

# Yang Qingyuan

June 21, 2021

## 1 import modules

```
[292]: import os
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from linearmodels.panel import PanelOLS
import statsmodels.formula.api as smf
import statsmodels.api as sm
from statsmodels.discrete.discrete_model import Probit
from sklearn.preprocessing import LabelEncoder
from statsmodels.stats.power import TTestIndPower
from scipy import optimize
```

## 2 global variables

```
[2]: #
main_path = r'D:\      '
os.chdir(main_path)

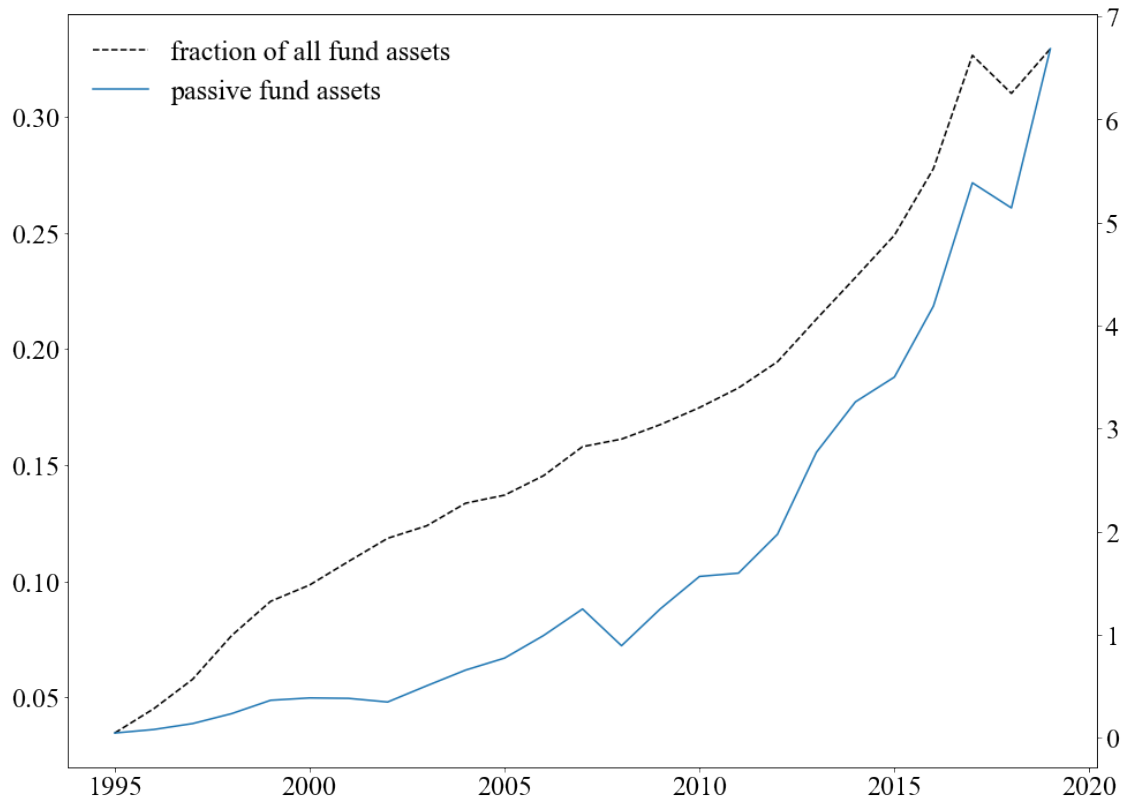
#
params={
    'font.family':'serif',
    'font.serif':'Times New Roman',
    'font.style':'normal', # or italic
    'font.weight':'normal', #or 'bold'
    'font.size': 24,#or large,small
}
plt.rcParams.update(params)
```

## 3 empirical study

### 3.1 Figure 1

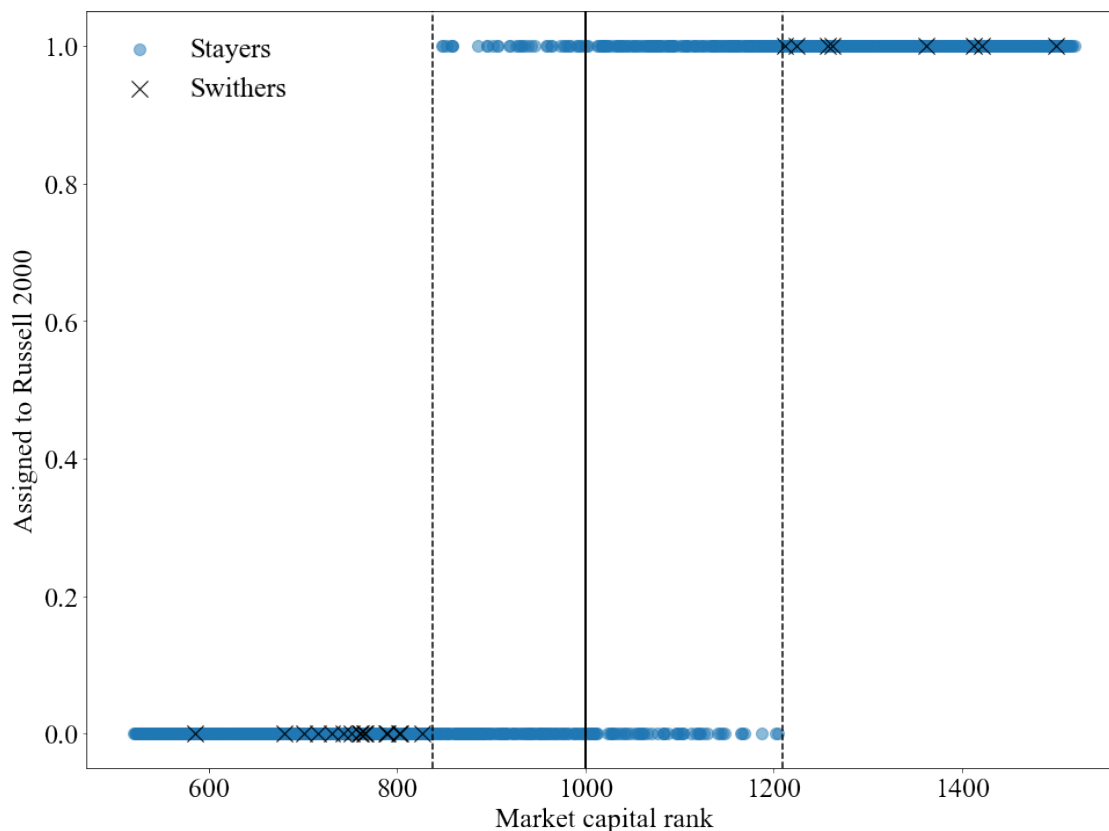
```
[177]: df = pd.read_stata('passiveaum.dta')

fig,ax = plt.subplots(figsize=(16,12))
ax2 = ax.twinx()
ax.plot(df['year'], df['frac_index'], '--', color='black', label='fraction of_
    ↳all fund assets')
ax2.plot(df['year'], df['aum1'], label='passive fund assets')
handles1, labels1 = ax.get_legend_handles_labels()
handles2, labels2 = ax2.get_legend_handles_labels()
plt.legend(handles1+handles2, labels1+labels2, framealpha=0)
plt.show()
fig.savefig('figure 1.png',dpi=1000, bbox_inches='tight')
```



### 3.2 Figure 2

```
[246]: # 2a
df = pd.read_stata('Russell107.dta')
fig,ax = plt.subplots(figsize=(16,12))
temp1 = df[df.switcher==0]
temp2 = df[df.switcher!=0]
ax.scatter(temp1['caprank'],temp1['R2000'], alpha=0.5, s=100, label='Stayers')
ax.scatter(temp2['caprank'],temp2['R2000'], marker='x', linewidth=1,
           color='black', s=200, label='Swithers')
plt.axvline(838, linestyle='--', color='black')
plt.axvline(1210, linestyle='--', color='black')
plt.axvline(1000, linestyle='-', linewidth=2, color='black')
ax.set_xlabel('Market capital rank')
ax.set_ylabel('Assigned to Russell 2000')
plt.legend(framealpha=0)
plt.show()
fig.savefig('figure 2a.png',dpi=1000, bbox_inches='tight')
```

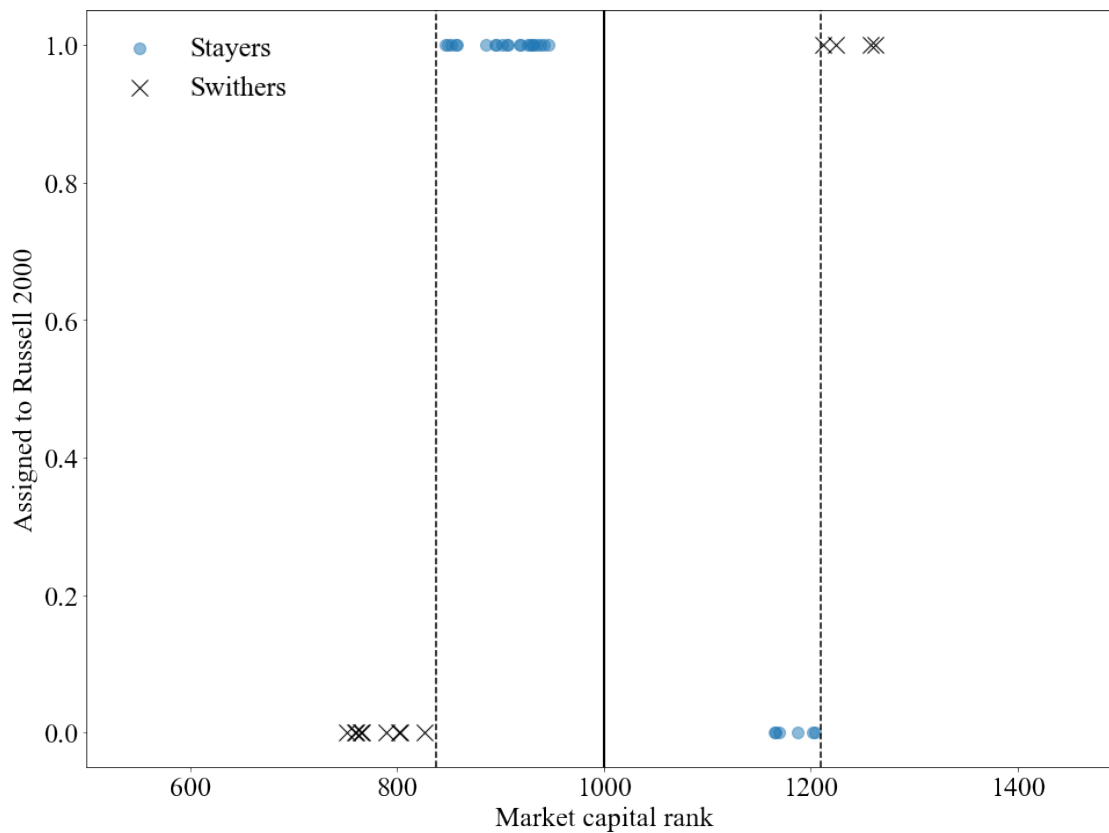


```
[260]: # 2b
df = pd.read_stata('Russell_sample_100.dta')
```

```

df = df[df.year==2007]
df = df[df.cohort==2007]
fig,ax = plt.subplots(figsize=(16,12))
temp1 = df[df.switcher==0]
temp2 = df[df.switcher!=0]
ax.scatter(temp1['caprank'],temp1['R2000'], alpha=0.5, s=100, label='Stayers')
ax.scatter(temp2['caprank'],temp2['R2000'], marker='x', linewidth=1,
    ↪color='black', s=200, label='Swithers')
plt.axvline(838, linestyle='--', color='black')
plt.axvline(1210, linestyle='--', color='black')
plt.axvline(1000, linestyle='-', linewidth=2, color='black')
ax.set_xlabel('Market capital rank')
ax.set_ylabel('Assigned to Russell 2000')
ax.set_xlim(500,1500)
plt.legend(framealpha=0, loc='upper left')
plt.show()
fig.savefig('figure 2b.png',dpi=1000, bbox_inches='tight')

