

```
In [1]: import numpy as np
import pandas as pd
from scipy import stats
# set up panel data
pdata = pd.read_csv("CB data-1.csv", index_col = ["ID", "Year"],
                    parse_dates = True)

pdata
# alternatively
# data = pd.read_csv("http://web.pdx.edu/~crkl/ceR/data/airline.txt", sep='\s+',
# Set data as panel data
# pdata = data.set_index(['I', 'T'], inplace=True)
```

Out[1]:

		SL	Market Capitalization	Market Cap Growth	Debt / Equity Ratio	Debt / FCF Ratio	Return on Equity (ROE)	Return on Assets (ROA)	Return on Capital (ROIC)	Log Total Assets	Current Circulat (\$)
ID	Year										
TMP	2019-01-01	1	1,367,031,594	0.193	1.112	7.657	0.125	0.012	0.014	15.230	7.4
	2018-01-01	1	1,145,996,404	-0.073	1.896	12.701	0.138	0.012	0.014	15.222	7.4
	2017-01-01	1	1,236,784,388	-0.131	2.024	22.846	0.091	0.008	0.013	15.308	7.3
	2016-01-01	1	1,423,389,891	0.701	1.810	13.276	0.108	0.010	0.013	15.311	7.3
	2015-01-01	1	836,855,660	0.025	1.379	9.279	0.115	0.011	0.015	15.317	7.2
...
BCS	2014-01-01	54	60,465,197,835	0.090	11.887	-65.859	-0.003	0.000	0.000	15.319	7.3
	2013-01-01	54	55,489,736,570	0.050	12.579	-27.133	0.010	0.000	0.001	15.200	7.3
	2012-01-01	54	52,823,723,087	0.578	17.739	-63.139	-0.012	0.000	0.000	14.881	7.0
	2011-01-01	54	33,469,882,543	-0.290	17.875	34.557	0.056	0.002	0.002	14.877	6.9
	2010-01-01	54	47,129,813,960	0.279	17.435	52.410	0.073	0.002	0.002	14.688	6.8

540 rows × 12 columns



```
In [5]: pdata.describe()
```

Out[5]:

	Market Cap Growth	Debt / Equity Ratio	Debt / FCF Ratio	Return on Equity (ROE)	Return on Assets (ROA)	Return on Capital (ROIC)	Log Total Assets	Cur Circu (
count	540.000000	540.000000	540.000000	540.000000	540.000000	540.000000	540.000000	540.00
mean	0.216502	1.136930	6.343000	0.086111	0.009646	0.013672	15.135300	7.2
std	0.376532	1.832502	38.175251	0.056684	0.006142	0.007287	0.219199	0.19
min	-0.466000	0.041000	-365.473000	-0.692000	-0.069000	-0.060000	14.688000	6.88
25%	-0.019250	0.433250	2.885500	0.072000	0.008000	0.011000	14.881000	7.04
50%	0.135500	0.768500	5.608000	0.090000	0.010000	0.014000	15.226000	7.22
75%	0.366500	1.148250	9.802000	0.109000	0.012000	0.017000	15.311000	7.38
max	3.730000	17.875000	530.212000	0.340000	0.036000	0.040000	15.319000	7.49



```
In [6]: column_names = {'Debt / Equity Ratio': 'DE',
                        'Debt / FCF Ratio': 'DF',
                        'Liquidity Ratio': 'LR',
                        'Return on Equity (ROE)': 'ROE',
                        'Return on Assets (ROA)': 'ROA',
                        'Return on Capital (ROIC)': 'ROIC',
                        'Log Total Assets': 'TA',
                        'Log Currency in Circulation ($ Bil)': 'CC',
                        'Effective Federal Funds Rate (%)': 'FFR',
                        'Loss Function': 'LF'}

# rename columns
pdata = pdata.rename(columns = column_names)

pdata
```

Out[6]:

