```

year 1978

### OLS Regression Results

```
=====
Dep. Variable:          rprice    R-squared:                0.115
Model:                  OLS      Adj. R-squared:            0.112
Method:                 Least Squares    F-statistic:          41.32
Date:                  Mon, 17 May 2021    Prob (F-statistic):    4.72e-10
Time:                  01:39:43    Log-Likelihood:        -3776.4
No. Observations:      321    AIC:                    7557.
Df Residuals:          319    BIC:                    7564.
Df Model:              1
Covariance Type:       nonrobust
=====
```

```
=====
coef    std err          t      P>|t|    [0.025    0.975]
=====
```

Intercept	9.104e+04	2080.730	43.752	0.000	8.69e+04	9.51e+04
near inc	-2.446e+04	3804.807	-6.428	0.000	-3.19e+04	-1.7e+04
=====						
Omnibus:		179.474	Durbin-Watson:			1.481
Prob(Omnibus):		0.000	Jarque-Bera (JB):			1761.079
Skew:		2.116	Prob(JB):			0.00
Kurtosis:		13.666	Cond. No.			2.42
=====						

Warnings:

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

```
print('year 1981')
y81 = (data2['year'] == 1981) # subset logic array
results81 = stat_ols('rprice ~ nearinc', data2, subset=y81)
```

year 1981

#### OLS Regression Results

Dep. Variable:	rprice	R-squared:	0.115			
Model:	OLS	Adj. R-squared:	0.112			
Method:	Least Squares	F-statistic:	41.32			
Date:	Mon, 17 May 2021	Prob (F-statistic):	4.72e-10			
Time:	01:39:09	Log-Likelihood:	-3776.4			
No. Observations:	321	AIC:	7557.			
Df Residuals:	319	BIC:	7564.			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]
-----						
Intercept	9.104e+04	2080.730	43.752	0.000	8.69e+04	9.51e+04
near inc	-2.446e+04	3804.807	-6.428	0.000	-3.19e+04	-1.7e+04
=====						
Omnibus:		179.474	Durbin-Watson:			1.481
Prob(Omnibus):		0.000	Jarque-Bera (JB):			1761.079
Skew:		2.116	Prob(JB):			0.00
Kurtosis:		13.666	Cond. No.			2.42

Warnings:

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

## ► joint regression including an interaction term:

- "C" denote category variables
- "\*" will include the individual columns that were multiplied together (Interaction)
- Ref: <https://www.statsmodels.org/stable/examples/notebooks/generated/formuas.html>

```
result_joint = stat_ols('rprice ~ nearinc * C(year)', data2)
```

## OLS Regression Results

Dep. Variable:	rprice	R-squared:	0.174
Model:	OLS	Adj. R-squared:	0.166
Method:	Least Squares	F-statistic:	22.25
Date:	Mon, 17 May 2021	Prob (F-statistic):	4.22e-13
Time:	01:40:38	Log-Likelihood:	-3765.2
No. Observations:	321	AIC:	7538.
Df Residuals:	317	BIC:	7554.
Df Model:	3		
Covariance Type:	nonrobust		

  

	coef	std err	t	P> t	[0.025	0.975]
Intercept	8.252e+04	2726.910	30.260	0.000	7.72e+04	8.79e+04
C(year)[T.1981]	1.879e+04	4050.065	4.640	0.000	1.08e+04	2.68e+04
near inc	-1.882e+04	4875.322	-3.861	0.000	-2.84e+04	-9232.293
near inc:C(year)[T.1981]	-1.186e+04	7456.646	-1.591	0.113	-2.65e+04	2806.867

  

Omnibus:	192.562	Durbin-Watson:	1.557
Prob(Omnibus):	0.000	Jarque-Bera (JB):	2462.071
Skew:	2.217	Prob(JB):	0.00
Kurtosis:	15.822	Cond. No.	6.05

## Warnings:

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

## ▼ difference in difference (DiD) with log price