```



### 3.3 Figure 3

```
[363]: df = pd.read_stata('Russell_sample_100.dta')
df = df[df.year==(df.cohort - 1)] # 1

# 3a
temp = df[df.upper==1].copy() #
temp['coarse_caprank'] = np.floor(temp.caprank_rel_to_upper / 10) * 10 + 5 #
↳
temp1 = temp[(temp.caprank_rel_to_upper<0) & (temp.caprank_rel_to_upper>=-100)]
temp2 = temp[(temp.caprank_rel_to_upper>0) & (temp.caprank_rel_to_upper<=100)]
fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='caprank_rel_to_upper', y='MFPassive', data=temp1, ci=99,
↳marker='', color='black')
sns.regplot(x='caprank_rel_to_upper', y='MFPassive', data=temp2, ci=99,
↳marker='', color='black')
temp = temp[(temp.coarse_caprank <= 100) & (temp.coarse_caprank >= -100)] #
↳
plt.axvline(0, linestyle='--', color='black')
ax.scatter(temp.groupby('coarse_caprank').mean().index, temp.
↳groupby('coarse_caprank').mean()['MFPassive'], color='black', s=100)
ax.set_ylim(0,15)
ax.set_xlabel('Rank Relative to Upper Band')
ax.set_ylabel('Passive Fund Ownership (% MktCap)')
plt.show()
fig.savefig('figure 3a.png',dpi=1000, bbox_inches='tight')

# 3b
temp = df[df.upper==0].copy() #
temp['coarse_caprank'] = np.floor(temp.caprank_rel_to_lower / 10) * 10 + 5 #
↳
temp1 = temp[(temp.caprank_rel_to_lower<0) & (temp.caprank_rel_to_lower>=-100)]
temp2 = temp[(temp.caprank_rel_to_lower>0) & (temp.caprank_rel_to_lower<=100)]
fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='caprank_rel_to_lower', y='MFPassive', data=temp1, ci=99,
↳marker='', color='black')
sns.regplot(x='caprank_rel_to_lower', y='MFPassive', data=temp2, ci=99,
↳marker='', color='black')
temp = temp[(temp.coarse_caprank <= 100) & (temp.coarse_caprank >= -100)] #
↳
plt.axvline(0, linestyle='--', color='black')
ax.scatter(temp.groupby('coarse_caprank').mean().index, temp.
↳groupby('coarse_caprank').mean()['MFPassive'], color='black', s=100)
ax.set_ylim(0,15)
ax.set_xlabel('Rank Relative to Lower Band')
ax.set_ylabel('Passive Fund Ownership (% MktCap)')
plt.show()
```

```

fig.savefig('figure 3b.png',dpi=1000, bbox_inches='tight')

# 3c E index
temp = df[df.upper==1].copy() #
temp['coarse_caprank'] = np.floor(temp.caprank_rel_to_upper / 10) * 10 + 5 #
↳
temp1 = temp[(temp.caprank_rel_to_upper<0) & (temp.caprank_rel_to_upper>=-100)]
temp2 = temp[(temp.caprank_rel_to_upper>0) & (temp.caprank_rel_to_upper<=100)]
fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='caprank_rel_to_upper', y='bcf', data=temp1, ci=99, marker='',
↳color='black')
sns.regplot(x='caprank_rel_to_upper', y='bcf', data=temp2, ci=99, marker='',
↳color='black')
temp = temp[(temp.coarse_caprank <= 100) & (temp.coarse_caprank >= -100)] #
↳
plt.axvline(0, linestyle='--', color='black')
ax.scatter(temp.groupby('coarse_caprank').mean().index, temp.
↳groupby('coarse_caprank').mean()['bcf'], color='black', s=100)
ax.set_ylim(0,6)
ax.set_xlabel('Rank Relative to Upper Band')
ax.set_ylabel('E-Index')
plt.show()
fig.savefig('figure 3c.png',dpi=1000, bbox_inches='tight')

# 3d E index
temp = df[df.upper==0].copy() #
temp['coarse_caprank'] = np.floor(temp.caprank_rel_to_lower / 10) * 10 + 5 #
↳
temp1 = temp[(temp.caprank_rel_to_lower<0) & (temp.caprank_rel_to_lower>=-100)]
temp2 = temp[(temp.caprank_rel_to_lower>0) & (temp.caprank_rel_to_lower<=100)]
fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='caprank_rel_to_lower', y='bcf', data=temp1, ci=99, marker='',
↳color='black')
sns.regplot(x='caprank_rel_to_lower', y='bcf', data=temp2, ci=99, marker='',
↳color='black')
temp = temp[(temp.coarse_caprank <= 100) & (temp.coarse_caprank >= -100)] #
↳
plt.axvline(0, linestyle='--', color='black')
ax.scatter(temp.groupby('coarse_caprank').mean().index, temp.
↳groupby('coarse_caprank').mean()['bcf'], color='black', s=100)
ax.set_ylim(0,6)
ax.set_xlabel('Rank Relative to Lower Band')
ax.set_ylabel('E-Index')
plt.show()
fig.savefig('figure 3d.png',dpi=1000, bbox_inches='tight')

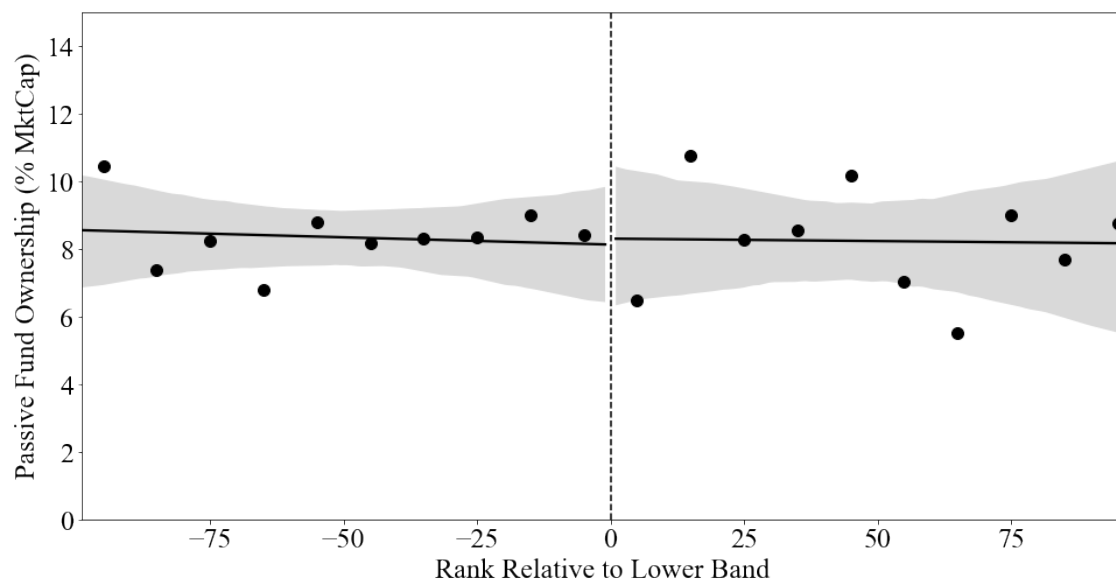
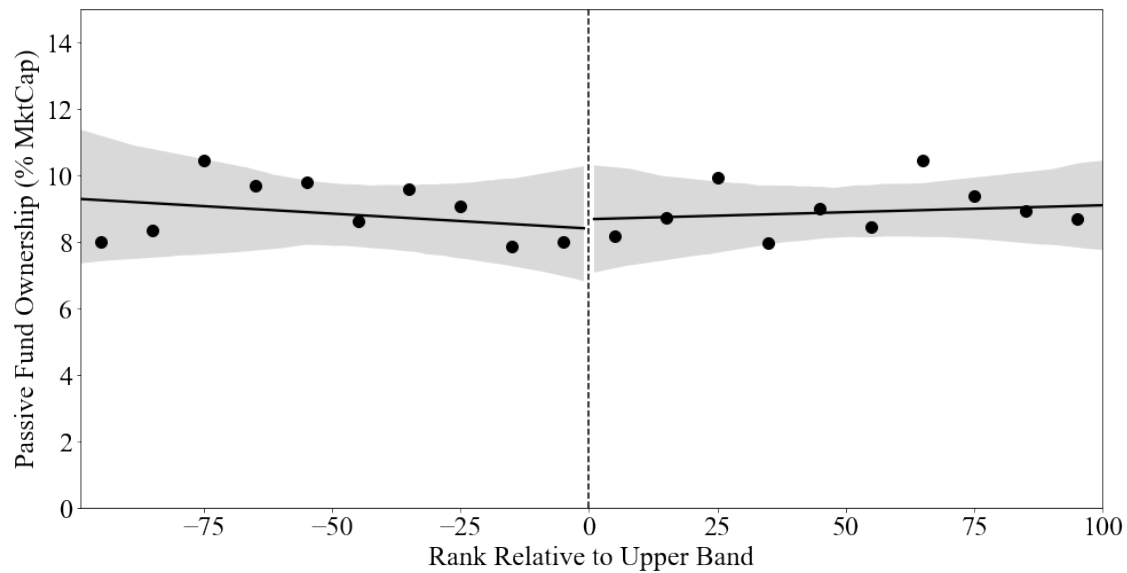
```

```

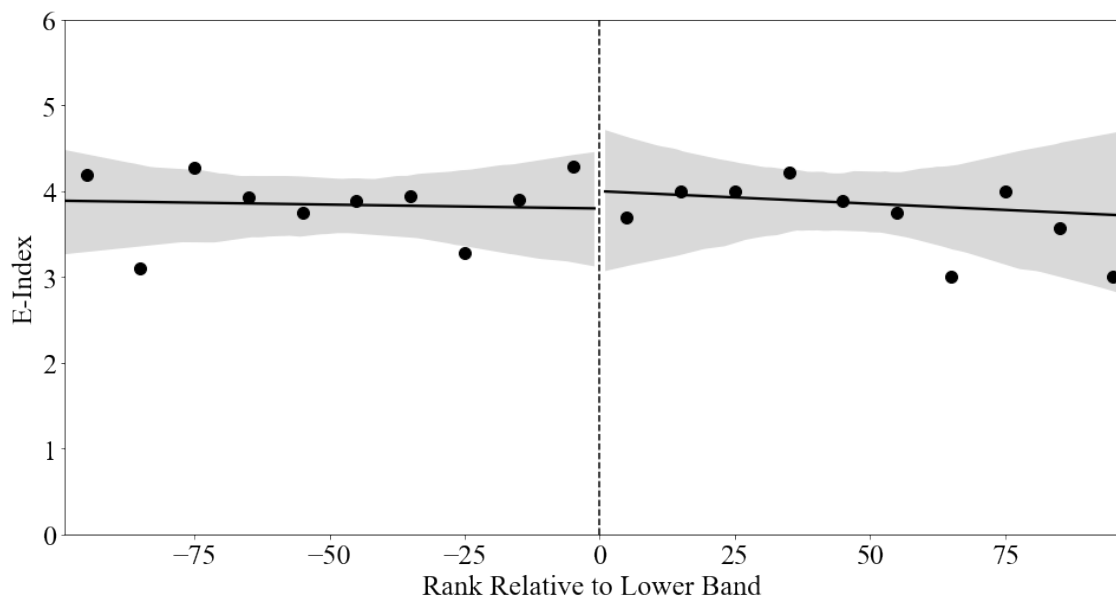
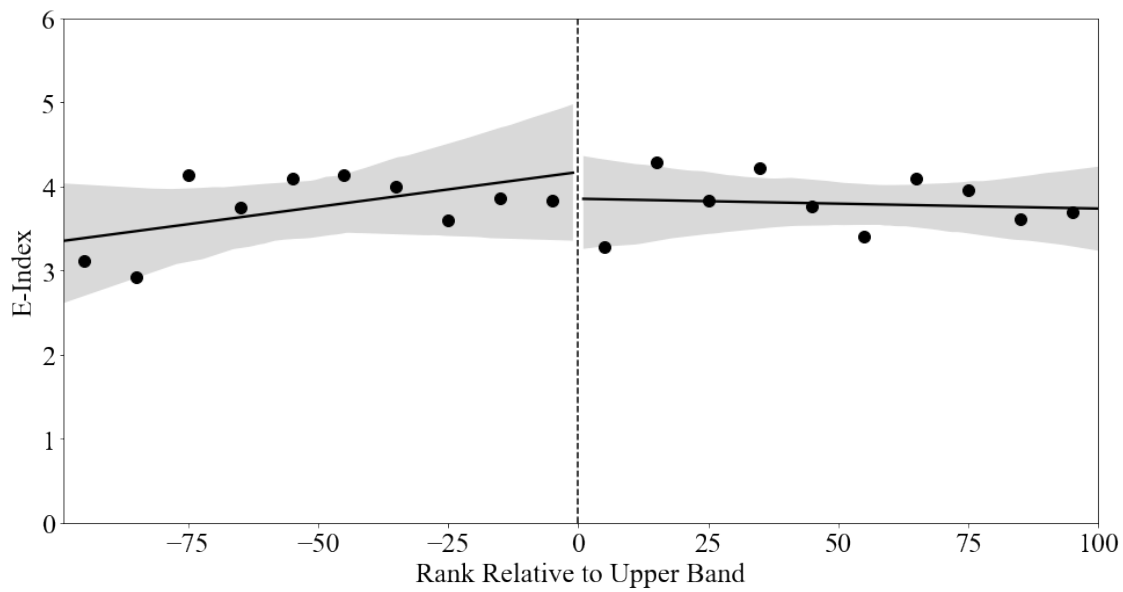
# 3e 11 Month Prior Return
temp = df[df.upper==1].copy() #
temp['coarse_caprank'] = np.floor(temp.caprank_rel_to_upper / 10) * 10 + 5 #
↳
temp1 = temp[(temp.caprank_rel_to_upper<0) & (temp.caprank_rel_to_upper>=-100)]
temp2 = temp[(temp.caprank_rel_to_upper>0) & (temp.caprank_rel_to_upper<=100)]
fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='caprank_rel_to_upper', y='rtn_julymay', data=temp1, ci=99,
↳marker='', color='black')
sns.regplot(x='caprank_rel_to_upper', y='rtn_julymay', data=temp2, ci=99,
↳marker='', color='black')
temp = temp[(temp.coarse_caprank <= 100) & (temp.coarse_caprank >= -100)] #
↳
plt.axvline(0, linestyle='--', color='black')
ax.scatter(temp.groupby('coarse_caprank').mean().index, temp.
↳groupby('coarse_caprank').mean()['rtn_julymay'], color='black', s=100)
ax.set_ylim(-0.4,0.4)
ax.set_xlabel('Rank Relative to Upper Band')
ax.set_ylabel('11 Month Prior Return')
plt.show()
fig.savefig('figure 3e.png',dpi=1000, bbox_inches='tight')

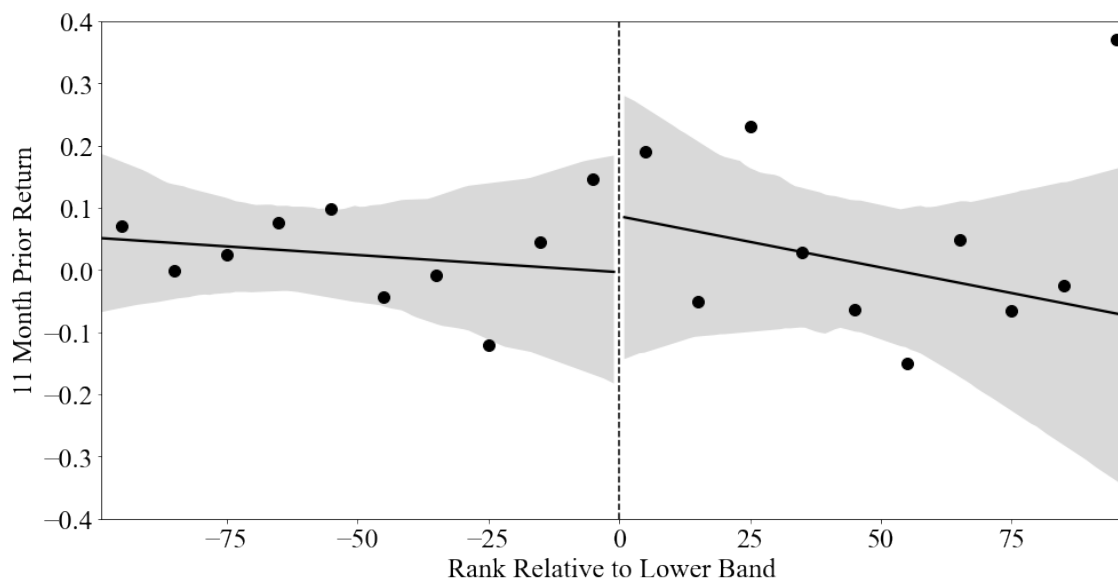
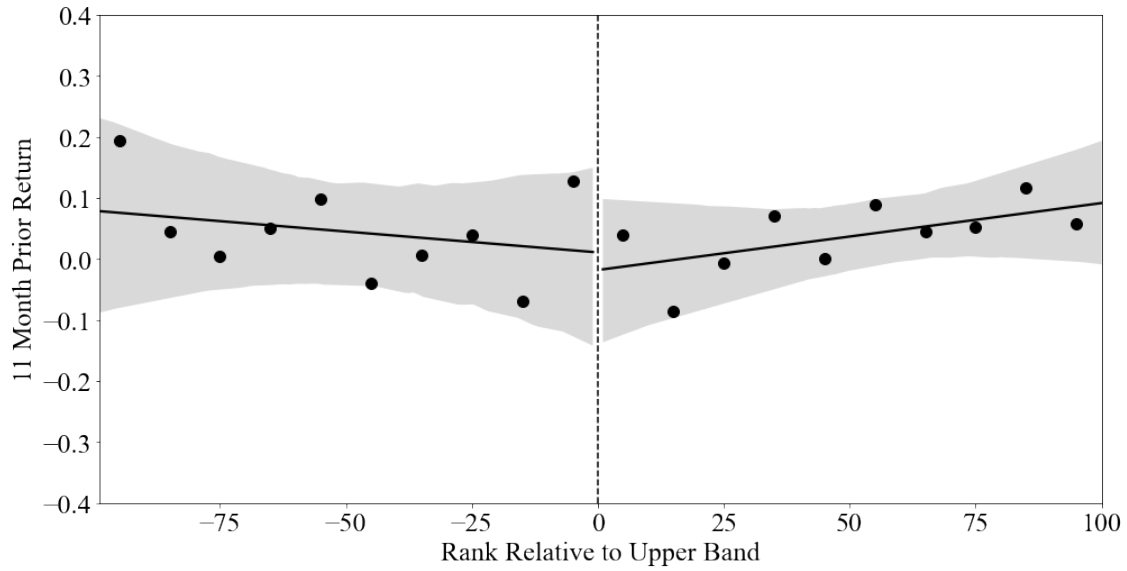
# 3f 11 Month Prior Return
temp = df[df.upper==0].copy() #
temp['coarse_caprank'] = np.floor(temp.caprank_rel_to_lower / 10) * 10 + 5 #
↳
temp1 = temp[(temp.caprank_rel_to_lower<0) & (temp.caprank_rel_to_lower>=-100)]
temp2 = temp[(temp.caprank_rel_to_lower>0) & (temp.caprank_rel_to_lower<=100)]
fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='caprank_rel_to_lower', y='rtn_julymay', data=temp1, ci=99,
↳marker='', color='black')
sns.regplot(x='caprank_rel_to_lower', y='rtn_julymay', data=temp2, ci=99,
↳marker='', color='black')
temp = temp[(temp.coarse_caprank <= 100) & (temp.coarse_caprank >= -100)] #
↳
plt.axvline(0, linestyle='--', color='black')
ax.scatter(temp.groupby('coarse_caprank').mean().index, temp.
↳groupby('coarse_caprank').mean()['rtn_julymay'], color='black', s=100)
ax.set_ylim(-0.4,0.4)
ax.set_xlabel('Rank Relative to Lower Band')
ax.set_ylabel('11 Month Prior Return')
plt.show()
fig.savefig('figure 3f.png',dpi=1000, bbox_inches='tight')