		Market Capitalization	Market Cap Growth	DE	DF	ROE	ROA	ROIC	TA	CC	FFR	
ID	Year											
TMP	2019-01-01	1,367,031,594	0.193	1.112	7.657	0.125	0.012	0.014	15.230	7.492	1.551	(
	2018-01-01	1,145,996,404	-0.073	1.896	12.701	0.138	0.012	0.014	15.222	7.442	2.274	(
	2017-01-01	1,236,784,388	-0.131	2.024	22.846	0.091	0.008	0.013	15.308	7.381	1.302	-1
	2016-01-01	1,423,389,891	0.701	1.810	13.276	0.108	0.010	0.013	15.311	7.313	0.540	-1
	2015-01-01	836,855,660	0.025	1.379	9.279	0.115	0.011	0.015	15.317	7.255	0.241	-1
...	
BCS	2014-01-01	60,465,197,835	0.090	11.887	-65.859	-0.003	0.000	0.000	15.319	7.190	0.123	-1
	2013-01-01	55,489,736,570	0.050	12.579	-27.133	0.010	0.000	0.001	15.200	7.115	0.085	-1
	2012-01-01	52,823,723,087	0.578	17.739	-63.139	-0.012	0.000	0.000	14.881	7.053	0.165	-1
	2011-01-01	33,469,882,543	-0.290	17.875	34.557	0.056	0.002	0.002	14.877	6.972	0.072	-2
	2010-01-01	47,129,813,960	0.279	17.435	52.410	0.073	0.002	0.002	14.688	6.887	0.183	-2

540 rows × 11 columns

```
In [8]: df = pdata
df
```

Out[8]:

		Market Capitalization	Market Cap Growth	DE	DF	ROE	ROA	ROIC	TA	CC	FFR	
ID	Year											
TMP	2019-01-01	1,367,031,594	0.193	1.112	7.657	0.125	0.012	0.014	15.230	7.492	1.551	(
	2018-01-01	1,145,996,404	-0.073	1.896	12.701	0.138	0.012	0.014	15.222	7.442	2.274	(
	2017-01-01	1,236,784,388	-0.131	2.024	22.846	0.091	0.008	0.013	15.308	7.381	1.302	-1
	2016-01-01	1,423,389,891	0.701	1.810	13.276	0.108	0.010	0.013	15.311	7.313	0.540	-1
	2015-01-01	836,855,660	0.025	1.379	9.279	0.115	0.011	0.015	15.317	7.255	0.241	-1
...	
BCS	2014-01-01	60,465,197,835	0.090	11.887	-65.859	-0.003	0.000	0.000	15.319	7.190	0.123	-1
	2013-01-01	55,489,736,570	0.050	12.579	-27.133	0.010	0.000	0.001	15.200	7.115	0.085	-1
	2012-01-01	52,823,723,087	0.578	17.739	-63.139	-0.012	0.000	0.000	14.881	7.053	0.165	-1
	2011-01-01	33,469,882,543	-0.290	17.875	34.557	0.056	0.002	0.002	14.877	6.972	0.072	-2
	2010-01-01	47,129,813,960	0.279	17.435	52.410	0.073	0.002	0.002	14.688	6.887	0.183	-2

540 rows × 11 columns



```
In [9]: # Descriptive statistics
df.describe()
```

Out[9]:

	Market Cap Growth	DE	DF	ROE	ROA	ROIC	TA	
count	540.000000	540.000000	540.000000	540.000000	540.000000	540.000000	540.000000	540.000000
mean	0.216502	1.136930	6.343000	0.086111	0.009646	0.013672	15.135300	7.210000
std	0.376532	1.832502	38.175251	0.056684	0.006142	0.007287	0.219199	0.190000
min	-0.466000	0.041000	-365.473000	-0.692000	-0.069000	-0.060000	14.688000	6.880000
25%	-0.019250	0.433250	2.885500	0.072000	0.008000	0.011000	14.881000	7.000000
50%	0.135500	0.768500	5.608000	0.090000	0.010000	0.014000	15.226000	7.210000
75%	0.366500	1.148250	9.802000	0.109000	0.012000	0.017000	15.311000	7.310000
max	3.730000	17.875000	530.212000	0.340000	0.036000	0.040000	15.319000	7.410000

```
In [10]: # Pooled OLS estimator
from linearmodels import PooledOLS
pooled = PooledOLS.from_formula('ROE ~ 1 + DE + DF + TA + CC + FFR', df).fit()
print(pooled)
```