```
# difference in difference (DiD):
```

```
result_did_with_log_price = stat_ols('np.log(rprice) ~ near inc*C(year)', data2)
```

## OLS Regression Results

Dep. Variable:	np.log(rprice)	R-squared:	0.246
Model:	OLS	Adj. R-squared:	0.239
Method:	Least Squares	F-statistic:	34.47
Date:	Mon, 17 May 2021	Prob (F-statistic):	2.62e-19
Time:	01:47:41	Log-Likelihood:	-105.68
No. Observations:	321	AIC:	219.4
Df Residuals:	317	BIC:	234.4
Df Model:	3		
Covariance Type:	nonrobust		

  

	coef	std err	t	P> t	[0.025	0.975]
Intercept	11.2854	0.031	369.839	0.000	11.225	11.345
C(year)[T.1981]	0.1931	0.045	4.261	0.000	0.104	0.282
near inc	-0.3399	0.055	-6.231	0.000	-0.447	-0.233
near inc:C(year)[T.1981]	-0.0626	0.083	-0.751	0.453	-0.227	0.102

```

Omnibus:                29.462    Durbin-Watson:                1.568
Prob(Omnibus):           0.000    Jarque-Bera (JB):           84.380
Skew:                    0.370    Prob(JB):                   4.75e-19
Kurtosis:                5.400    Cond. No.                   6.05
=====

```

Warnings:

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

## ➤ DiD with control variables

# DiD with control variables:

```

result_did_with_log_price_controls = stat_ols('np.log(rprice) ~ nearinc*C(year) + age +
        'l(age**2) + np.log(intst) + np.log(land) +
        'np.log(area) + rooms + baths', data2)

```

### OLS Regression Results

Dep. Variable:	np.log(rprice)	R-squared:	0.733
Model:	OLS	Adj. R-squared:	0.724
Method:	Least Squares	F-statistic:	84.91
Date:	Mon, 17 May 2021	Prob (F-statistic):	1.24e-82
Time:	01:51:30	Log-Likelihood:	60.690
No. Observations:	321	AIC:	-99.38
Df Residuals:	310	BIC:	-57.89
Df Model:	10		
Covariance Type:	nonrobust		

  

	coef	std err	t	P> t	[0.025	0.975]
Intercept	7.6517	0.416	18.399	0.000	6.833	8.470
C(year)[T.1981]	0.1621	0.028	5.687	0.000	0.106	0.218
near inc	0.0322	0.047	0.679	0.498	-0.061	0.126
near inc:C(year)[T.1981]	-0.1315	0.052	-2.531	0.012	-0.234	-0.029
age	-0.0084	0.001	-5.924	0.000	-0.011	-0.006
l(age ** 2)	3.763e-05	8.67e-06	4.342	0.000	2.06e-05	5.47e-05
np.log(intst)	-0.0614	0.032	-1.950	0.052	-0.123	0.001
np.log(land)	0.0998	0.024	4.077	0.000	0.052	0.148
np.log(area)	0.3508	0.051	6.813	0.000	0.249	0.452
rooms	0.0473	0.017	2.732	0.007	0.013	0.081
baths	0.0943	0.028	3.400	0.001	0.040	0.149

  

Omnibus:	67.366	Durbin-Watson:	1.710
Prob(Omnibus):	0.000	Jarque-Bera (JB):	356.621
Skew:	-0.734	Prob(JB):	3.64e-78
Kurtosis:	7.951	Cond. No.	1.83e+05

Warnings:

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



[2] The condition number is large,  $1.83e+05$ . This might indicate that there are strong multicollinearity or other numerical problems.

## 13.4 Panel: First Differenced Estimator

```
# (if not installed) pip install linearmodels
import linearmodels as plm
```

```
data3 = woo.dataWoo('crime4')
data_summary(data3)
```

↳ (630, 59)

```
Index(['county', 'year', 'crmte', 'prbarr', 'prbconv', 'prbpris', 'avgsgen',
      'polpc', 'density', 'taxpc', 'west', 'central', 'urban', 'pctmin80',
      'wcon', 'wtuc', 'wtrd', 'wfir', 'wser', 'wmfg', 'wfed', 'wsta', 'wloc',
      'mix', 'pctymle', 'd82', 'd83', 'd84', 'd85', 'd86', 'd87', 'lcrmte',
      'lprbarr', 'lprbconv', 'lprbpris', 'lavgsgen', 'lpolpc', 'ldensity',
      'ltaxpc', 'lwcon', 'lwtuc', 'lwtrd', 'lwfir', 'lwser', 'lwmfg', 'lwfed',
      'lwsta', 'lwloc', 'lmix', 'lpctymle', 'lpctmin', 'clcrmte', 'clprbarr',
      'clprbcon', 'clprbpri', 'clavgsgen', 'clpolpc', 'cltaxpc', 'clmix'],
      dtype='object')
```

	county	year	crmte	prbarr	prbconv	prbpris	avgsgen	polpc	density	taxpc
0	1	81	0.039885	0.289696	0.402062	0.472222	5.61	0.001787	2.307159	25.69
1	1	82	0.038345	0.338111	0.433005	0.506993	5.59	0.001767	2.330254	24.87
2	1	83	0.030305	0.330449	0.525703	0.479705	5.80	0.001836	2.341801	26.45