```









### 3.4 Figure 4

```
[371]: df = pd.read_stata('simulated_data.dta')

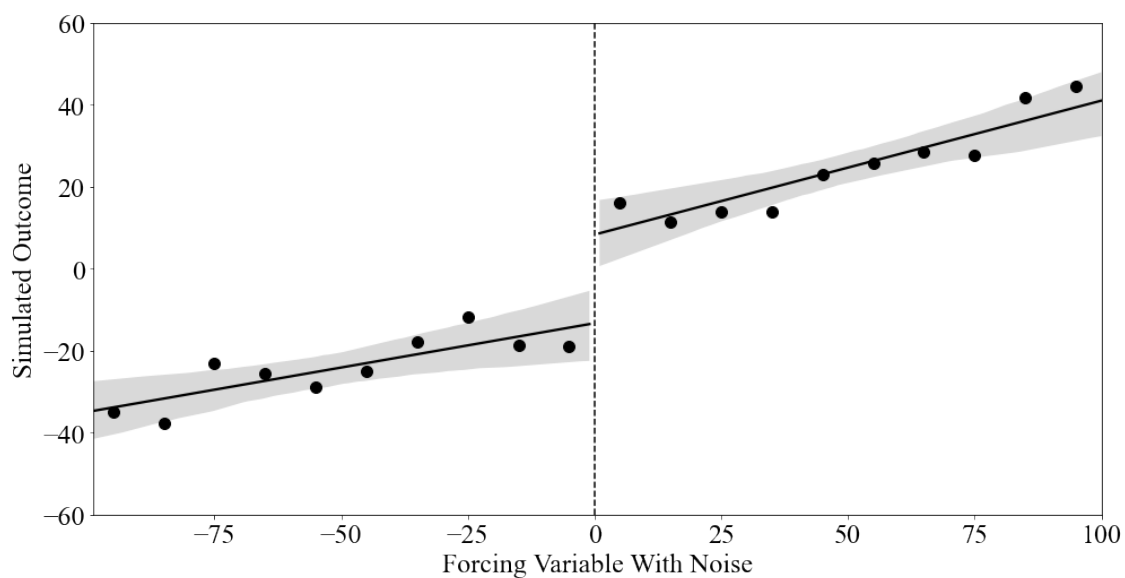
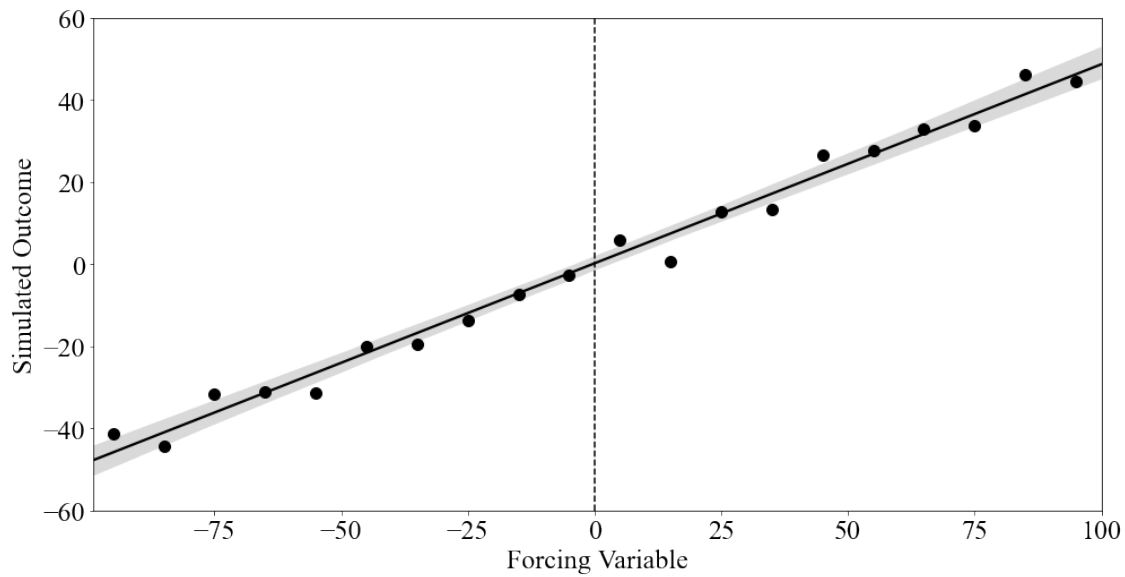
# 4a
temp = df.copy()
temp = temp[(temp.forcing <= 100) & (temp.forcing >= -100)]
temp['coarse_forcing'] = np.floor(temp.forcing / 10) * 10 + 5 #
```

```

fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='forcing', y='outcome', data=temp, ci=99, marker='',
            color='black')
temp = temp[(temp.coarse_forcing <= 100) & (temp.coarse_forcing >= -100)] #
ax.scatter(temp.groupby('coarse_forcing').mean().index, temp.
            groupby('coarse_forcing').mean()['outcome'], color='black', s=100)
plt.axvline(0, linestyle='--', color='black')
ax.set_xlabel('Forcing Variable')
ax.set_ylabel('Simulated Outcome')
ax.set_ylim(-60,60)
plt.show()
fig.savefig('figure 4a.png',dpi=1000, bbox_inches='tight')

# 4b
temp = df.copy()
temp1 = temp[(temp.forcing2<0) & (temp.forcing2>=-100)]
temp2 = temp[(temp.forcing2>0) & (temp.forcing2<=100)]
temp['coarse_forcing2'] = np.floor(temp.forcing2 / 10) * 10 + 5 #
fig,ax = plt.subplots(figsize=(16,8))
sns.regplot(x='forcing2', y='outcome', data=temp1, ci=99, marker='',
            color='black')
sns.regplot(x='forcing2', y='outcome', data=temp2, ci=99, marker='',
            color='black')
temp = temp[(temp.coarse_forcing2 <= 100) & (temp.coarse_forcing2 >= -100)] #
ax.scatter(temp.groupby('coarse_forcing2').mean().index, temp.
            groupby('coarse_forcing2').mean()['outcome'], color='black', s=100)
plt.axvline(0, linestyle='--', color='black')
ax.set_xlabel('Forcing Variable With Noise')
ax.set_ylabel('Simulated Outcome')
ax.set_ylim(-60,60)
plt.show()
fig.savefig('figure 4b.png',dpi=1000, bbox_inches='tight')

```



### 3.5 Figure 5

```
[59]: df = pd.read_stata('Russell_sample_100.dta')
df['rel_year'] = df['year'] - df['cohort']
df = df.groupby(['upper', 'switcher', 'rel_year']).
    ↪agg(['mean', 'sem'])['passivefund_own_R2000']
df = df.reset_index()
# upper = 1
```

```

temp = df[df['upper']==1].copy()
temp['switcher'] = temp['switcher'].replace(-1,1)
temp.drop(columns=['upper'], inplace=True)
temp = temp.set_index(['rel_year', 'switcher']).unstack().
    ↳sort_values(by='switcher',axis=1)
temp['upper'] = 1
# upper = 0
temp1 = df[df['upper']==0].copy()
temp1['switcher'] = temp1['switcher'].replace(-1,1)
temp1.drop(columns=['upper'], inplace=True)
temp1 = temp1.set_index(['rel_year', 'switcher']).unstack().
    ↳sort_values(by='switcher',axis=1)
temp1['upper'] = 0

temp = temp.append(temp1).sort_values(by='upper')
temp.columns = ['mean0', 'sem0', 'mean1', 'sem1', 'upper']
temp['lb0'] = temp['mean0'] - 2 * temp['sem0']
temp['ub0'] = temp['mean0'] + 2 * temp['sem0']
temp['lb1'] = temp['mean1'] - 2 * temp['sem1']
temp['ub1'] = temp['mean1'] + 2 * temp['sem1']
temp['lower'] = temp['upper']==0
temp.loc[temp.lower==True, 'lower2'] = "Lower Band (R1000 {\&rarr} R2000)"
temp.loc[temp.lower==False, 'lower2'] = "Lower Band (R2000 {\&rarr} R1000)"
temp

```

```

[59]:

```

	mean0	sem0	mean1	sem1	upper	lb0	ub0	\
rel_year								
-3	0.156030	0.035166	0.141302	0.051118	0	0.085697	0.226362	
-2	0.066345	0.024558	0.045150	0.028412	0	0.017230	0.115460	
-1	0.000000	0.000000	0.123961	0.054510	0	0.000000	0.000000	
0	0.154263	0.043493	2.099414	0.065674	0	0.067278	0.241248	
1	1.042622	0.087734	2.142434	0.090097	0	0.867154	1.218089	
2	1.298348	0.095086	2.202878	0.117981	0	1.108175	1.488520	
-3	1.467304	0.044630	1.427165	0.060834	1	1.378045	1.556564	
-2	1.581118	0.043268	1.656058	0.054325	1	1.494582	1.667655	
-1	1.787587	0.039851	1.693187	0.048941	1	1.707884	1.867289	
0	1.851963	0.047595	0.000002	0.000002	1	1.756772	1.947153	
1	1.290208	0.069332	0.037541	0.019472	1	1.151545	1.428872	
2	1.053380	0.075274	0.181435	0.045752	1	0.902832	1.203929	

	lb1	ub1	lower	lower2
rel_year				
-3	0.039066	0.243538	True	Lower Band (R1000 {\&rarr} R2000)
-2	-0.011674	0.101974	True	Lower Band (R1000 {\&rarr} R2000)
-1	0.014941	0.232981	True	Lower Band (R1000 {\&rarr} R2000)
0	1.968066	2.230761	True	Lower Band (R1000 {\&rarr} R2000)
1	1.962241	2.322627	True	Lower Band (R1000 {\&rarr} R2000)

2	1.966915	2.438841	True	Lower Band (R1000 $\{\&rarr\}$ R2000)
-3	1.305497	1.548833	False	Lower Band (R2000 $\{\&rarr\}$ R1000)
-2	1.547408	1.764708	False	Lower Band (R2000 $\{\&rarr\}$ R1000)
-1	1.595305	1.791070	False	Lower Band (R2000 $\{\&rarr\}$ R1000)
0	-0.000002	0.000006	False	Lower Band (R2000 $\{\&rarr\}$ R1000)
1	-0.001403	0.076486	False	Lower Band (R2000 $\{\&rarr\}$ R1000)
2	0.089931	0.272939	False	Lower Band (R2000 $\{\&rarr\}$ R1000)