PooledOLS Estimation Summary

```
=====
=
Dep. Variable:          ROE    R-squared:          0.077
1
Estimator:              PooledOLS    R-squared (Between):    0.053
7
No. Observations:      540    R-squared (Within):    0.083
3
Date:                  Thu, Apr 21 2022    R-squared (Overall):    0.077
1
Time:                  00:00:12    Log-likelihood          805.8
9
Cov. Estimator:        Unadjusted
                                F-statistic:          8.926
9
Entities:              54    P-value          0.000
0
Avg Obs:              10.0000    Distribution:          F(5,53
4)
Min Obs:              10.0000
Max Obs:              10.0000    F-statistic (robust):    8.926
9
                                P-value          0.000
0
Time periods:          10    Distribution:          F(5,53
4)
Avg Obs:              54.000
Min Obs:              54.000
Max Obs:              54.000
```

Parameter Estimates

```
=====
Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
Intercept  -0.5927    0.1799   -3.2940    0.0011   -0.9462   -0.2392
CC          0.0719    0.0525    1.3690    0.1716   -0.0313    0.1750
DE         -0.0021    0.0013   -1.6112    0.1077   -0.0046    0.0005
DF          1.683e-05  6.231e-05  0.2700    0.7872   -0.0001    0.0001
FFR         -0.0018    0.0088   -0.2072    0.8359   -0.0190    0.0154
TA          0.0108    0.0294    0.3679    0.7131   -0.0470    0.0687
=====
```



```
In [11]: # Between estimator
from linearmodels import BetweenOLS
between = BetweenOLS.from_formula('ROE ~ 1 + DE + DF + TA + CC + FFR', df).fit()
print(between)
```

BetweenOLS Estimation Summary

```
=====
=
Dep. Variable:          ROE    R-squared:          0.073
4
Estimator:             BetweenOLS    R-squared (Between):    0.073
4
No. Observations:      54    R-squared (Within):    0.001
1
Date:                  Thu, Apr 21 2022    R-squared (Overall):    0.016
2
Time:                  00:01:14    Log-likelihood    122.8
2
Cov. Estimator:        Unadjusted
                                F-statistic:          0.760
7
Entities:              54    P-value    0.582
6
Avg Obs:               10.0000    Distribution:    F(5,4
8)
Min Obs:               10.0000
Max Obs:               10.0000    F-statistic (robust):    0.760
7
                                P-value    0.582
6
Time periods:          10    Distribution:    F(5,4
8)
Avg Obs:               54.000
Min Obs:               54.000
Max Obs:               54.000
```

Parameter Estimates

```
=====
Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
Intercept   0.0003  6.266e+04  5.193e-09    1.0000   -1.26e+05   1.26e+05
CC          0.0023  1.219e+04  1.924e-07    1.0000   -2.451e+04  2.451e+04
DE         -0.0044   0.0023  -1.9310    0.0594   -0.0089    0.0002
DF         -0.0001   0.0004  -0.3806    0.7052   -0.0008    0.0006
FFR         0.0002
TA          0.0049   1647.2   2.99e-06    1.0000   -3311.9    3311.9
=====
```

```
C:\Users\HP\anaconda3\lib\site-packages\linearmodels\panel\results.py:87: RuntimeWarning: invalid value encountered in sqrt
    return Series(np.sqrt(np.diag(self.cov)), self._var_names, name="std_error")
```



```
In [12]: # First differences estimator (without constant term)
from linearmodels import FirstDifferenceOLS
firstdiff = FirstDifferenceOLS.from_formula('ROE ~ DE + DF + TA + CC + FFR', df)
print(firstdiff)
```

FirstDifferenceOLS Estimation Summary

```
=====
=
Dep. Variable:          ROE    R-squared:          0.041
5
Estimator:      FirstDifferenceOLS    R-squared (Between):    -54.45
8
No. Observations:      486    R-squared (Within):      0.046
4
Date:      Thu, Apr 21 2022    R-squared (Overall):    -41.43
1
Time:      00:02:40    Log-likelihood      833.5
2
Cov. Estimator:      Unadjusted
F-statistic:          4.163
9
Entities:          54    P-value          0.001
0
Avg Obs:          10.0000    Distribution:      F(5,48
1)
Min Obs:          10.0000
Max Obs:          10.0000    F-statistic (robust):    4.163
9
P-value          0.001
0
Time periods:          10    Distribution:      F(5,48
1)
Avg Obs:          54.000
Min Obs:          54.000
Max Obs:          54.000
```