```
data3 = data3.set_index(['county', 'year'], drop=False)
data3.head()
```

		county	year	crm rte	prb arr	prb conv	prb pris	avg sen	pol pc	den
county	year									
1	81	1	81	0.039885	0.289696	0.402062	0.472222	5.61	0.001787	2.30
	82	1	82	0.038345	0.338111	0.433005	0.506993	5.59	0.001767	2.33
	83	1	83	0.030305	0.330449	0.525703	0.479705	5.80	0.001836	2.34
	84	1	84	0.034726	0.362525	0.604706	0.520104	6.89	0.001886	2.34
	85	1	85	0.036573	0.325395	0.578723	0.497059	6.55	0.001924	2.36

```
reg = plm.FirstDifferenceOLS.from_formula('np.log(crmte) ~ year + d83 + d84 + d85 + d86 + d87 + lprbar',
first_difference_results = reg.fit()
print(first_difference_results)
```

## FirstDifferenceOLS Estimation Summary

Dep. Variable:	np.log(crmrte)	R-squared:	0.4326
Estimator:	FirstDifferenceOLS	R-squared (Between):	0.6003
No. Observations:	540	R-squared (Within):	0.4281
Date:	Mon, May 17 2021	R-squared (Overall):	0.6000
Time:	02:05:38	Log-likelihood	248.48
Cov. Estimator:	Unadjusted		
		F-statistic:	36.661
Entities:	90	P-value	0.0000
Avg Obs:	7.0000	Distribution:	F(11,529)
Min Obs:	7.0000		
Max Obs:	7.0000	F-statistic (robust):	36.661
		P-value	0.0000
Time periods:	7	Distribution:	F(11,529)
Avg Obs:	90.000		
Min Obs:	90.000		
Max Obs:	90.000		

## Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
year	0.0077	0.0171	0.4522	0.6513	-0.0258	0.0412
d83	-0.0999	0.0239	-4.1793	0.0000	-0.1468	-0.0529
d84	-0.1478	0.0413	-3.5806	0.0004	-0.2289	-0.0667
d85	-0.1524	0.0584	-2.6098	0.0093	-0.2671	-0.0377
d86	-0.1249	0.0760	-1.6433	0.1009	-0.2742	0.0244
d87	-0.0841	0.0940	-0.8944	0.3715	-0.2687	0.1006
lprbarr	-0.3275	0.0300	-10.924	0.0000	-0.3864	-0.2686
lprbconv	-0.2381	0.0182	-13.058	0.0000	-0.2739	-0.2023
lprbpris	-0.1650	0.0260	-6.3555	0.0000	-0.2161	-0.1140
lavgsen	-0.0218	0.0221	-0.9850	0.3251	-0.0652	0.0216
lpolpc	0.3984	0.0269	14.821	0.0000	0.3456	0.4512

## ▼ User-Define Functions for Statmodels

- Name: Jikhan Jeong
- Date: 5-16-2021 (Updated)

```
import numpy as np
import pandas as pd
import statsmodels.formula.api as smf
```

```
def data_summary(df):
    print(df.shape)
    print(df.columns)
    return df.head(3)
```

```
def stat_ols(formulas,df,subset=None, drop_cols=None):
```

```
'''
Name: Jikhan Jeong
* Simpler ols
* Requires: import statsmodels.formula.api as smf
'''

reg = smf.ols(formulas, df, subset=None, drop_cols=None)
results = reg.fit()
print(results.summary())
return results

# (ex) stat_ols('lwage ~ y85*(educ+female) + exper + 1*((exper**2)/100) + union', data)
```

---

✓ 0초    오후 7:05에 완료됨

● ✕