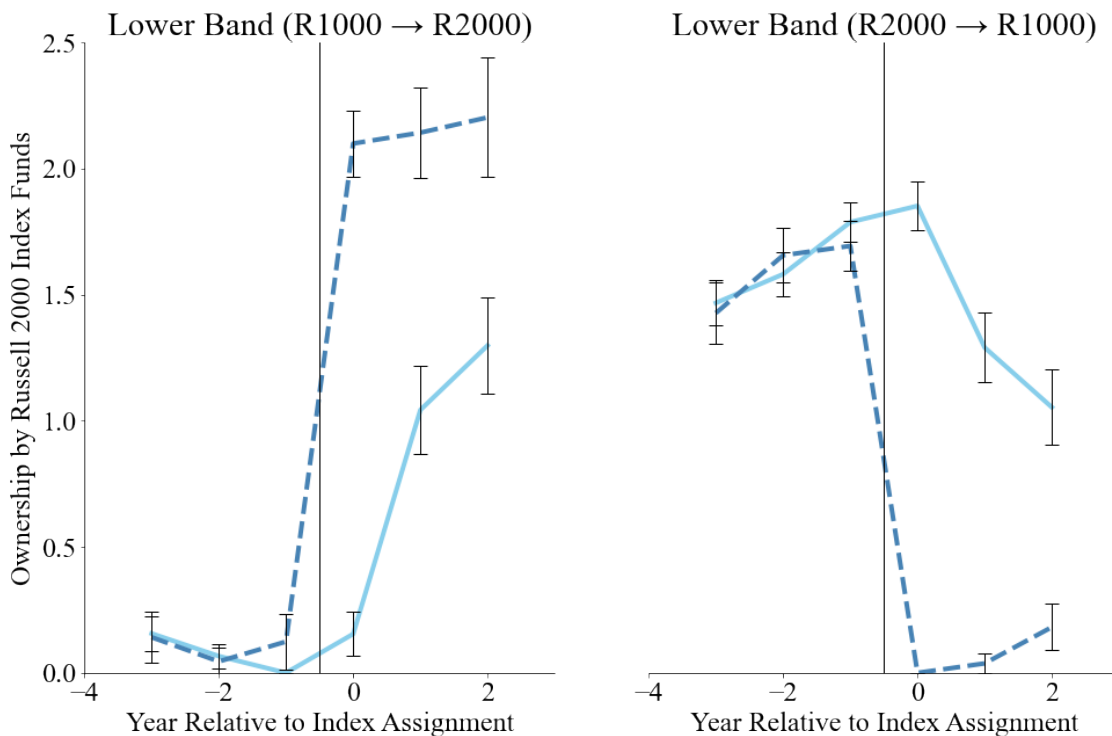
```
[151]: fig,ax = plt.subplots(1,2,figsize=(16,10))
#
temp1 = temp[temp.lower].copy()
ax[0].plot(temp1.reset_index().rel_year, temp1['mean0'], linewidth=4,
           color='skyblue')
ax[0].plot(temp1.reset_index().rel_year, temp1['mean1'], '--', linewidth=4,
           color='steelblue')
ax[0].set_xlim(-4,3)
ax[0].set_ylim(0,2.5)
ax[0].spines['right'].set_visible(False)
ax[0].spines['top'].set_visible(False)
ax[0].set_title("Lower Band (R1000  $\rightarrow$  R2000)")
ax[0].set_xlabel('Year Relative to Index Assignment')
ax[0].set_ylabel('Ownership by Russell 2000 Index Funds')
ax[0].axvline(-0.5, color='black', linewidth=1)
for i in temp1.index: # stayer
    ax[0].axvline(i, color='black', linewidth=1, marker='_', markersize=12,
                 ymin=temp1.loc[i,'lb0']/2.5, ymax=temp1.loc[i,'ub0']/2.5)
for i in temp1.index: # switcher
    ax[0].axvline(i, color='black', linewidth=1, marker='_', markersize=12,
                 ymin=temp1.loc[i,'lb1']/2.5, ymax=temp1.loc[i,'ub1']/2.5)
#
temp2 = temp[~temp.lower]
ax[1].plot(temp2.reset_index().rel_year, temp2['mean0'], linewidth=4,
           color='skyblue')
ax[1].plot(temp2.reset_index().rel_year, temp2['mean1'], '--', linewidth=4,
           color='steelblue')
ax[1].set_xlim(-4,3)
ax[1].set_ylim(0,2.5)
plt.gca().axes.get_yaxis().set_visible(False) # y
ax[1].spines['right'].set_visible(False)
ax[1].spines['left'].set_visible(False)
ax[1].spines['top'].set_visible(False)
ax[1].axvline(-0.5, color='black', linewidth=1)
for i in temp2.index: # stayer
    ax[1].axvline(i, color='black', linewidth=1, marker='_', markersize=12,
                 ymin=temp2.loc[i,'lb0']/2.5, ymax=temp2.loc[i,'ub0']/2.5)
for i in temp2.index: # switcher
```

```

ax[1].axvline(i, color='black', linewidth=1, marker='_', markersize=12,
↳ymin=temp2.loc[i,'lb1']/2.5, ymax=temp2.loc[i,'ub1']/2.5)

ax[1].set_title("Lower Band (R2000 → R1000)")
ax[1].set_xlabel('Year Relative to Index Assignment')
plt.show()
fig.savefig('figure 5.png',dpi=1000, bbox_inches='tight')

```



### 3.6 Figure 6

```

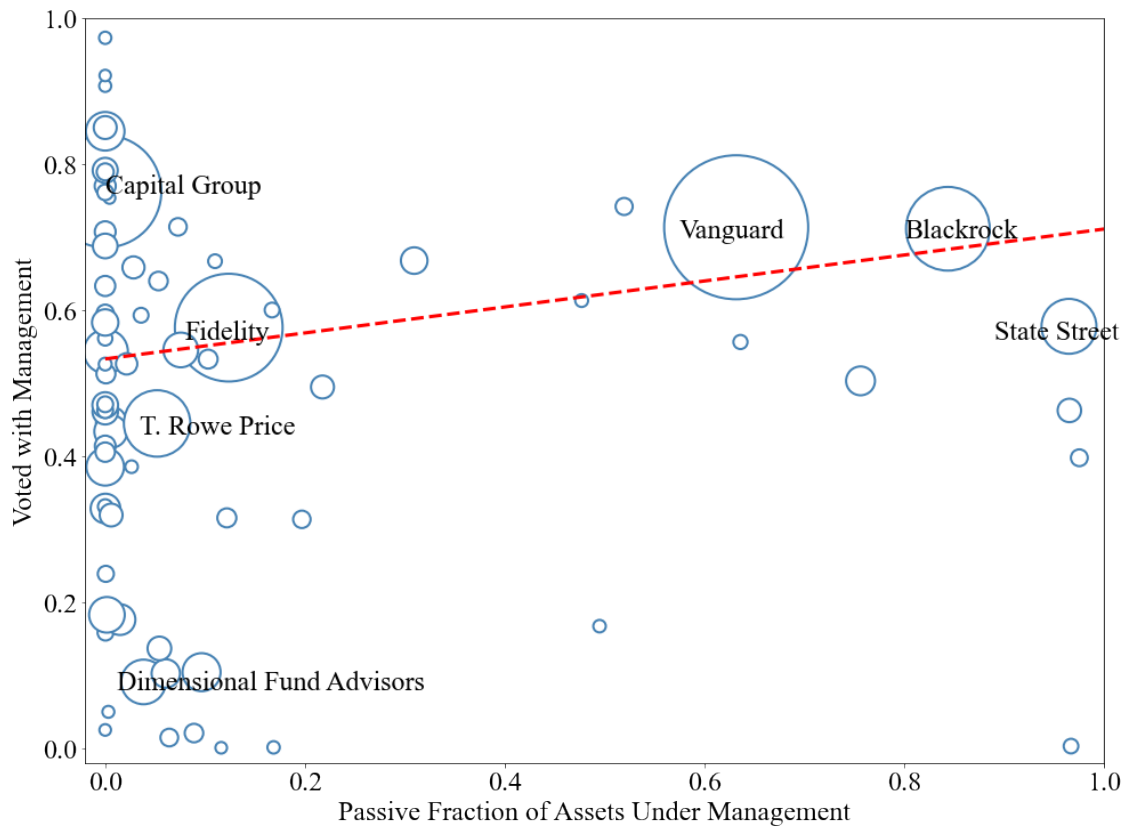
[149]: df = pd.read_stata('fundfamilies.dta')
fig,ax = plt.subplots(figsize=(16,12))
ax.scatter(df['frac_passive'], df['voted_with_mgmt'], s=df['tna_total']/100,
↳marker='o', color='white', edgecolor='steelblue',linewidth=2)
ax.text(0.575,0.7,"Vanguard", ha='left')
ax.text(0.8,0.7,"Blackrock", ha='left')
ax.text(0.89,0.56, "State Street", ha='left')
ax.text(0.08,0.56, "Fidelity")
ax.text(0.012,0.08, "Dimensional Fund Advisors")
ax.text(0,0.76, "Capital Group")
ax.text(0.035,0.43, "T. Rowe Price")
mod = smf.wls(formula='voted_with_mgmt ~ frac_passive', weights=df.tna_total,
↳data=df) # WLS

```

```

temp = np.linspace(0,100,100)
temp = pd.DataFrame(temp, columns=['x'])
temp['y'] = mod.fit().params[0] + temp['x'] * mod.fit().params[1]
ax.plot(temp['x'], temp['y'], '--', color='red', linewidth=3)
ax.set_xlim(-0.02,1)
ax.set_ylim(-0.02,1)
ax.set_xlabel('Passive Fraction of Assets Under Management')
ax.set_ylabel('Voted with Management')
plt.show()
fig.savefig('figure 6.png',dpi=1000, bbox_inches='tight')

```



### 3.7 Table 1

```

[131]: # AUM expense ratio
df = pd.read_stata('fundyear_panel.dta')
temp = df[['tna_latest', 'exp_ratio', 'indexfund']].groupby('indexfund').
    ↳describe(percentiles=[.1,.5,.9]) # indexfund
#temp2 = df[['tna_latest', 'exp_ratio']].describe(percentiles=[.1,.5,.9]) #
#temp2.columns = [0,1] # stata 0,1
table1 = (temp['tna_latest'].round(0)).append(temp['exp_ratio'].round(2)) #
    ↳latest 0 exp_ratio 2

```



```

table1.index.name = 'indexfund'

#
df = pd.read_stata('fundholdings_byyear.dta')
temp1 = df[['nholdings', 'indexfund']].groupby('indexfund').
    ↳describe(percentiles=[.1, .5, .9])
table1 = table1.append(temp1['nholdings'].round(0)).
    ↳sort_values(by='indexfund', ascending=False)
table1['variable'] = ['AUM ($M)', 'Expense ratio (%)', '# stocks hold', 'AUM_
    ↳($M)', 'Expense ratio (%)', '# stocks hold']
table1 = table1[['variable', 'mean', 'std', '10%', '50%', '90%', 'count']]

# indexfund activefund
df = pd.read_stata('fundyear_panel.dta')
print('index fund {} active fund {}'.format(len(df[df.
    ↳indexfund==1]['fundid_anon'].drop_duplicates()), len(df[df.
    ↳activefund==1]['fundid_anon'].drop_duplicates())))
table1.to_excel('table1.xlsx')

```

index fund 613 active fund 2646

### 3.8 Table 2

```

[13]: df = pd.read_stata('Russell_sample_100.dta')
temp = df.describe(percentiles=[.1, .5, .9])
temp =
    ↳temp[['ME', 'passivefund_own_R2000', 'passivefund_own_R1000', 'MFPassive', 'MFActive', 'delta', '
        ↳
        ↳'bcf', 'bcf5', 'supermajority_req', 'limitspecialmeet', 'writtenconsent', 'dualclass']]
temp.T[['mean', 'std', '10%', '50%', '90%', 'count']].to_excel('table 2.xlsx')

```

### 3.9 Table 3

```

[68]: def table3_reg(y_name):
    df = pd.read_stata('Russell_sample_100.dta')
    df['firmid_anon#cohort'] = df['firmid_anon'] * df['cohort']
    df.set_index(['firmid_anon#cohort', df.year], inplace=True) #
    X = df[['R1000_to_R2000', 'R2000_to_R1000']]
    y = df[[y_name]]
    mod = PanelOLS(y, X, entity_effects=True, time_effects=True,
    ↳singletons=False)
    res = mod.fit(cov_type='clustered', clusters=df[['firmid_anon', 'year']])
    return res

```

```

[70]: # index own 2000
table3_reg('passivefund_own_R2000')

```

```
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 1 singleton observations dropped
warn.warn(
```

[70]:

#### PanelOLS Estimation Summary

```
=====
===
Dep. Variable:    passivefund_own_R2000    R-squared:
0.3209
Estimator:                PanelOLS    R-squared (Between):
-0.2171
No. Observations:                4649    R-squared (Within):
0.2977
Date:                Sat, Jun 12 2021    R-squared (Overall):
-0.0607
Time:                16:24:49    Log-likelihood
-4888.1
Cov. Estimator:                Clustered
F-statistic:
900.08
Entities:                824    P-value
0.0000
Avg Obs:                5.6420    Distribution:
F(2,3809)
Min Obs:                2.0000
Max Obs:                6.0000    F-statistic (robust):
186.28
P-value
0.0000
Time periods:                15    Distribution:
F(2,3809)
Avg Obs:                309.93
Min Obs:                39.000
Max Obs:                504.00
```

#### Parameter Estimates

```
=====
==
CI
-----
--
R1000_to_R2000    1.7181    0.1453    11.824    0.0000    1.4332
2.0030
R2000_to_R1000    -1.6300    0.0888    -18.348    0.0000    -1.8042
-1.4559
=====
```

==

F-test for Poolability: 5.4117

P-value: 0.0000

Distribution: F(837,3809)

Included effects: Entity, Time

PanelEffectsResults, id: 0x16c1aa532b0

```
[71]: # index own 1000
      table3_reg('passivefund_own_R1000')
```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 1 singleton observations dropped

warn.warn()

[71]: PanelOLS Estimation Summary

=====

===

Dep. Variable:	passivefund_own_R1000	R-squared:
	0.3387	
Estimator:	PanelOLS	R-squared (Between):
	0.0012	
No. Observations:	4649	R-squared (Within):
	0.3525	
Date:	Sat, Jun 12 2021	R-squared (Overall):
	0.1489	
Time:	16:25:04	Log-likelihood
	4728.6	
Cov. Estimator:	Clustered	F-statistic:
		975.24
Entities:	824	P-value
		0.0000
Avg Obs:	5.6420	Distribution:
		F(2,3809)
Min Obs:	2.0000	
Max Obs:	6.0000	F-statistic (robust):
		121.53
		P-value
		0.0000
Time periods:	15	Distribution:
		F(2,3809)
Avg Obs:	309.93	
Min Obs:	39.000	
Max Obs:	504.00	

Parameter Estimates

```
=====
```

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper
CI						
--						
R1000_to_R2000	-0.2141	0.0216	-9.9313	0.0000	-0.2564	
-0.1719						
R2000_to_R1000	0.2217	0.0143	15.538	0.0000	0.1937	
0.2497						

```
=====
```

F-test for Poolability: 5.0986  
P-value: 0.0000  
Distribution: F(837,3809)

Included effects: Entity, Time  
PanelEffectsResults, id: 0x16c1f9b8d60

```
[72]: # index own mfpactive
table3_reg('MFPactive')
```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:  
SingletonWarning: 1 singleton observations dropped  
warn.warn(

```
[72]:
```

PanelOLS Estimation Summary			
=====			
Dep. Variable:	MFPactive	R-squared:	0.0299
Estimator:	PanelOLS	R-squared (Between):	-0.0102
No. Observations:	4649	R-squared (Within):	-0.0106
Date:	Sat, Jun 12 2021	R-squared (Overall):	-0.0104
Time:	16:25:36	Log-likelihood	-9884.7
Cov. Estimator:	Clustered		
		F-statistic:	58.798
Entities:	824	P-value	0.0000
Avg Obs:	5.6420	Distribution:	F(2,3809)
Min Obs:	2.0000		
Max Obs:	6.0000	F-statistic (robust):	17.265
		P-value	0.0000
Time periods:	15	Distribution:	F(2,3809)
Avg Obs:	309.93		
Min Obs:	39.000		
Max Obs:	504.00		