Parameter Estimates

```
=====
Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
CC          0.1047    0.0391    2.6750    0.0077    0.0278    0.1816
DE         -0.0008    0.0047   -0.1625    0.8710   -0.0100    0.0084
DF        3.782e-05  3.494e-05  1.0822    0.2797  -3.084e-05  0.0001
FFR         0.0067    0.0048    1.3907    0.1650   -0.0028    0.0161
TA         -0.0002    0.0193   -0.0114    0.9909   -0.0381    0.0377
=====
```

```
In [13]: from linearmodels.panel.results import compare
res1 = {'Pooled':pooled, 'Between':between, 'firstdiff':firstdiff}
print(compare(res1))
```

Model Comparison			
	Pooled	Between	firstdiff
Dep. Variable	ROE	ROE	ROE
Estimator	PooledOLS	BetweenOLS	FirstDifferenceOLS
No. Observations	540	54	486
Cov. Est.	Unadjusted	Unadjusted	Unadjusted
R-squared	0.0771	0.0734	0.0415
R-Squared (Within)	0.0833	0.0011	0.0464
R-Squared (Between)	0.0537	0.0734	-54.458
R-Squared (Overall)	0.0771	0.0162	-41.431
F-statistic	8.9269	0.7607	4.1639
P-value (F-stat)	0.0000	0.5826	0.0010
Intercept	-0.5927 (-3.2940)	0.0003 (5.193e-09)	
CC	0.0719 (1.3690)	0.0023 (1.924e-07)	0.1047 (2.6750)
DE	-0.0021 (-1.6112)	-0.0044 (-1.9310)	-0.0008 (-0.1625)
DF	1.683e-05 (0.2700)	-0.0001 (-0.3806)	3.782e-05 (1.0822)
FFR	-0.0018 (-0.2072)	0.0002	0.0067 (1.3907)
TA	0.0108 (0.3679)	0.0049 (2.99e-06)	-0.0002 (-0.0114)

T-stats reported in parentheses

```
In [15]: # Fixed effects or within estimator
# with constant included or not, will have the same results
# with constant term suppressed
from linearmodels import PanelOLS
fixed = PanelOLS.from_formula('ROE ~ DE + DF + TA + CC + FFR + EntityEffects', data=df)
print(fixed)
# extract fixed effects
fixed.estimated_effects
fixed_effects = fixed.estimated_effects.unstack(level=0).values[0]
print(fixed_effects)
# F test for fixed effects versus OLS
print(fixed.f_pooled)
```

PanelOLS Estimation Summary

```
=====
=
Dep. Variable:          ROE    R-squared:          0.109
9
Estimator:              PanelOLS    R-squared (Between):    -67.31
2
No. Observations:      540    R-squared (Within):      0.109
9
Date:                  Thu, Apr 21 2022    R-squared (Overall):    -51.19
8
Time:                  00:05:20    Log-likelihood          878.7
6
Cov. Estimator:        Unadjusted
9
                        F-statistic:          11.87
Entities:              54    P-value          0.000
0
Avg Obs:               10.0000    Distribution:          F(5,48
1)
Min Obs:               10.0000
Max Obs:               10.0000    F-statistic (robust):    11.87
9
                        P-value          0.000
0
Time periods:          10    Distribution:          F(5,48
1)
Avg Obs:               54.000
Min Obs:               54.000
Max Obs:               54.000
```

Parameter Estimates

```
=====
Parameter    Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
CC            0.0849    0.0484    1.7524    0.0803    -0.0103    0.1801
DE            0.0101    0.0032    3.1429    0.0018    0.0038    0.0165
DF           -6.722e-07    6.001e-05    -0.0112    0.9911    -0.0001    0.0001
FFR           -0.0022    0.0081    -0.2688    0.7882    -0.0180    0.0137
TA            0.0136    0.0271    0.5024    0.6156    -0.0397    0.0669
=====
```

F-test for Poolability: 2.8116

P-value: 0.0000

Distribution: F(53,481)

Included effects: Entity

[-0.75102022	-0.75544832	-0.7694817	-0.92007384	-0.7310765	-0.7423119
-0.72793233	-0.71660087	-0.73842161	-0.7471241	-0.7196085	-0.71755112
-0.69902119	-0.72929546	-0.77850417	-0.75664868	-0.73428696	-0.74676802
-0.70457462	-0.73489494	-0.72053978	-0.7360623	-0.75119472	-0.71635121
-0.76546751	-0.73813533	-0.75354623	-0.74317909	-0.69689894	-0.72652242
-0.78695378	-0.71942456	-0.73453788	-0.73736042	-0.7041424	-0.74723873
-0.74485564	-0.69565661	-0.77202361	-0.73935229	-0.72717435	-0.73073287
-0.71866045	-0.74560486	-0.83803986	-0.76517839	-0.75365448	-0.74699964
-0.70380216	-0.71451607	-0.74568589	-0.74999114	-0.7408618	-0.75306237]

Pooled F-statistic

H0: Effects are zero

Statistic: 2.8116

P-value: 0.0000

Distributed: F(53,481)



```
In [16]: # with constant term included
fixed1 = PanelOLS.from_formula('ROE ~ 1 + DE + DF + TA + CC + FFR + EntityEffects', data)
print(fixed1)
# extract fixed effects
fixed1.estimated_effects
fixed1_effects = fixed1.params.Intercept + fixed1.estimated_effects.unstack(level1)
print(fixed1_effects)
```

PanelOLS Estimation Summary

```
=====
=
Dep. Variable:          ROE    R-squared:          0.109
9
Estimator:              PanelOLS    R-squared (Between):    -0.787
0
No. Observations:      540    R-squared (Within):      0.109
9
Date:                  Thu, Apr 21 2022    R-squared (Overall):    -0.077
0
Time:                  00:08:04    Log-likelihood          878.7
6
Cov. Estimator:        Unadjusted
9
                        F-statistic:          11.87
Entities:              54    P-value          0.000
0
Avg Obs:              10.0000    Distribution:          F(5,48
1)
Min Obs:              10.0000
Max Obs:              10.0000    F-statistic (robust):    11.87
9
                        P-value          0.000
0
Time periods:          10    Distribution:          F(5,48
1)
Avg Obs:              54.000
Min Obs:              54.000
Max Obs:              54.000
```

Parameter Estimates

```
=====
Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
Intercept  -0.7423    0.1695   -4.3781    0.0000   -1.0754   -0.4092
CC          0.0849    0.0484    1.7524    0.0803   -0.0103    0.1801
DE          0.0101    0.0032    3.1429    0.0018    0.0038    0.0165
DF        -6.722e-07  6.001e-05  -0.0112    0.9911   -0.0001    0.0001
FFR        -0.0022    0.0081   -0.2688    0.7882   -0.0180    0.0137
TA          0.0136    0.0271    0.5024    0.6156   -0.0397    0.0669
=====
```

F-test for Poolability: 2.8116

P-value: 0.0000

Distribution: F(53,481)

Included effects: Entity

[-0.75102022 -0.75544832 -0.7694817 -0.92007384 -0.7310765 -0.7423119

```
-0.72793233 -0.71660087 -0.73842161 -0.7471241 -0.7196085 -0.71755112  
-0.69902119 -0.72929546 -0.77850417 -0.75664868 -0.73428696 -0.74676802  
-0.70457462 -0.73489494 -0.72053978 -0.7360623 -0.75119472 -0.71635121  
-0.76546751 -0.73813533 -0.75354623 -0.74317909 -0.69689894 -0.72652242  
-0.78695378 -0.71942456 -0.73453788 -0.73736042 -0.7041424 -0.74723873  
-0.74485564 -0.69565661 -0.77202361 -0.73935229 -0.72717435 -0.73073287  
-0.71866045 -0.74560486 -0.83803986 -0.76517839 -0.75365448 -0.74699964  
-0.70380216 -0.71451607 -0.74568589 -0.74999114 -0.7408618 -0.75306237]
```

```
In [17]: # F test for fixed effects versus OLS  
print(fixed1.f_pooled)
```

```
Pooled F-statistic  
H0: Effects are zero  
Statistic: 2.8116  
P-value: 0.0000  
Distributed: F(53,481)
```

```
In [18]: # Random effects estimator, constant term must be included
# should not have EntityEffects or TimeEffects in the formula
from linearmodels import RandomEffects
random = RandomEffects.from_formula('ROE ~ 1 + DE + DF + TA + CC + FFR', df).fit()
print(random)
# extract fixed effects
random.estimated_effects
random_effects = random.params.Intercept + random.estimated_effects.unstack(level=0)
print(random_effects)
print(random.variance_decomposition)
```

```

                                RandomEffects Estimation Summary
=====
=
Dep. Variable:                  ROE    R-squared:                  0.083
7
Estimator:                    RandomEffects    R-squared (Between):    -0.009
0
No. Observations:              540    R-squared (Within):        0.092
6
Date:                        Thu, Apr 21 2022    R-squared (Overall):        0.071
4
Time:                        00:08:51    Log-likelihood              846.2
1
Cov. Estimator:              Unadjusted
                                F-statistic:                  9.755
9
Entities:                    54    P-value                  0.000
0
Avg Obs:                    10.0000    Distribution:              F(5,53
4)
Min Obs:                    10.0000
Max Obs:                    10.0000    F-statistic (robust):      9.755
9
                                P-value                  0.000
0
Time periods:                10    Distribution:              F(5,53
4)
Avg Obs:                    54.000
Min Obs:                    54.000
Max Obs:                    54.000