Parameter Estimates

```
=====
```

```

==
                Parameter  Std. Err.      T-stat      P-value      Lower CI      Upper
CI
-----
--
R1000_to_R2000      1.3119      0.3677      3.5678      0.0004      0.5910
2.0328
R2000_to_R1000     -1.2041      0.2599     -4.6330      0.0000     -1.7136
-0.6945
=====
==

```

F-test for Poolability: 29.718

P-value: 0.0000

Distribution: F(837,3809)

Included effects: Entity, Time

PanelEffectsResults, id: 0x16c1f9f8ac0

```

[73]: # index own mfpassive
      table3_reg('MFAActive')

```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 1 singleton observations dropped

warn.warn()

```

[73]:                PanelOLS Estimation Summary
=====
Dep. Variable:                MFAActive      R-squared:                0.0077
Estimator:                    PanelOLS       R-squared (Between):      0.0041
No. Observations:              4649          R-squared (Within):       0.0098
Date:                          Sat, Jun 12 2021  R-squared (Overall):      0.0045
Time:                          16:25:47        Log-likelihood            -1.492e+04
Cov. Estimator:                Clustered
                                F-statistic:                14.832
Entities:                      824             P-value                   0.0000
Avg Obs:                       5.6420          Distribution:              F(2,3809)
Min Obs:                       2.0000
Max Obs:                       6.0000          F-statistic (robust):     7.1720
                                P-value                   0.0008
Time periods:                  15             Distribution:              F(2,3809)
Avg Obs:                       309.93
Min Obs:                       39.000
Max Obs:                       504.00

```

Parameter Estimates

```

=====
==

```

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper
CI						
-----						
--						
R1000_to_R2000	-2.2053	0.7960	-2.7706	0.0056	-3.7659	
-0.6447						
R2000_to_R1000	1.6008	0.6064	2.6398	0.0083	0.4119	
2.7896						
=====						
==						

F-test for Poolability: 14.162

P-value: 0.0000

Distribution: F(837,3809)

Included effects: Entity, Time

PanelEffectsResults, id: 0x16c1f9cbb50

```
[74]: # index own mfpassive
      table3_reg('allother')
```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 1 singleton observations dropped

warn.warn()

[74]: PanelOLS Estimation Summary

Dep. Variable:	allother	R-squared:	0.0007
Estimator:	PanelOLS	R-squared (Between):	0.0002
No. Observations:	4649	R-squared (Within):	-1.307e-05
Date:	Sat, Jun 12 2021	R-squared (Overall):	0.0002
Time:	16:25:57	Log-likelihood	-1.553e+04
Cov. Estimator:	Clustered		
		F-statistic:	1.2844
Entities:	824	P-value	0.2769
Avg Obs:	5.6420	Distribution:	F(2,3809)
Min Obs:	2.0000		
Max Obs:	6.0000	F-statistic (robust):	0.6447
		P-value	0.5249
Time periods:	15	Distribution:	F(2,3809)
Avg Obs:	309.93		
Min Obs:	39.000		
Max Obs:	504.00		

#### Parameter Estimates

=====						
==						
	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper

CI

```
-----  
--  
R1000_to_R2000      0.8934      0.9433      0.9471      0.3436      -0.9560  
2.7428  
R2000_to_R1000     -0.3967      0.7169     -0.5534      0.5801      -1.8023  
1.0089  
=====
```

F-test for Poolability: 16.442

P-value: 0.0000

Distribution: F(837,3809)

Included effects: Entity, Time

PanelEffectsResults, id: 0x16c1aa66430

### 3.10 Table 4

```
[43]: df = pd.read_stata('fundvotes.dta')  
df = df[['votedyes', 'votedno', 'abstained', 'didnotvote', 'indexfund',  
        ↪ 'mgmt_against', 'iss_against']]  
indexdf = df[df.indexfund==1].copy() # index fund  
activedf = df[df.indexfund==0].copy() # active fund
```

```
[44]: ##  
#  
tabledf = pd.DataFrame()  
temp1 = pd.DataFrame(indexdf.sum(), columns=['index fund'])  
temp2 = pd.DataFrame(activedf.sum(), columns=['active fund'])  
temp = pd.concat([temp1, temp2], axis=1)  
temp = temp.loc[['votedyes', 'votedno', 'abstained', 'didnotvote']]  
temp = temp / temp.sum()  
temp = pd.DataFrame(temp.unstack(), columns=['all']).T  
temp['Difference PctYes'] = temp['index fund']['votedyes'] - temp['active_↪  
fund']['votedyes']  
temp['N'] = df.count()[0]  
tabledf = tabledf.append(temp)  
tabledf
```

```
[44]:      index fund      active fund  
      votedyes  votedno  abstained  didnotvote  votedyes  votedno  abstained  \  
all  0.898148  0.066869  0.032586  0.002397  0.88877  0.075739  0.03122  
  
      Difference PctYes      N  
      didnotvote  
all  0.004271      0.009378  27297366
```

```
[45]: # yes yes
temp1 = pd.DataFrame(indexdf[(indexdf.mgmt_against==0) & (indexdf.
    ↳iss_against==0)].sum(),columns=['index fund'])
temp2 = pd.DataFrame(activatedf[(activatedf.mgmt_against==0) & (activatedf.
    ↳iss_against==0)].sum(),columns=['active fund'])
temp = pd.concat([temp1,temp2],axis=1)
temp = temp.loc[['votedyes', 'votedno', 'abstained', 'didnotvote']]
temp = temp / temp.sum()
temp = pd.DataFrame(temp.unstack(), columns=['yes yes']).T
temp['Difference PctYes'] = temp['index fund']['votedyes'] - temp['active_
    ↳fund']['votedyes']
temp['N'] = indexdf[(indexdf.mgmt_against==0) & (indexdf.iss_against==0)].
    ↳count()[0] + activatedf[(activatedf.mgmt_against==0) & (activatedf.
    ↳iss_against==0)].count()[0]
tabledf = tabledf.append(temp)
tabledf
```

```
[45]:
```

	index fund				active fund		
	votedyes	votedno	abstained	didnotvote	votedyes	votedno	\
all	0.898148	0.066869	0.032586	0.002397	0.888770	0.075739	
yes yes	0.948730	0.036646	0.013684	0.000940	0.952452	0.033084	

	Difference PctYes				N
	abstained	didnotvote			
all	0.031220	0.004271	0.009378	27297366	
yes yes	0.011785	0.002679	-0.003723	24293163	

```
[46]: # yes yes
temp1 = pd.DataFrame(indexdf[(indexdf.mgmt_against==1) & (indexdf.
    ↳iss_against==1)].sum(),columns=['index fund'])
temp2 = pd.DataFrame(activatedf[(activatedf.mgmt_against==1) & (activatedf.
    ↳iss_against==1)].sum(),columns=['active fund'])
temp = pd.concat([temp1,temp2],axis=1)
temp = temp.loc[['votedyes', 'votedno', 'abstained', 'didnotvote']]
temp = temp / temp.sum()
temp = pd.DataFrame(temp.unstack(), columns=['no no']).T
temp['Difference PctYes'] = temp['index fund']['votedyes'] - temp['active_
    ↳fund']['votedyes']
temp['N'] = indexdf[(indexdf.mgmt_against==1) & (indexdf.iss_against==1)].
    ↳count()[0] + activatedf[(activatedf.mgmt_against==1) & (activatedf.
    ↳iss_against==1)].count()[0]
tabledf = tabledf.append(temp)
tabledf
```

```
[46]:
```

	index fund				active fund		
	votedyes	votedno	abstained	didnotvote	votedyes	votedno	\
all	0.898148	0.066869	0.032586	0.002397	0.888770	0.075739	



yes yes	0.948730	0.036646	0.013684	0.000940	0.952452	0.033084
no no	0.050827	0.826508	0.099179	0.023486	0.053712	0.829371

			Difference PctYes	N
	abstained	didnotvote		
all	0.031220	0.004271	0.009378	27297366
yes yes	0.011785	0.002679	-0.003723	24293163
no no	0.101189	0.015728	-0.002885	398666

```
[47]: # yes no
temp1 = pd.DataFrame(indexdf[(indexdf.mgmt_against==0) & (indexdf.
    ↳iss_against==1)].sum(),columns=['index fund'])
temp2 = pd.DataFrame(activatedf[(activatedf.mgmt_against==0) & (activatedf.
    ↳iss_against==1)].sum(),columns=['active fund'])
temp = pd.concat([temp1,temp2],axis=1)
temp = temp.loc[['votedyes', 'votedno', 'abstained', 'didnotvote']]
temp = temp / temp.sum()
temp = pd.DataFrame(temp.unstack(), columns=['yes no']).T
temp['Difference PctYes'] = temp['index fund']['votedyes'] - temp['active_
    ↳fund']['votedyes']
temp['N'] = indexdf[(indexdf.mgmt_against==0) & (indexdf.iss_against==1)].
    ↳count()[0] + activatedf[(activatedf.mgmt_against==0) & (activatedf.
    ↳iss_against==1)].count()[0]
tabledf = tabledf.append(temp)
tabledf
```

```
[47]: index fund active fund \
      votedyes votedno abstained didnotvote votedyes votedno
all      0.898148 0.066869 0.032586 0.002397 0.888770 0.075739
yes yes   0.948730 0.036646 0.013684 0.000940 0.952452 0.033084
no no     0.050827 0.826508 0.099179 0.023486 0.053712 0.829371
yes no    0.533978 0.200873 0.247289 0.017860 0.438823 0.248889

      Difference PctYes N
      abstained didnotvote
all      0.031220 0.004271 0.009378 27297366
yes yes   0.011785 0.002679 -0.003723 24293163
no no     0.101189 0.015728 -0.002885 398666
yes no    0.286819 0.025469 0.095156 1761341
```

```
[48]: # no yes
temp1 = pd.DataFrame(indexdf[(indexdf.mgmt_against==1) & (indexdf.
    ↳iss_against==0)].sum(),columns=['index fund'])
temp2 = pd.DataFrame(activatedf[(activatedf.mgmt_against==1) & (activatedf.
    ↳iss_against==0)].sum(),columns=['active fund'])
temp = pd.concat([temp1,temp2],axis=1)
temp = temp.loc[['votedyes', 'votedno', 'abstained', 'didnotvote']]
```

```
temp = temp / temp.sum()
temp = pd.DataFrame(temp.unstack(), columns=['no yes']).T
temp['Difference PctYes'] = temp['index fund']['votedyes'] - temp['active_fund']['votedyes']
temp['N'] = indexdf[(indexdf.mgmt_against==1) & (indexdf.iss_against==0)].count()[0] + activedf[(activatedf.mgmt_against==1) & (activatedf.iss_against==0)].count()[0]
tabledf = tabledf.append(temp)
tabledf
```

```
[48]:
```

	index fund				active fund		
	votedyes	votedno	abstained	didnotvote	votedyes	votedno	\
all	0.898148	0.066869	0.032586	0.002397	0.888770	0.075739	
yes yes	0.948730	0.036646	0.013684	0.000940	0.952452	0.033084	
no no	0.050827	0.826508	0.099179	0.023486	0.053712	0.829371	
yes no	0.533978	0.200873	0.247289	0.017860	0.438823	0.248889	
no yes	0.431135	0.512965	0.054824	0.001076	0.472298	0.467568	

	Difference PctYes			N
	abstained	didnotvote		
all	0.031220	0.004271	0.009378 27297366	
yes yes	0.011785	0.002679	-0.003723 24293163	
no no	0.101189	0.015728	-0.002885 398666	
yes no	0.286819	0.025469	0.095156 1761341	
no yes	0.057004	0.003130	-0.041163 844196	

```
[51]: tabledf.round(3).to_excel('table 4.xlsx')
```

### 3.11 Table 5

```
[173]: #
df = pd.read_stata('fundvotes.dta')
df = df[df.contentious_vote==1] #
df['activefund'] = (df['indexfund']==0)
df.set_index([df.firmid_anon2, df.year], inplace=True) #
X = df[['indexfund']]
y = df[['voted_with_mgmt']]
mod = PanelOLS(y, X, entity_effects=True, time_effects=True, singletons=False)
res = mod.fit(cov_type='clustered', clusters=df[['fundid_anon2', 'firmid_anon2']])
res.summary
```

```
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 97 singleton observations dropped
warn.warn()
```

```
[173]: <class 'linearmodels.compat.statsmodels.Summary'>
      ""
```

PanelOLS Estimation Summary

```

=====
Dep. Variable:          voted_with_mgmt    R-squared:                0.0092
Estimator:              PanelOLS          R-squared (Between):      0.1728
No. Observations:       2601806          R-squared (Within):       0.0093
Date:                   Sun, Jun 13 2021  R-squared (Overall):      0.0768
Time:                   13:02:30          Log-likelihood            -1.804e+06
Cov. Estimator:         Clustered

                               F-statistic:                2.417e+04
Entities:                5590              P-value                  0.0000
Avg Obs:                 465.44            Distribution:             F(1,2596201)
Min Obs:                 2.0000
Max Obs:                 2.843e+04         F-statistic (robust):     17.794
                               P-value                  0.0000
Time periods:           15                Distribution:             F(1,2596201)
Avg Obs:                 1.735e+05
Min Obs:                 8210.0
Max Obs:                 2.82e+05

```

#### Parameter Estimates

```

=====
              Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
indexfund      0.1006      0.0238     4.2183    0.0000     0.0538     0.1473
=====

```

F-test for Poolability: 28.846

P-value: 0.0000

Distribution: F(5603,2596201)

Included effects: Entity, Time

"""

```

[176]: #
df['exp_ratio#indexfund'] = df['exp_ratio'] * df['indexfund']
df['exp_ratio#activefund'] = df['exp_ratio'] * df['activefund']
X = df[['indexfund', 'exp_ratio#indexfund', 'exp_ratio#activefund']]
y = df[['voted_with_mgmt']]
mod = PanelOLS(y, X, entity_effects=True, time_effects=True, singletons=False)
res = mod.fit(cov_type='clustered', clusters=df[['fundid_anon2', 'firmid_anon2']])
res.summary

```

```

[176]: <class 'linearmodels.compat.statsmodels.Summary'>

```

"""

#### PanelOLS Estimation Summary

```

=====
Dep. Variable:          voted_with_mgmt    R-squared:                0.0204
Estimator:              PanelOLS          R-squared (Between):      0.1841

```

No. Observations:	2601806	R-squared (Within):	0.0206
Date:	Sun, Jun 13 2021	R-squared (Overall):	0.0831
Time:	13:11:09	Log-likelihood	-1.789e+06
Cov. Estimator:	Clustered		
		F-statistic:	1.805e+04
Entities:	5590	P-value	0.0000
Avg Obs:	465.44	Distribution:	F(3,2596199)
Min Obs:	2.0000		
Max Obs:	2.843e+04	F-statistic (robust):	12.170
		P-value	0.0000
Time periods:	15	Distribution:	F(3,2596199)
Avg Obs:	1.735e+05		
Min Obs:	8210.0		
Max Obs:	2.82e+05		

#### Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI
Upper CI					
indexfund	0.1017	0.0235	4.3287	0.0000	0.0557
0.1478					
exp_ratio#indexfund	-0.2863	0.0619	-4.6243	0.0000	-0.4076
-0.1649					
exp_ratio#activefund	-0.0225	0.0356	-0.6317	0.5276	-0.0922
0.0473					

F-test for Poolability: 29.813  
P-value: 0.0000  
Distribution: F(5603,2596199)

Included effects: Entity, Time  
"""

```
[174]: #
df[['firmid_anon2#year']] = df['firmid_anon2'] * df['year']
X = df[['indexfund']]
y = df[['voted_with_mgmt']]
mod = PanelOLS(y, X, entity_effects=False, time_effects=False,
               ↳singletons=False, other_effects=df[['firmid_anon2#year']])
res = mod.fit(cov_type='clustered', clusters=df[['fundid_anon2', 'firmid_anon2']])
res.summary
```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 193 singleton observations dropped  
warn.warn()

[174]: <class 'linearmodels.compat.statsmodels.Summary'>  
"""

```

                                PanelOLS Estimation Summary
=====
Dep. Variable:          voted_with_mgmt    R-squared:                0.0093
Estimator:              PanelOLS           R-squared (Between):      0.1677
No. Observations:       2601710           R-squared (Within):       0.0093
Date:                   Sun, Jun 13 2021   R-squared (Overall):      0.0757
Time:                   13:02:54          Log-likelihood            -1.753e+06
Cov. Estimator:         Clustered

                                F-statistic:                2.417e+04
Entities:                5627              P-value                  0.0000
Avg Obs:                 462.36            Distribution:             F(1,2586133)
Min Obs:                 1.0000
Max Obs:                 2.843e+04         F-statistic (robust):     17.388
                                P-value                  0.0000
Time periods:            15              Distribution:             F(1,2586133)
Avg Obs:                 1.734e+05
Min Obs:                 8170.0
Max Obs:                 2.82e+05

```

```

                                Parameter Estimates
=====
                Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
indexfund      0.0990      0.0237     4.1699    0.0000     0.0525     0.1455
=====

```

F-test for Poolability: 17.414  
P-value: 0.0000  
Distribution: F(15575,2586133)

Included effects: Other Effect (firmid\_anon2#year)  
Model includes 5 other effects  
Other Effect Observations per group (firmid\_anon2#year):  
Avg Obs: 164.99, Min Obs: 0.0000, Max Obs: 7168.0, Groups: 15769  
"""

```

[177]: #
X = df[['indexfund', 'exp_ratiow#indexfund', 'exp_ratiow#activefund']]
y = df[['voted_with_mgmt']]
mod = PanelOLS(y, X, entity_effects=False, time_effects=False,
↳ singletons=False, other_effects=df[['firmid_anon2#year']])
res = mod.fit(cov_type='clustered', clusters=df[['fundid_anon2', 'firmid_anon2']])

```

```
res.summary
```

```
[177]: <class 'linearmodels.compat.statsmodels.Summary'>
```

```
"""
```

#### PanelOLS Estimation Summary

```
=====
Dep. Variable:      voted_with_mgmt    R-squared:              0.0209
Estimator:          PanelOLS           R-squared (Between):    0.1781
No. Observations:   2601710           R-squared (Within):     0.0206
Date:               Sun, Jun 13 2021   R-squared (Overall):    0.0818
Time:               13:11:29           Log-likelihood          -1.737e+06
Cov. Estimator:     Clustered

                               F-statistic:      1.836e+04
Entities:            5627                 P-value              0.0000
Avg Obs:             462.36               Distribution:         F(3,2586131)
Min Obs:             1.0000
Max Obs:             2.843e+04            F-statistic (robust):  12.023
                               P-value              0.0000
Time periods:        15                  Distribution:         F(3,2586131)
Avg Obs:             1.734e+05
Min Obs:             8170.0
Max Obs:             2.82e+05
```

#### Parameter Estimates

```
=====
=====
                               Parameter  Std. Err.    T-stat    P-value    Lower CI
Upper CI
-----
-----
indexfund              0.0999      0.0234      4.2711    0.0000      0.0541
0.1458
exp_ratio#indexfund   -0.2859      0.0614     -4.6529    0.0000     -0.4063
-0.1655
exp_ratio#activefund  -0.0228      0.0352     -0.6480    0.5170     -0.0919
0.0462
=====
=====
```

F-test for Poolability: 17.846

P-value: 0.0000

Distribution: F(15575,2586131)

Included effects: Other Effect (firmid\_anon2#year)

Model includes 5 other effects

Other Effect Observations per group (firmid\_anon2#year):

Avg Obs: 164.99, Min Obs: 0.0000, Max Obs: 7168.0, Groups: 15769

"""

### 3.12 Table 6

```
[183]: #
df = pd.read_stata('fundvotes.dta')
df = df[df.contentious_vote==1] #
df[['firmid_anon2#year']] = df['firmid_anon2'] * df['year']
df.set_index([df.firmid_anon2, df.year], inplace=True) #
for vote_type in ['item_board', 'item_compensation', 'item_disclosure',
↳ 'item_E']:
    temp = df[df[vote_type]==1]
    X = temp[['indexfund']]
    y = temp[['voted_with_mgmt']]
    mod = PanelOLS(y, X, entity_effects=False, time_effects=False,
↳ singletons=False, other_effects=temp['firmid_anon2#year'])
    res = mod.
↳ fit(cov_type='clustered', clusters=temp[['fundid_anon2', 'firmid_anon2']])
    print(res.summary)
```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:  
SingletonWarning: 181 singleton observations dropped  
warn.warn()

#### PanelOLS Estimation Summary

```
=====
Dep. Variable:          voted_with_mgmt    R-squared:                0.0117
Estimator:              PanelOLS          R-squared (Between):      0.1743
No. Observations:       1428111           R-squared (Within):      0.0114
Date:                   Sun, Jun 13 2021   R-squared (Overall):     0.0927
Time:                   13:39:42           Log-likelihood           -9.552e+05
Cov. Estimator:         Clustered

                                F-statistic:          1.685e+04
Entities:                4619                P-value                0.0000
Avg Obs:                  309.18              Distribution:           F(1,1417144)
Min Obs:                  1.0000
Max Obs:                  1.73e+04             F-statistic (robust):   15.922
                                P-value                0.0001
Time periods:             15                 Distribution:           F(1,1417144)
Avg Obs:                  9.521e+04
Min Obs:                  5444.0
Max Obs:                  1.561e+05
```

#### Parameter Estimates

```
=====
Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
indexfund  0.1094    0.0274    3.9903    0.0001    0.0557    0.1631
```

```
=====
F-test for Poolability: 14.373
P-value: 0.0000
Distribution: F(10965,1417144)
```

```
Included effects: Other Effect (firmid_anon2#year)
Model includes 5 other effects
Other Effect Observations per group (firmid_anon2#year):
Avg Obs: 128.12, Min Obs: 0.0000, Max Obs: 5058.0, Groups: 11147
```

```
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 2 singleton observations dropped
warn.warn(
```

#### PanelOLS Estimation Summary

```
=====
Dep. Variable:      voted_with_mgmt      R-squared:      0.0105
Estimator:          PanelOLS             R-squared (Between): 0.1422
No. Observations:   35131               R-squared (Within): 0.0106
Date:              Sun, Jun 13 2021      R-squared (Overall): 0.0680
Time:              13:39:48              Log-likelihood     -2.388e+04
Cov. Estimator:     Clustered

F-statistic:      370.99
Entities:         214      P-value      0.0000
Avg Obs:          164.16   Distribution: F(1,34856)
Min Obs:          2.0000
Max Obs:          2516.0   F-statistic (robust): 13.080
P-value          0.0003
Time periods:     15      Distribution: F(1,34856)
Avg Obs:          2342.1
Min Obs:          126.00
Max Obs:          7811.0
```

#### Parameter Estimates

```
=====
Parameter Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
indexfund    0.1136    0.0314    3.6167    0.0003    0.0520    0.1751
=====
```

```
F-test for Poolability: 6.1292
P-value: 0.0000
Distribution: F(273,34856)
```

```
Included effects: Other Effect (firmid_anon2#year)
Model includes 5 other effects
Other Effect Observations per group (firmid_anon2#year):
Avg Obs: 127.29, Min Obs: 0.0000, Max Obs: 1106.0, Groups: 276
```



# PanelOLS Estimation Summary

```

=====
Dep. Variable:      voted_with_mgmt    R-squared:      0.0035
Estimator:          PanelOLS           R-squared (Between): 0.0589
No. Observations:   122322            R-squared (Within):  0.0034
Date:               Sun, Jun 13 2021   R-squared (Overall): 0.0346
Time:               13:39:49           Log-likelihood      -7.856e+04
Cov. Estimator:     Clustered

                               F-statistic:      430.62
Entities:           5479              P-value         0.0000
Avg Obs:            22.326            Distribution:      F(1,121793)
Min Obs:            0.0000
Max Obs:            4126.0            F-statistic (robust): 3.9597
                               P-value         0.0466
Time periods:       15                Distribution:      F(1,121793)
Avg Obs:            8154.8
Min Obs:            14.000
Max Obs:            1.812e+04

```

## Parameter Estimates

```

=====
               Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper CI
-----
indexfund      0.0629      0.0316     1.9899    0.0466     0.0009     0.1248
=====

```

F-test for Poolability: 4.8949

P-value: 0.0000

Distribution: F(527,121793)

Included effects: Other Effect (firmid\_anon2#year)

Model includes 5 other effects

Other Effect Observations per group (firmid\_anon2#year):

Avg Obs: 231.67, Min Obs: 2.0000, Max Obs: 852.00, Groups: 528

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 11 singleton observations dropped

warn.warn(

# PanelOLS Estimation Summary

```

=====
Dep. Variable:      voted_with_mgmt    R-squared:      0.0039
Estimator:          PanelOLS           R-squared (Between): 0.1335
No. Observations:   80767            R-squared (Within):  0.0043
Date:               Sun, Jun 13 2021   R-squared (Overall): 0.0409
Time:               13:39:50           Log-likelihood      -3.265e+04
Cov. Estimator:     Clustered

                               F-statistic:      315.61
Entities:           676              P-value         0.0000

```

Avg Obs:	119.48	Distribution:	F(1,79863)
Min Obs:	2.0000		
Max Obs:	2519.0	F-statistic (robust):	7.8399
		P-value	0.0051
Time periods:	15	Distribution:	F(1,79863)
Avg Obs:	5384.5		
Min Obs:	319.00		
Max Obs:	1.024e+04		

#### Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
indexfund	0.0483	0.0173	2.8000	0.0051	0.0145	0.0822

F-test for Poolability: 13.021

P-value: 0.0000

Distribution: F(902,79863)

Included effects: Other Effect (firmid\_anon2#year)

Model includes 5 other effects

Other Effect Observations per group (firmid\_anon2#year):

Avg Obs: 88.367, Min Obs: 0.0000, Max Obs: 534.00, Groups: 914

### 3.13 Table 7

```
[254]: df = pd.read_stata('blockholding_disclosure.dta')
#
X = df[['frac_passive']]
y = df[['frac_13D']].copy()
threshold = y.median()[0]
y[y>threshold] = 1
y[y<=threshold] = 0
mod = Probit(y, sm.add_constant(X))
res = mod.fit()
print(res.summary())
print(res.get_margeff().summary())

#
X = df[['frac_passive', 'logAUM']]
y = df[['frac_13D']].copy()
threshold = y.median()[0]
y[y>threshold] = 1
y[y<=threshold] = 0
mod = Probit(y, sm.add_constant(X))
res = mod.fit()
```

```

print(res.summary())
print(res.get_margeff().summary())
#
X = df[['frac_passive', 'n_blocks']]
y = df[['frac_13D']].copy()
threshold = y.median()[0]
y[y>threshold] = 1
y[y<=threshold] = 0
mod = Probit(y, sm.add_constant(X))
res = mod.fit()
print(res.summary())
print(res.get_margeff().summary())
# lf=LabelEncoder().fit(df['mgmt_cd'])
# df['mgmt_label'] = lf.transform(df['mgmt_cd'])
# cov_type='cluster', groups=np.array(df['mgmt_label'])
# mod = PanelOLS(y, X, entity_effects=True, time_effects=True, singletons=False)

```

Optimization terminated successfully.

Current function value: 0.444528

Iterations 6

#### Probit Regression Results

```

=====
Dep. Variable:          frac_13D    No. Observations:          1070
Model:                  Probit      Df Residuals:              1068
Method:                  MLE        Df Model:                  1
Date:                   Sun, 13 Jun 2021    Pseudo R-squ.:          0.01229
Time:                   15:24:38    Log-Likelihood:         -475.64
converged:               True        LL-Null:                 -481.56
Covariance Type:        nonrobust    LLR p-value:            0.0005802
=====

```

	coef	std err	z	P> z	[0.025	0.975]
const	-0.9247	0.047	-19.593	0.000	-1.017	-0.832
frac_passive	-0.9965	0.332	-3.005	0.003	-1.646	-0.347

#### Probit Marginal Effects

```

=====
Dep. Variable:          frac_13D
Method:                  dydx
At:                     overall
=====

```

	dy/dx	std err	z	P> z	[0.025	0.975]
frac_passive	-0.2459	0.082	-3.010	0.003	-0.406	-0.086

Optimization terminated successfully.

Current function value: 0.443178

Iterations 6

# Probit Regression Results

```

=====
Dep. Variable:          frac_13D    No. Observations:          1070
Model:                  Probit      Df Residuals:              1067
Method:                  MLE        Df Model:                  2
Date:                   Sun, 13 Jun 2021    Pseudo R-squ.:          0.01529
Time:                   15:24:38    Log-Likelihood:         -474.20
converged:              True        LL-Null:                 -481.56
Covariance Type:        nonrobust    LLR p-value:            0.0006339
=====

```

	coef	std err	z	P> z	[0.025	0.975]
const	-0.5438	0.229	-2.378	0.017	-0.992	-0.096
frac_passive	-0.9050	0.331	-2.737	0.006	-1.553	-0.257
logAUM	-0.0423	0.025	-1.696	0.090	-0.091	0.007

## Probit Marginal Effects

```

=====
Dep. Variable:          frac_13D
Method:                  dydx
At:                      overall
=====

```

	dy/dx	std err	z	P> z	[0.025	0.975]
frac_passive	-0.2225	0.081	-2.740	0.006	-0.382	-0.063
logAUM	-0.0104	0.006	-1.700	0.089	-0.022	0.002

Optimization terminated successfully.

Current function value: 0.443931

Iterations 6

# Probit Regression Results

```

=====
Dep. Variable:          frac_13D    No. Observations:          1070
Model:                  Probit      Df Residuals:              1067
Method:                  MLE        Df Model:                  2
Date:                   Sun, 13 Jun 2021    Pseudo R-squ.:          0.01362
Time:                   15:24:38    Log-Likelihood:         -475.01
converged:              True        LL-Null:                 -481.56
Covariance Type:        nonrobust    LLR p-value:            0.001419
=====

```

	coef	std err	z	P> z	[0.025	0.975]
const	-0.9415	0.050	-19.007	0.000	-1.039	-0.844
frac_passive	-1.0258	0.335	-3.062	0.002	-1.682	-0.369
n_blocks	0.0003	0.000	1.165	0.244	-0.000	0.001

## Probit Marginal Effects

Dep. Variable:	frac_13D					
Method:	dydx					
At:	overall					
=====						
	dy/dx	std err	z	P> z	[0.025	0.975]
-----						
frac_passive	-0.2527	0.082	-3.068	0.002	-0.414	-0.091
n_blocks	7.912e-05	6.78e-05	1.167	0.243	-5.38e-05	0.000
=====						

### 3.14 Table 8

```
[259]: # table 8a management proposals
df = pd.read_stata('fundvotes.dta')
df = df[df.contentious_vote==1] #
df[['firmid_anon2#year']] = df['firmid_anon2'] * df['year']
df.set_index([df.firmid_anon2, df.year], inplace=True) #
for y_name in ['votedyes', 'votedno', 'abstained']:
    df = df[df['shprop']==0] # management proposals
    X = df[['indexfund']]
    y = df[[y_name]]
    mod = PanelOLS(y, X, entity_effects=False, time_effects=False,
    ↪ singletons=False, other_effects=df['firmid_anon2#year'])
    res = mod.
    ↪ fit(cov_type='clustered', clusters=df[['fundid_anon2', 'firmid_anon2']])
    print('*****' + y_name +
    ↪ '*****')
    print(res.summary)
```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:  
SingletonWarning: 203 singleton observations dropped  
warn.warn()

\*\*\*\*\*votedyes\*\*\*\*\*  
\*\*\*\*\*

PanelOLS Estimation Summary			
=====			
Dep. Variable:	votedyes	R-squared:	0.0133
Estimator:	PanelOLS	R-squared (Between):	0.1942
No. Observations:	1738874	R-squared (Within):	0.0130
Date:	Sun, Jun 13 2021	R-squared (Overall):	0.1010
Time:	15:31:39	Log-likelihood	-1.166e+06
Cov. Estimator:	Clustered		
		F-statistic:	2.332e+04
Entities:	5456	P-value	0.0000
Avg Obs:	318.71	Distribution:	F(1,1724993)
Min Obs:	1.0000		
Max Obs:	1.924e+04	F-statistic (robust):	16.489

Time periods: 15 P-value 0.0000  
 Distribution: F(1,1724993)  
 Avg Obs: 1.159e+05  
 Min Obs: 5638.0  
 Max Obs: 2.046e+05

#### Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
indexfund	0.1161	0.0286	4.0606	0.0000	0.0600	0.1721

F-test for Poolability: 13.034  
 P-value: 0.0000  
 Distribution: F(13879,1724993)

Included effects: Other Effect (firmid\_anon2#year)  
 Model includes 5 other effects  
 Other Effect Observations per group (firmid\_anon2#year):  
 Avg Obs: 123.47, Min Obs: 0.0000, Max Obs: 5376.0, Groups: 14083

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:  
 SingletonWarning: 203 singleton observations dropped  
 warn.warn()

\*\*\*\*\*votedno\*\*\*\*\*  
 \*\*\*\*\*

#### PanelOLS Estimation Summary

```

=====
Dep. Variable:          votedno    R-squared:                0.0034
Estimator:              PanelOLS   R-squared (Between):      -0.0854
No. Observations:       1738874    R-squared (Within):       0.0024
Date:                   Sun, Jun 13 2021  R-squared (Overall):     -0.0362
Time:                   15:31:56    Log-likelihood            -6.14e+05
Cov. Estimator:         Clustered

                               F-statistic:          5940.5
Entities:                5456    P-value                0.0000
Avg Obs:                  318.71  Distribution:           F(1,1724993)
Min Obs:                   1.0000
Max Obs:                   1.924e+04  F-statistic (robust):    16.033
                               P-value                0.0001
Time periods:              15    Distribution:           F(1,1724993)
Avg Obs:                    1.159e+05
Min Obs:                     5638.0
Max Obs:                     2.046e+05
  
```

#### Parameter Estimates

=====

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
indexfund	-0.0426	0.0106	-4.0042	0.0001	-0.0635	-0.0218

F-test for Poolability: 58.865

P-value: 0.0000

Distribution: F(13879,1724993)

Included effects: Other Effect (firmid\_anon2#year)

Model includes 5 other effects

Other Effect Observations per group (firmid\_anon2#year):

Avg Obs: 123.47, Min Obs: 0.0000, Max Obs: 5376.0, Groups: 14083

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 203 singleton observations dropped

warn.warn()

\*\*\*\*\*abstained\*\*\*\*\*  
\*\*\*\*\*

#### PanelOLS Estimation Summary

```

=====
Dep. Variable:          abstained    R-squared:                0.0068
Estimator:              PanelOLS     R-squared (Between):      -0.1496
No. Observations:       1738874      R-squared (Within):       0.0071
Date:                   Sun, Jun 13 2021  R-squared (Overall):      -0.0600
Time:                   15:32:13         Log-likelihood            -7.769e+05
Cov. Estimator:         Clustered

                               F-statistic:          1.177e+04
Entities:                5456         P-value                0.0000
Avg Obs:                 318.71        Distribution:           F(1,1724993)
Min Obs:                 1.0000
Max Obs:                 1.924e+04      F-statistic (robust):    13.697
                               P-value                0.0002
Time periods:            15          Distribution:           F(1,1724993)
Avg Obs:                 1.159e+05
Min Obs:                 5638.0
Max Obs:                 2.046e+05

```

#### Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
indexfund	-0.0659	0.0178	-3.7009	0.0002	-0.1008	-0.0310

F-test for Poolability: 47.309

P-value: 0.0000

Distribution: F(13879,1724993)

Included effects: Other Effect (firmid\_anon2#year)  
 Model includes 5 other effects  
 Other Effect Observations per group (firmid\_anon2#year):  
 Avg Obs: 123.47, Min Obs: 0.0000, Max Obs: 5376.0, Groups: 14083

```
[260]: # table 8a shareholders proposals
df = pd.read_stata('fundvotes.dta')
df = df[df.contentious_vote==1] #
df[['firmid_anon2#year']] = df['firmid_anon2'] * df['year']
df.set_index([df.firmid_anon2, df.year], inplace=True) #
for y_name in ['votedyes', 'votedno', 'abstained']:
    df = df[df['shprop']==1] # shareholders proposals
    X = df[['indexfund']]
    y = df[[y_name]]
    mod = PanelOLS(y, X, entity_effects=False, time_effects=False,
    ↪singletons=False, other_effects=df['firmid_anon2#year'])
    res = mod.
    ↪fit(cov_type='clustered', clusters=df[['fundid_anon2', 'firmid_anon2']])
    print('*****' + y_name +
    ↪'*****')
    print(res.summary)
```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:  
 SingletonWarning: 5 singleton observations dropped  
 warn.warn()

\*\*\*\*\*votedyes\*\*\*\*\*  
 \*\*\*\*\*

#### PanelOLS Estimation Summary

```
=====
Dep. Variable:          votedyes      R-squared:                0.0039
Estimator:              PanelOLS      R-squared (Between):      -0.0859
No. Observations:       862821        R-squared (Within):       0.0040
Date:                   Sun, Jun 13 2021  R-squared (Overall):     -0.0378
Time:                   15:39:26         Log-likelihood            -5.68e+05
Cov. Estimator:         Clustered

                               F-statistic:          3338.1
Entities:                1120      P-value              0.0000
Avg Obs:                  770.38    Distribution:          F(1,859462)
Min Obs:                  1.0000
Max Obs:                  1.875e+04  F-statistic (robust):    7.9822
                               P-value              0.0047
Time periods:             15      Distribution:          F(1,859462)
Avg Obs:                  5.752e+04
Min Obs:                  2524.0
Max Obs:                  1.056e+05
```



# Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
indexfund	-0.0664	0.0235	-2.8253	0.0047	-0.1125	-0.0203

F-test for Poolability: 33.783

P-value: 0.0000

Distribution: F(3357,859462)

Included effects: Other Effect (firmid\_anon2#year)

Model includes 5 other effects

Other Effect Observations per group (firmid\_anon2#year):

Avg Obs: 256.56, Min Obs: 0.0000, Max Obs: 3236.0, Groups: 3363

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 5 singleton observations dropped

warn.warn()

\*\*\*\*\*votedno\*\*\*\*\*  
\*\*\*\*\*

## PanelOLS Estimation Summary

Dep. Variable:	votedno	R-squared:	0.0040
Estimator:	PanelOLS	R-squared (Between):	0.1069
No. Observations:	862821	R-squared (Within):	0.0043
Date:	Sun, Jun 13 2021	R-squared (Overall):	0.0391
Time:	15:39:34	Log-likelihood	-5.825e+05
Cov. Estimator:	Clustered		
		F-statistic:	3462.5
Entities:	1120	P-value	0.0000
Avg Obs:	770.38	Distribution:	F(1,859462)
Min Obs:	1.0000		
Max Obs:	1.875e+04	F-statistic (robust):	9.9115
		P-value	0.0016
Time periods:	15	Distribution:	F(1,859462)
Avg Obs:	5.752e+04		
Min Obs:	2524.0		
Max Obs:	1.056e+05		

# Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
indexfund	0.0688	0.0219	3.1482	0.0016	0.0260	0.1116

F-test for Poolability: 26.183

P-value: 0.0000

Distribution: F(3357,859462)

Included effects: Other Effect (firmid\_anon2#year)

Model includes 5 other effects

Other Effect Observations per group (firmid\_anon2#year):

Avg Obs: 256.56, Min Obs: 0.0000, Max Obs: 3236.0, Groups: 3363

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 5 singleton observations dropped

warn.warn()

\*\*\*\*\*abstained\*\*\*\*\*  
\*\*\*\*\*

#### PanelOLS Estimation Summary

Dep. Variable:	abstained	R-squared:	6.279e-06
Estimator:	PanelOLS	R-squared (Between):	0.0040
No. Observations:	862821	R-squared (Within):	-4.23e-06
Date:	Sun, Jun 13 2021	R-squared (Overall):	0.0007
Time:	15:39:42	Log-likelihood	8.433e+04
Cov. Estimator:	Clustered		
		F-statistic:	5.3970
Entities:	1120	P-value	0.0202
Avg Obs:	770.38	Distribution:	F(1,859462)
Min Obs:	1.0000		
Max Obs:	1.875e+04	F-statistic (robust):	0.0256
		P-value	0.8730
Time periods:	15	Distribution:	F(1,859462)
Avg Obs:	5.752e+04		
Min Obs:	2524.0		
Max Obs:	1.056e+05		

#### Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
indexfund	0.0013	0.0078	0.1599	0.8730	-0.0141	0.0166

F-test for Poolability: 30.720

P-value: 0.0000

Distribution: F(3357,859462)

Included effects: Other Effect (firmid\_anon2#year)

Model includes 5 other effects

Other Effect Observations per group (firmid\_anon2#year):

Avg Obs: 256.56, Min Obs: 0.0000, Max Obs: 3236.0, Groups: 3363

```
[266]: # table 8b shareholders proposals
df = pd.read_stata('Russell_sample_100.dta')
df['firmid_anon#cohort'] = df['firmid_anon'] * df['cohort']
df.set_index([df['firmid_anon#cohort'],df.year], inplace=True) #
X = df[['R1000_to_R2000','R2000_to_R1000']]
for y_name in ['n_contentious_mgmtprop', 'frac_contentious_mgmtprop',
↳ 'frac_mgmtprop_passed', 'n_contentious_shprop', 'frac_contentious_shprop',
↳ 'frac_shprop_passed']:
    y = df[[y_name]]
    mod = PanelOLS(y, X, entity_effects=True, time_effects=True,
↳ singletons=False)
    res = mod.fit(cov_type='clustered',clusters=df[['firmid_anon','year']])
    print('*****' + y_name +
↳ '*****')
    print(res.summary)
```

D:\program files\Anaconda\lib\site-

packages\linearmodels\shared\exceptions.py:35: MissingValueWarning:

Inputs contain missing values. Dropping rows with missing observations.

warnings.warn(missing\_value\_warning\_msg, MissingValueWarning)

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:

SingletonWarning: 11 singleton observations dropped

warn.warn(

\*\*\*\*\*n\_contentious\_mgmtprop\*\*\*\*\*  
\*\*\*\*\*

#### PanelOLS Estimation Summary

=====

Dep. Variable:	n_contentious_mgmtprop	R-squared:	
0.0004			
Estimator:	PanelOLS	R-squared (Between):	
-0.0056			
No. Observations:	4137	R-squared (Within):	
0.0005			
Date:	Sun, Jun 13 2021	R-squared (Overall):	
-0.0034			
Time:	15:56:24	Log-likelihood	
-6314.2			
Cov. Estimator:	Clustered	F-statistic:	
0.6370			
Entities:	798	P-value	
0.5290			
Avg Obs:	5.1842	Distribution:	
F(2,3323)			
Min Obs:	2.0000		
Max Obs:	6.0000	F-statistic (robust):	

0.3404

P-value

0.7115

Time periods:

15

Distribution:

F(2,3323)

Avg Obs:

275.80

Min Obs:

18.000

Max Obs:

474.00

#### Parameter Estimates

=====						
==						
	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper
CI						
-----						
--						
R1000_to_R2000	-0.1359	0.1665	-0.8160	0.4146	-0.4624	
0.1906						
R2000_to_R1000	0.0051	0.1019	0.0497	0.9603	-0.1946	
0.2048						
=====						
==						

F-test for Poolability: 4.7694

P-value: 0.0000

Distribution: F(811,3323)

Included effects: Entity, Time

\*\*\*\*\*frac\_contentious\_mgmtprop\*\*\*\*\*

\*\*\*\*\*

#### PanelOLS Estimation Summary

=====			
=====			
Dep. Variable:	frac_contentious_mgmtprop	R-squared:	
0.0004			
Estimator:	PanelOLS	R-squared (Between):	
-0.0047			
No. Observations:	4137	R-squared (Within):	
0.0007			
Date:	Sun, Jun 13 2021	R-squared (Overall):	
-0.0027			
Time:	15:56:25	Log-likelihood	
2733.6			
Cov. Estimator:	Clustered	F-statistic:	
0.5851			
Entities:	798	P-value	
0.5571			

Avg Obs: 5.1842 Distribution:  
 F(2,3323)  
 Min Obs: 2.0000  
 Max Obs: 6.0000 F-statistic (robust):  
 0.4814  
 P-value  
 0.6179  
 Time periods: 15 Distribution:  
 F(2,3323)  
 Avg Obs: 275.80  
 Min Obs: 18.000  
 Max Obs: 474.00

#### Parameter Estimates

=====						
==						
	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
-----						
--						
R1000_to_R2000	-0.0144	0.0154	-0.9368	0.3489	-0.0446	0.0157
R2000_to_R1000	0.0012	0.0101	0.1228	0.9023	-0.0185	0.0210
=====						
==						

F-test for Poolability: 5.0647  
 P-value: 0.0000  
 Distribution: F(811,3323)

Included effects: Entity, Time

\*\*\*\*\*frac\_mgmtprop\_passed\*\*\*\*\*  
 \*\*\*\*\*

#### PanelOLS Estimation Summary

=====  
 ==  
 Dep. Variable: frac\_mgmtprop\_passed R-squared:  
 3.137e-06  
 Estimator: PanelOLS R-squared (Between):  
 -0.0002  
 No. Observations: 4137 R-squared (Within):  
 0.0001  
 Date: Sun, Jun 13 2021 R-squared (Overall):  
 -0.0002  
 Time: 15:56:25 Log-likelihood  
 5148.3  
 Cov. Estimator: Clustered

```

                                F-statistic:
0.0052
Entities:                        798    P-value
0.9948
Avg Obs:                        5.1842  Distribution:
F(2,3323)
Min Obs:                        2.0000
Max Obs:                        6.0000  F-statistic (robust):
0.0061
                                P-value
0.9939
Time periods:                    15    Distribution:
F(2,3323)
Avg Obs:                        275.80
Min Obs:                        18.000
Max Obs:                        474.00

```

#### Parameter Estimates

```

=====
==
                                Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper
CI
-----
--
R1000_to_R2000    -0.0005    0.0072    -0.0739    0.9411    -0.0146
0.0136
R2000_to_R1000    -0.0005    0.0066    -0.0735    0.9414    -0.0135
0.0125
=====
==

```

```

F-test for Poolability: 1.7625
P-value: 0.0000
Distribution: F(811,3323)

```

Included effects: Entity, Time

```

D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 106 singleton observations dropped
  warn.warn(

```

```

*****n_contentious_shprop*****
*****

```

#### PanelOLS Estimation Summary

```

=====
==
Dep. Variable:    n_contentious_shprop    R-squared:
0.0095
Estimator:        PanelOLS    R-squared (Between):

```

0.0566  
 No. Observations: 198 R-squared (Within):  
 0.0073  
 Date: Sun, Jun 13 2021 R-squared (Overall):  
 0.0473  
 Time: 15:56:25 Log-likelihood  
 -112.95  
 Cov. Estimator: Clustered  
 F-statistic:  
 0.5799  
 Entities: 63 P-value  
 0.5615  
 Avg Obs: 3.1429 Distribution:  
 F(2,121)  
 Min Obs: 2.0000  
 Max Obs: 6.0000 F-statistic (robust):  
 0.5995  
 P-value  
 0.5507  
 Time periods: 13 Distribution:  
 F(2,121)  
 Avg Obs: 15.231  
 Min Obs: 3.0000  
 Max Obs: 27.000

#### Parameter Estimates

=====						
==						
	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper
CI	-----					
--						
R1000_to_R2000	0.2197	0.3928	0.5594	0.5769	-0.5580	
0.9975						
R2000_to_R1000	0.2448	0.3238	0.7559	0.4512	-0.3963	
0.8859						
=====						
==						

F-test for Poolability: 1.4162  
 P-value: 0.0447  
 Distribution: F(74,121)

Included effects: Entity, Time

\*\*\*\*\*frac\_contentious\_shprop\*\*\*\*\*  
 \*\*\*\*\*

#### PanelOLS Estimation Summary

=====

```

=====
Dep. Variable:      frac_contentious_shprop   R-squared:
0.0041
Estimator:          PanelOLS                 R-squared (Between):
0.0306
No. Observations:   198                     R-squared (Within):
0.0096
Date:               Sun, Jun 13 2021         R-squared (Overall):
0.0268
Time:               15:56:25                 Log-likelihood
-6.7535
Cov. Estimator:     Clustered
F-statistic:
0.2496
Entities:           63                     P-value
0.7795
Avg Obs:            3.1429                 Distribution:
F(2,121)
Min Obs:            2.0000
Max Obs:            6.0000                 F-statistic (robust):
0.2906
P-value
0.7484
Time periods:       13                     Distribution:
F(2,121)
Avg Obs:            15.231
Min Obs:            3.0000
Max Obs:            27.000

```

#### Parameter Estimates

	Parameter	Std. Err.	T-stat	P-value	Lower CI	Upper CI
CI						
--						
R1000_to_R2000	0.0776	0.1270	0.6105	0.5427	-0.1740	
0.3291						
R2000_to_R1000	0.0960	0.2259	0.4252	0.6715	-0.3512	
0.5433						

```

F-test for Poolability: 1.6454
P-value: 0.0075
Distribution: F(74,121)

```

Included effects: Entity, Time



```
*****frac_shprop_passed*****
*****
```

#### PanelOLS Estimation Summary

```
=====
Dep. Variable:    frac_shprop_passed    R-squared:                0.0081
Estimator:        PanelOLS              R-squared (Between):      -0.0351
No. Observations: 198                  R-squared (Within):       0.0151
Date:             Sun, Jun 13 2021      R-squared (Overall):      -0.0238
Time:             15:56:26              Log-likelihood            28.340
Cov. Estimator:   Clustered

F-statistic:                0.4958
Entities:                  63      P-value                    0.6103
Avg Obs:                   3.1429  Distribution:              F(2,121)
Min Obs:                   2.0000
Max Obs:                   6.0000  F-statistic (robust):     1.3213
                                   P-value                    0.2706
Time periods:              13      Distribution:              F(2,121)
Avg Obs:                   15.231
Min Obs:                   3.0000
Max Obs:                   27.000
```

#### Parameter Estimates

```
=====
==
               Parameter  Std. Err.    T-stat    P-value    Lower CI    Upper
CI
-----
--
R1000_to_R2000    -0.1748     0.1113    -1.5709    0.1188    -0.3950
0.0455
R2000_to_R1000    -0.0597     0.1483    -0.4028    0.6878    -0.3533
0.2339
=====
==
```

F-test for Poolability: 3.8073

P-value: 0.0000

Distribution: F(74,121)

Included effects: Entity, Time

### 3.15 Table 9

```
[375]: # Table 9a
df = pd.read_stata('Russell_sample_100.dta')
df['firmid_anon#cohort'] = df['firmid_anon'] * df['cohort']
df.set_index([df['firmid_anon#cohort'],df.year], inplace=True) #
```

```

X = df[['R1000_to_R2000', 'R2000_to_R1000']]
tempdf = pd.DataFrame()
for y_name in ['delta', 'totalcomp', 'equityvscash', 'bcf6', 'ceo_turnover']:
    y = df[[y_name]]
    mod = PanelOLS(y, X, entity_effects=True, time_effects=True,
↳ singletons=False)
    res = mod.fit(cov_type='clustered', clusters=df[['firmid_anon', 'year']])
    #print('*****' + y_name +
↳ '*****')
    #print(res.summary)
    temp = []
    temp.append(round(res.params[0], 2))
    temp.append('(' + str(round(res.std_errors[0], 2)) + ')')
    temp.append(round(res.params[1], 2))
    temp.append('(' + str(round(res.std_errors[1], 2)) + ')')
    temp.append(res.nobs)
    temp.append(round(res.rsquared, 3))
    temp.append('Yes')
    temp.append('Yes')
    temp = pd.DataFrame(temp, index=['1000-2000', '1000-2000 ste', '2000-1000',
↳ '2000-1000 ste', 'obs', 'R2', 'Firm Cohort FE', 'Year FE'], columns=[y_name])
    tempdf = pd.concat([tempdf, temp], axis=1)

```

```

D:\program files\Anaconda\lib\site-
packages\linearmodels\shared\exceptions.py:35: MissingValueWarning:
Inputs contain missing values. Dropping rows with missing observations.
    warnings.warn(missing_value_warning_msg, MissingValueWarning)
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 6 singleton observations dropped
    warn.warn(
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 5 singleton observations dropped
    warn.warn(
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 98 singleton observations dropped
    warn.warn(
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 7 singleton observations dropped
    warn.warn(

```

[377]: tempdf

```

[377]:
          delta totalcomp equityvscash    bcf6 ceo_turnover
1000-2000    -0.43      0.56      -0.06      0      -0.06
1000-2000 ste (0.11)   (0.08)   (0.02) (0.03)   (0.05)
2000-1000     0.27     -0.41      0.03    0.02      0.02
2000-1000 ste (0.11)   (0.06)   (0.01) (0.03)   (0.03)

```

obs	3445	3219	3138	2592	3923
R2	0.017	0.038	0.011	0	0.001
Firm Cohort FE	Yes	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes	Yes

```
[376]: tempdf.to_excel('table 9a.xlsx')
```

```
[378]: # table 9b
df = pd.read_stata('Russell_sample_100.dta')
df['firmid_anon#cohort'] = df['firmid_anon'] * df['cohort']
df.set_index([df['firmid_anon#cohort'], df.year], inplace=True) #
X = df[['R1000_to_R2000', 'R2000_to_R1000']]
tempdf = pd.DataFrame()
for y_name in ['indpt_board_pct', 'bcf', 'bcf5', 'supermajority_req',
↳ 'limitspecialmeet', 'writtenconsent', 'dualclass']:
    y = df[[y_name]]
    mod = PanelOLS(y, X, entity_effects=True, time_effects=True,
↳ singletons=False)
    res = mod.fit(cov_type='clustered', clusters=df[['firmid_anon', 'year']])
    #print('*****' + y_name +
↳ '*****')
    #print(res.summary)
    temp = []
    temp.append(round(res.params[0], 2))
    temp.append('(' + str(round(res.std_errors[0], 2)) + ')')
    temp.append(round(res.params[1], 2))
    temp.append('(' + str(round(res.std_errors[1], 2)) + ')')
    temp.append(res.nobs)
    temp.append(round(res.rsquared, 3))
    temp.append('Yes')
    temp.append('Yes')
    temp = pd.DataFrame(temp, index=['1000-2000', '1000-2000 ste', '2000-1000',
↳ '2000-1000 ste', 'obs', 'R2', 'Firm Cohort FE', 'Year FE'], columns=[y_name])
    tempdf = pd.concat([tempdf, temp], axis=1)
```

```
D:\program files\Anaconda\lib\site-
packages\linearmodels\shared\exceptions.py:35: MissingValueWarning:
Inputs contain missing values. Dropping rows with missing observations.
warnings.warn(missing_value_warning_msg, MissingValueWarning)
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 22 singleton observations dropped
warn.warn(
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 98 singleton observations dropped
warn.warn(
```

```
[379]: tempdf
```

```
[379]:
```

	indpt_board_pct	bcf	bcf5	supermajority_req \
1000-2000	-0.03	-0.07	-0.06	-0.01
1000-2000 ste	(0.01)	(0.08)	(0.05)	(0.01)
2000-1000	0	0.05	-0.01	0.01
2000-1000 ste	(0.01)	(0.06)	(0.04)	(0.02)
obs	2613	2592	2592	2592
R2	0.007	0.001	0.003	0.001
Firm Cohort FE	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes

	limitspecialmeet	writtenconsent	dualclass
1000-2000	-0	-0.05	0
1000-2000 ste	(0.02)	(0.03)	(0.0)
2000-1000	0	0.06	-0.01
2000-1000 ste	(0.03)	(0.04)	(0.01)
obs	2592	2592	2592
R2	0	0.005	0.002
Firm Cohort FE	Yes	Yes	Yes
Year FE	Yes	Yes	Yes

```
[380]: tempdf.to_excel('table 9b.xlsx')
```

### 3.16 Table 10

```
[381]: # table 10
df = pd.read_stata('Russell_sample_100.dta')
df['firmid_anon#cohort'] = df['firmid_anon'] * df['cohort']
df.set_index([df['firmid_anon#cohort'], df.year], inplace=True) #
X = df[['R1000_to_R2000', 'R2000_to_R1000']]
tempdf = pd.DataFrame()
for y_name in ['logQ', 'logqtot', 'logMB', 'ROA']:
    y = df[[y_name]]
    mod = PanelOLS(y, X, entity_effects=True, time_effects=True,
→singletons=False)
    res = mod.fit(cov_type='clustered', clusters=df[['firmid_anon', 'year']])
    #print('*****' + y_name +
→'*****')
    #print(res.summary)
    temp = []
    temp.append(round(res.params[0], 2))
    temp.append('(' + str(round(res.std_errors[0], 2)) + ')')
    temp.append(round(res.params[1], 2))
    temp.append('(' + str(round(res.std_errors[1], 2)) + ')')
    temp.append(res.nobs)
    temp.append(round(res.rsquared, 3))
    temp.append('Yes')
    temp.append('Yes')
```

```
temp = pd.DataFrame(temp, index=['1000-2000', '1000-2000 ste', '2000-1000',
→ '2000-1000 ste', 'obs', 'R2', 'Firm Cohort FE', 'Year FE'], columns=[y_name])
tempdf = pd.concat([tempdf,temp], axis=1)
```

```
D:\program files\Anaconda\lib\site-
packages\linearmodels\shared\exceptions.py:35: MissingValueWarning:
Inputs contain missing values. Dropping rows with missing observations.
  warnings.warn(missing_value_warning_msg, MissingValueWarning)
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 16 singleton observations dropped
  warn.warn(
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 4 singleton observations dropped
  warn.warn(
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 2 singleton observations dropped
  warn.warn(
D:\program files\Anaconda\lib\site-packages\linearmodels\panel\model.py:1195:
SingletonWarning: 6 singleton observations dropped
  warn.warn(
```

```
[382]: tempdf
```

```
[382]:
```

	logQ	logqtot	logMB	ROA
1000-2000	-0.1	-0.21	-0.12	-0.03
1000-2000 ste	(0.03)	(0.06)	(0.05)	(0.01)
2000-1000	0.01	0.06	-0.03	0
2000-1000 ste	(0.01)	(0.03)	(0.03)	(0.01)
obs	4296	3403	4552	4188
R2	0.014	0.024	0.006	0.008
Firm Cohort FE	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes

```
[383]: tempdf.to_excel('table 10.xlsx')
```