```

```

                                Parameter Estimates
=====
                                Parameter    Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
Intercept    -0.6232    0.1678    -3.7139    0.0002    -0.9529    -0.2936
CC            0.0746    0.0488    1.5295    0.1267    -0.0212    0.1703
DE            0.0003    0.0018    0.1505    0.8804    -0.0032    0.0038
DF            1.866e-05    5.93e-05    0.3146    0.7532    -9.783e-05    0.0001
FFR           -0.0019    0.0081    -0.2355    0.8139    -0.0179    0.0141
TA            0.0114    0.0273    0.4172    0.6767    -0.0423    0.0651
=====
[-0.62895178 -0.62967337 -0.63722145 -0.65141869 -0.61752029 -0.62472332
-0.61988231 -0.6158958 -0.62344079 -0.6288646 -0.61593738 -0.61426796
-0.60856247 -0.62060387 -0.63703279 -0.62929595 -0.61837852 -0.62632259
-0.6093808 -0.62065634 -0.61495394 -0.6181889 -0.63067895 -0.61436772

```



```
-0.63423202 -0.62163232 -0.62976787 -0.62442547 -0.60738745 -0.61887655
-0.62207593 -0.61455128 -0.62356593 -0.62279974 -0.6102221 -0.62442348
-0.62646543 -0.60772057 -0.63663799 -0.62489058 -0.61555647 -0.6219407
-0.61275883 -0.62623118 -0.66376693 -0.63537912 -0.62591989 -0.62512677
-0.61094108 -0.60989269 -0.62347709 -0.62778243 -0.62130009 -0.62928946]
Effects          0.000443
Residual         0.002537
Percent due to Effects  0.148685
Name: Variance Decomposition, dtype: float64
```



```
In [19]: # compare fixed effects and random effects models
res2 = {'Pooled':pooled,'Fixed+1':fixed1,'Fixed':fixed,'Random':random}
print(compare(res2))

effects = pd.DataFrame({'Fixed Effects':fixed_effects,'Random Effects':random_eff
                        index=pdata.index.levels[0])
print(effects)
```

Model Comparison				
	Pooled	Fixed+1	Fixed	
Random				
Dep. Variable	ROE	ROE	ROE	
ROE				
Estimator	PooledOLS	PanelOLS	PanelOLS	Random
Effects				
No. Observations	540	540	540	
540				
Cov. Est.	Unadjusted	Unadjusted	Unadjusted	Una
djusted				
R-squared	0.0771	0.1099	0.1099	
0.0837				
R-Squared (Within)	0.0833	0.1099	0.1099	
0.0926				
R-Squared (Between)	0.0537	-0.7870	-67.312	
-0.0090				
R-Squared (Overall)	0.0771	-0.0770	-51.198	
0.0714				
F-statistic	8.9269	11.879	11.879	
9.7559				
P-value (F-stat)	0.0000	0.0000	0.0000	
0.0000				
Intercept	-0.5927	-0.7423		
-0.6232				
	(-3.2940)	(-4.3781)		(-
3.7139)				
CC	0.0719	0.0849	0.0849	
0.0746				
	(1.3690)	(1.7524)	(1.7524)	
(1.5295)				
DE	-0.0021	0.0101	0.0101	
0.0003				
	(-1.6112)	(3.1429)	(3.1429)	
(0.1505)				
DF	1.683e-05	-6.722e-07	-6.722e-07	1.
866e-05				
	(0.2700)	(-0.0112)	(-0.0112)	
(0.3146)				
FFR	-0.0018	-0.0022	-0.0022	
-0.0019				
	(-0.2072)	(-0.2688)	(-0.2688)	(-
0.2355)				

TA	0.0108	0.0136	0.0136
0.0114			
	(0.3679)	(0.5024)	(0.5024)
(0.4172)			

```

=====
=====
Effects                                     Entity      Entity
-----
-----

```

T-stats reported in parentheses

	Fixed Effects	Random Effects
--	---------------	----------------

ID		
ABCB	-0.751020	-0.628952
AUB	-0.755448	-0.629673
BANR	-0.769482	-0.637221
BCS	-0.920074	-0.651419
BOKF	-0.731077	-0.617520
CATY	-0.742312	-0.624723
CBU	-0.727932	-0.619882
CFR	-0.716601	-0.615896
CMA	-0.738422	-0.623441
COLB	-0.747124	-0.628865
CTBI	-0.719608	-0.615937
CVBF	-0.717551	-0.614268
EWBC	-0.699021	-0.608562
FCNCA	-0.729295	-0.620604
FHN	-0.778504	-0.637033
FNB	-0.756649	-0.629296
FNLC	-0.734287	-0.618379
FULT	-0.746768	-0.626323
GABC	-0.704575	-0.609381
GBCI	-0.734895	-0.620656
GSBC	-0.720540	-0.614954
HBNC	-0.736062	-0.618189
HFWA	-0.751195	-0.630679
HOMB	-0.716351	-0.614368
HTH	-0.765468	-0.634232
HTLF	-0.738135	-0.621632
HWC	-0.753546	-0.629768
IBAI	-0.743179	-0.624425
IKFN	-0.696899	-0.607387
INDB	-0.726522	-0.618877
MFG	-0.786954	-0.622076
NTRS	-0.719425	-0.614551
PB	-0.734538	-0.623566
PPBI	-0.737360	-0.622800
RJF	-0.704142	-0.610222
SASR	-0.747239	-0.624423
SFNC	-0.744856	-0.626465
SIVB	-0.695657	-0.607721
SNV	-0.772024	-0.636638
SSB	-0.739352	-0.624891
TCBI	-0.727174	-0.615556
THFF	-0.730733	-0.621941
TMP	-0.718660	-0.612759
UBSI	-0.745605	-0.626231

UCBI	-0.838040	-0.663767
UMPQ	-0.765178	-0.635379
VLY	-0.753654	-0.625920
WAFD	-0.747000	-0.625127
WAL	-0.703802	-0.610941
WASH	-0.714516	-0.609893
WBS	-0.745686	-0.623477
WSBC	-0.749991	-0.627782
WSFS	-0.740862	-0.621300
WTFC	-0.753062	-0.629289

In [20]: *# LM test for random effects versus OLS*

```
n = pdata.index.levels[0].size
T = pdata.index.levels[1].size
D = np.kron(np.eye(n), np.ones(T)).T
e = pooled.resids
LM = (e.dot(D).dot(D.T).dot(e) / e.dot(e) - 1) ** 2 * n * T / 2 / (T - 1)
LM_pvalue = stats.chi2(1).sf(LM)
print("LM Test: chisq = {0}, df = 1, p-value = {1}".format(LM, LM_pvalue))
```

LM Test: chisq = 38.781752018223955, df = 1, p-value = 4.739377680850834e-10

In [21]: *# Hausman test for fixed versus random effects model*

```
# null hypothesis: random effects model
psi = fixed.cov - random.cov.iloc[1:,1:]
diff = fixed.params - random.params[1:]
# psi = fixed1.cov.iloc[1:,1:] - random.cov.iloc[1:,1:]
# diff = fixed1.params[1:] - random.params[1:]
W = diff.dot(np.linalg.inv(psi)).dot(diff)
dof = random.params.size - 1
pvalue = stats.chi2(dof).sf(W)
print("Hausman Test: chisq = {0}, df = {1}, p-value = {2}".format(W, dof, pvalue))
```

Hausman Test: chisq = 14.014889787657228, df = 5, p-value = 0.01551511032784552

5

```
In [24]: # panel robust hetero cov
fixed_robust = PanelOLS.from_formula('ROE ~ 1 + DE + DF + TA + CC + FFR + Entity', data)
print(fixed_robust)
random_robust = RandomEffects.from_formula('ROE ~ 1 + DE + DF + TA + CC + FFR', data)
print(random_robust)
```

PanelOLS Estimation Summary

```
=====
Dep. Variable:          ROE    R-squared:          0.109
Estimator:             PanelOLS  R-squared (Between): -0.787
No. Observations:      540    R-squared (Within):    0.109
Date:                  Thu, Apr 21 2022  R-squared (Overall): -0.077
Time:                  00:21:47    Log-likelihood      878.7
Cov. Estimator:        Clustered
F-statistic:           11.87
Entities:              54    P-value              0.000
Avg Obs:              10.0000  Distribution:      F(5,48)
Min Obs:              10.0000
Max Obs:              10.0000  F-statistic (robust): 12.01
P-value              0.000
Time periods:         10    Distribution:      F(5,48)
Avg Obs:              54.000
Min Obs:              54.000
Max Obs:              54.000
```

Parameter Estimates

```
=====
Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
Intercept  -0.7423    0.3341   -2.2216    0.0268    -1.3988    -0.0858
CC          0.0849    0.0250    3.4005    0.0007     0.0358     0.1340
DE          0.0101    0.0037    2.7018    0.0071     0.0028     0.0175
DF        -6.722e-07  4.252e-05 -0.0158    0.9874   -8.423e-05  8.288e-05
FFR        -0.0022    0.0039   -0.5541    0.5798    -0.0099     0.0055
TA          0.0136    0.0177    0.7713    0.4409    -0.0211     0.0483
=====
```

F-test for Poolability: 2.8116

P-value: 0.0000

Distribution: F(53,481)

Included effects: Entity

RandomEffects Estimation Summary

```
=====
```

```

=
Dep. Variable:                ROE    R-squared:                0.083
7
Estimator:                    RandomEffects    R-squared (Between):    -0.009
0
No. Observations:            540    R-squared (Within):    0.092
6
Date:                        Thu, Apr 21 2022    R-squared (Overall):    0.071
4
Time:                        00:21:47    Log-likelihood            846.2
1
Cov. Estimator:              Clustered
                                F-statistic:                9.755
9
Entities:                    54    P-value                    0.000
0
Avg Obs:                     10.0000    Distribution:                F(5,53
4)
Min Obs:                     10.0000
Max Obs:                     10.0000    F-statistic (robust):    7.045
1
                                P-value                    0.000
0
Time periods:                10    Distribution:                F(5,53
4)
Avg Obs:                     54.000
Min Obs:                     54.000
Max Obs:                     54.000

```

Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
Intercept	-0.6232	0.3373	-1.8476	0.0652	-1.2859	0.0394
CC	0.0746	0.0248	3.0025	0.0028	0.0258	0.1234
DE	0.0003	0.0028	0.0951	0.9243	-0.0053	0.0058
DF	1.866e-05	3.538e-05	0.5273	0.5982	-5.085e-05	8.816e-05
FFR	-0.0019	0.0039	-0.4854	0.6276	-0.0097	0.0058
TA	0.0114	0.0179	0.6357	0.5252	-0.0238	0.0466



```
In [25]: # compare fixed effects and random effects models
res3 = {'Fixed (Panel-Robust)':fixed_robust,'Random (Panel-Robust)':random_robust}
print(compare(res3))
```

Model Comparison		
	Fixed (Panel-Robust)	Random (Panel-Robust)
Dep. Variable	ROE	ROE
Estimator	PanelOLS	RandomEffects
No. Observations	540	540
Cov. Est.	Clustered	Clustered
R-squared	0.1099	0.0837
R-Squared (Within)	0.1099	0.0926
R-Squared (Between)	-0.7870	-0.0090
R-Squared (Overall)	-0.0770	0.0714
F-statistic	11.879	9.7559
P-value (F-stat)	0.0000	0.0000
Intercept	-0.7423 (-2.2216)	-0.6232 (-1.8476)
CC	0.0849 (3.4005)	0.0746 (3.0025)
DE	0.0101 (2.7018)	0.0003 (0.0951)
DF	-6.722e-07 (-0.0158)	1.866e-05 (0.5273)
FFR	-0.0022 (-0.5541)	-0.0019 (-0.4854)
TA	0.0136 (0.7713)	0.0114 (0.6357)
Effects	Entity	

T-stats reported in parentheses

In []: