How are stocks connected? Evidence from an emerging market

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- Stock return co-movement is caused by direct or indirect common ownership?
 - common ownership:
 - ullet We connect stocks through the common ownership by blockholders (ownership >1%) for direct common ownership
 - We connect stocks through the ultimate owner for indirect common ownership
 - We focus on excess return co-movement for a pair of the stocks
 - \bullet We use common ownership (direct or indirect) to forecast cross-sectional variation in the realized correlation of four-factor + industry residuals
 - We demonstrate that correlated trading can be a channel of co-movement

Why does it matter?

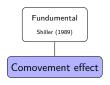
- Covariance
 - Covariance is a key component of risk in many financial applications. (Portfolio selection, Risk management, Hedging and Asset pricing)
 - Covariance is a significant input in risk measurement models (Such as Value-at-Risk)
- Return predictability
 - If it's valid, we can build a profitable buy-sell strategy

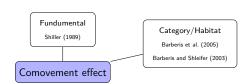
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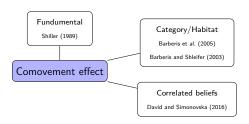
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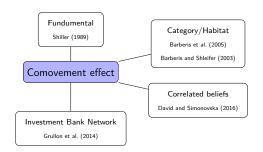


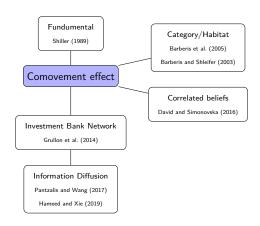
Comovement effect



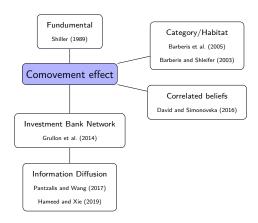


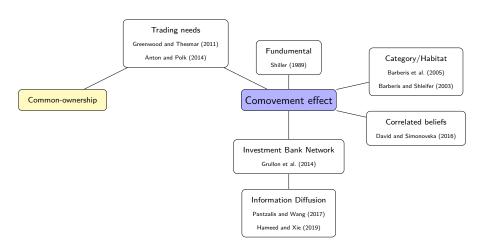


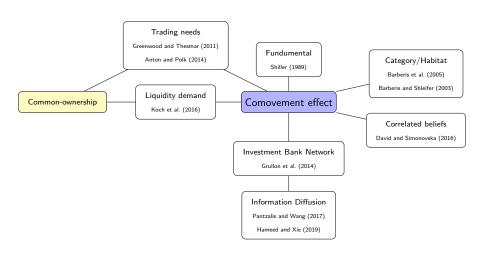




Common-ownership







Our work

- We use daily records of block-holder ownership for firms
- We are not restricted to mutual funds ownership
- Furthermore, 85% of market belongs to the business groups
 - Would business groups be able to raise the co-movement of stock returns?
 - Cho and Mooney (2015):
 The strong co-movement between group returns and firm returns is explained by correlated fundamentals.
 - Kim et al. (2015):
 The increase in correlation appears to be driven more by non-fundamental factors such as correlated trading, rather than fundamental factors such as related-party transactions
 - Common ownership or business group (indirect common ownership) ?
 - Through which channel?

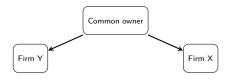
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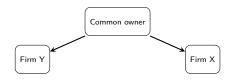
Pair composition

• Firms with at least one common owner



Pair composition

• Firms with at least one common owner

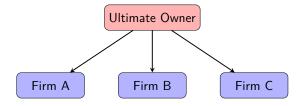


- In a business group, how can one pair be defined?
 - What is the business group?

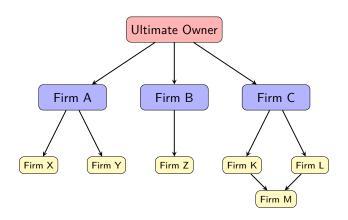
Business Group

Ultimate Owner

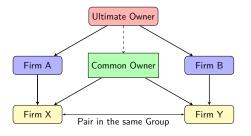
Business Group



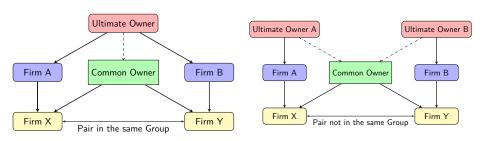
Business Group



Pair in the Business Group



Pair in the Business Group



Data Summary

- We use blockholders' data from 2014/03/25 (1393/01/06) to 2020/03/18 (1398/12/28)
 - Includes of 72 Months
 - Consists of 618 firm inculding 562 firm with common owners

Year	2014	2015	2016	2017	2018	2019
No. of Firms	365	376	447	552	587	618
No. of Blockholders	777	803	984	1297	1454	1458
No. of Groups	38	41	43	44	40	43
No. of Firms in Groups	249	268	300	336	346	375
Ave. Number of group Members	7	7	7	8	9	9
Ave. ownership of each Blockholders	21	22	22	21	22	23
Med. ownership of each Blockholders	7	8	8	8	8	9
Ave. Number of Owners	5	5	5	5	5	5
Ave. Block. Ownership	76	77	75	75	75	71

Pair Composition

- Pairs consist of two firms with at least one common owner
 - 93442 unique pairs which is 25% of possible pairs ($\frac{612+611}{2} = 373932$)

	mean	min	Median	max
Number of unique paris	24139	13272	23024	45795

Year	2014	2015	2016	2017	2018	2019
No. of Pairs	20876	21187	27784	41449	47234	67232
No. of Pairs not in Groups	11452	11192	15351	26530	29182	43433
No. of Pairs not in the same Group	7962	8731	10971	12916	15366	20745
No. of Pairs in the same Group	923	955	1099	1260	1536	1774
Ave. Number of Common owner	1	1	1	1	1	1

Measuring Common-ownership

Anton and Polk (2014)

$$FCAP_{ij,t} = \frac{\sum_{f=1}^{F} (S_{i,t}^{f} P_{i,t} + S_{j,t}^{f} P_{j,t})}{S_{i,t} + S_{i,t} + S_{i,t}}$$

SQRT

$$FCAP_{ij,t} = \frac{\sum_{f=1}^{F} (S_{i,t}^{f} P_{i,t} + S_{j,t}^{f} P_{j,t})}{S_{i,t}Pi, t + S_{j,t}Pj, t} \left[MFCAP_{ij,t} = \left[\frac{\sum_{f=1}^{F} (\sqrt{S_{i,t}^{f} P_{i,t}} + \sqrt{S_{j,t}^{f} P_{j,t}})}{\sqrt{S_{i,t}Pi, t} + \sqrt{S_{j,t}Pj, t}} \right]^{2} \right]$$

Measuring Common-ownership

Anton and Polk (2014)

SQRT

$$FCAP_{ij,t} = \frac{\sum_{f=1}^{F} (S_{i,t}^{f} P_{i,t} + S_{j,t}^{f} P_{j,t})}{S_{i,t} P_{i,t} + S_{j,t} P_{j,t}}$$

$$MFCAP_{ij,t} = \left[\frac{\sum_{f=1}^{F} (\sqrt{S_{i,t}^{f} P_{i,t}} + \sqrt{S_{j,t}^{f} P_{j,t}})}{\sqrt{S_{i,t} P_{i,t}} + \sqrt{S_{j,t} P_{j,t}}}\right]^{2}$$

Intuition

If for a pair of stocks with n mutual owners, all owners have even shares of each firm's market cap, then the proposed indexes will be equal to n. ${\color{red}Proof}$

MFCAP vs. FCAP Summary

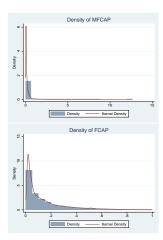
	MFCAP				FCAP					
	mean	std	min	median	max	mean	std	min	median	max
All	0.158	0.234	0.002	0.079	12.65	0.144	0.166	0.002	0.077	1.0
Same Group	0.474	0.478	0.005	0.367	6.174	0.346	0.265	0.004	0.321	1.0
Not Same Group	0.147	0.212	0.002	0.077	12.65	0.137	0.157	0.002	0.074	1.0
Same Industry	0.274	0.383	0.003	0.126	6.262	0.207	0.215	0.003	0.12	0.999
Not Same Industry	0.15	0.217	0.002	0.077	12.65	0.14	0.161	0.002	0.074	1.0

Results

- By the proposed measurement, common ownership increases
- Common ownership is greater in pairs that are in the same business group and insutry

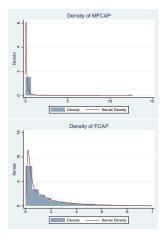
MFCAP vs. FCAP Distributions

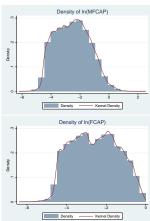
Monthly



MFCAP vs. FCAP Distributions

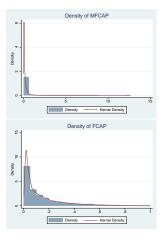
Monthly

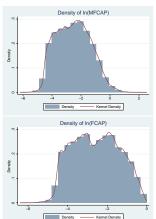


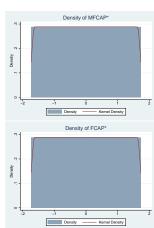


MFCAP vs. FCAP Distributions

Monthly







Correlation Calculation

4 Factor + Industry

Frist Step:

Estimate this model on periods of three month (From two months earlier):

• 4 Factor + Industry :

$$\begin{aligned} \textit{R}_{\textit{i},t} &= \alpha_{\textit{i}} + \beta_{\textit{mkt},\textit{i}} \textit{R}_{\textit{M},t} + \beta_{\textit{Ind},\textit{i}} \textit{R}_{\textit{Ind},t} \\ &+ \beta_{\textit{HML},\textit{i}} \textit{HML}_{t} + \beta_{\textit{SMB},\textit{i}} \textit{SMB}_{t} + \beta_{\textit{UMD},\textit{i}} \textit{UMD}_{t} + \boxed{\varepsilon_{\textit{i},t}} \end{aligned}$$

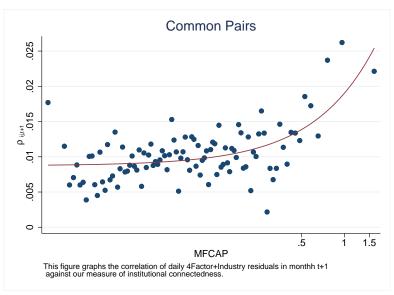
 Second Step: Calculate monthly correlation of each stock pair's daily abnormal returns (residuals)

	mean	std	min	median	max
CAPM + Industry	0.021	0.200	-1.0	0.016	1.0
4 Factor	0.032	0.202	-1.0	0.025	1.0
4 Factor + Industry	0.016	0.199	-1.0	0.010	1.0

Conclusion

We use the 4 Factor + Industry model to control for exposure to systematic risk because it almost captures all correlations between two firms in each pair.

Future Correlation via MFCAP



Controls

- **SameGroup**: Dummy variable for whether the two stocks belong to the same business group.
- **SameIndustry**: Dummy variable for whether the two stocks belong to the same Industry.
- SameSize: The negative of absolute difference in percentile ranking of size across a pair
- SameBookToMarket :The negative of absolute difference in percentile ranking of the book to market ratio across a pair
- **CrossOwnership**: The maximum percent of cross-ownership between two firms

Industry & Business group

	Yes	No
SameIndustry	4541	74837
	(5.7%)	(94.3%)
SameGroup	1834	27157
	(6.3%)	(93.7%)
SameGroup & SameIndustry	696	79378
	(0.9%)	(99.1%)

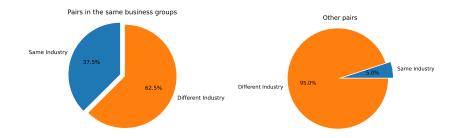


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Fama-MacBeth Estimation

- Fama-MacBeth regression analysis is implemented using a two-step procedure.
 - The first step is to run periodic cross-sectional regression for dependent variables using data of each period.
 - The second step is to analyze the time series of each regression coefficient to determine whether the average coefficient differs from zero.

Fama-MacBeth (1973)

- Two Step Regression
 - First Step

$$Y_{i1} = \delta_{0,1} + \delta_{1,1}^{1} X_{i,1}^{1} + \dots + \delta_{k,1}^{k} X_{i,1}^{k} + \varepsilon_{i,1}$$

$$\vdots$$

$$Y_{iT} = \delta_{0,1} + \delta_{1,T}^{1} X_{i,T}^{1} + \dots + \delta_{k,T}^{k} X_{i,T}^{k} + \varepsilon_{i,T}$$

Second Step

$$\begin{bmatrix} \bar{Y}_1 \\ \vdots \\ \bar{Y}_T \end{bmatrix}_{T \times 1} = \begin{bmatrix} 1 & \delta_1^0 & \delta_1^1 & \dots & \delta_1^k \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & \delta_T^0 & \delta_T^1 & \dots & \delta_T^k \end{bmatrix}_{T \times (k+2)} \times \begin{bmatrix} \lambda \\ \lambda_0 \\ \lambda_1 \\ \vdots \\ \lambda_k \end{bmatrix}_{(k+2) \times 1}$$

• Fama-MacBeth technique was developed to account for correlation between observations on different firms in the same period

Calculating standard errors

- In most cases, the standard errors are adjusted following Newey and West (1987).
 - Newey and West (1987) adjustment to the results of the regression produces a new standard error for the estimated mean that is adjusted for autocorrelation and heteroscedasticity.
 - Only input is the number of lags to use when performing the adjustment

$$Lag = 4(T/100)^{\frac{2}{9}}$$

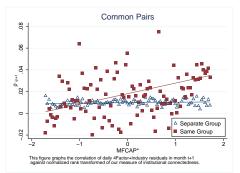
where T is the number of periods in the time series

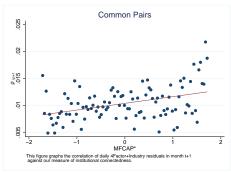
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Future Correlation via MFCAP

Normalized Rank-Transformed





Estimation model

Use Fama-MacBeth to estimate this model

$$\begin{split} \rho_{ij,t+1} &= \beta_0 + \beta_1 * \mathsf{MFCAP}^*_{ij,t} + \beta_2 * \mathsf{SameGroup}_{ij} \\ &+ \beta_3 * \mathsf{MFCAP}^*_{ij,t} \times \mathsf{SameGroup}_{ij} \\ &+ \sum_{k=1}^n \alpha_k * \mathsf{Control}_{ij,t} + \varepsilon_{ij,t+1} \end{split} \tag{1}$$

- Estimate the model on a monthly frequency
- Adjust standard errors by Newey and West adjustment with 4 lags $(4(70/100)^{\frac{2}{9}}=3.69\sim4)$

Model Estimation

Normalized Rank-Transformed

		Dependent	Variable: Fu	ture Pairs's	co-movemer	it
	(1)	(2)	(3)	(4)	(5)	(6)
MFCAP*	0.00104*	0.000633			0.000166	-0.000146
	(2.44)	(1.51)			(0.38)	(-0.34)
Same Group			0.0182***	0.0169***	0.0168***	0.0182***
			(8.60)	(7.60)	(7.19)	(8.13)
Controls	No	Yes	No	Yes	Yes	Yes
PairType Control	No	No	No	No	No	Yes
Observations	1665996	1665996	1665996	1665996	1665996	1665996

t statistics in parentheses

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

Model Estimation

Normalized Rank-Transformed

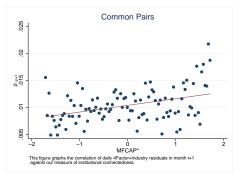
	Depe	ndent Variab	le: Future Pa	irs's co-move	ment
	(1)	(2)	(3)	(4)	(5)
MFCAP*	0.00946***	-0.000535	-0.000602	-0.000484	-0.000484
	(7.52)	(-1.24)	(-1.39)	(-1.16)	(-1.16)
$(MFCAP^*) \times SameGroup$			0.0109***	0.0110***	0.0110***
,			(10.22)	(9.49)	(9.49)
Same Group			0.00908***	0.00858***	0.00858***
			(4.00)	(3.61)	(3.61)
Sub-sample	SameGroup	Others	All	All	All
Business Group FE	No	No	No	Yes	Yes
Observations	58337	1607659	1665996	1665996	1665996

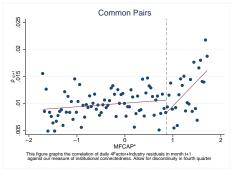
t statistics in parentheses

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Future Correlation via MFCAP

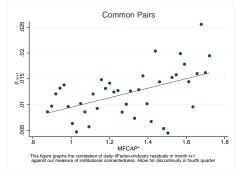
Discontinuity

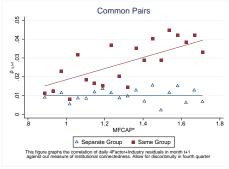




4 Factor + Industry Future Correlation via MFCAP*

Discontinuity & Business Groups





Fama-MacBeth Estimation

Discontinuity (sub-sample)

	Dependent	: Variable:	Future Pairs's	co-movement
	(1)	(2)	(3)	(4)
Same Group	0.0261***		-0.0280**	-0.0252*
	(8.83)		(-2.81)	(-2.38)
MFCAP*		0.00892*	-0.00407	-0.00353
		(2.60)	(-1.15)	(-1.02)
$(MFCAP^*) \times SameGroup$			0.0363***	0.0340***
			(5.03)	(4.33)
Controls	No	No	Yes	Yes
Business Group FE	No	No	No	Yes
Observations	417377	417377	417377	417377

t statistics in parentheses

 $^{^*}$ p < 0.05, ** p < 0.01, *** p < 0.001

All non-common owner pairs

regression

		De	ependent Var	iable: Future P	airs' co-move	ment	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
SameGroup	0.0156***		0.0158***			0.0138***	0.0131***
	(9.84)		(10.22)			(8.27)	(7.68)
MFCAP*		-0.0000723	-0.000277	0.00169	-0.000322*	-0.000390**	-0.000427
		(-0.44)	(-1.80)	(1.42)	(-2.19)	(-2.70)	(-2.29)
(MFCAP*) × SameGroup						0.00313**	0.00364**
						(2.80)	(3.34)
Sub-Sample	Total	Total	Total	SameGroups	Others	Total	Total
Business Group FE	No	No	No	No	No	No	Yes
Controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	6018646	6018646	6018646	114526	5904120	6018646	6018646

t statistics in parentheses

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

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TrunOver

$$\Delta \mathsf{TurnOver} = \mathsf{In}(\frac{\mathsf{TurnOver}_{i,t}}{\mathsf{TurnOver}_{i,t-1}}) = \mathsf{In}(\frac{\mathsf{volume}_{i,t}}{\mathsf{MarketCap}_{i,t}}) - \mathsf{In}(\frac{\mathsf{volume}_{i,t-1}}{\mathsf{MarketCap}_{i,t-1}})$$

TrunOver

$$\Delta \mathsf{TurnOver} = \mathsf{In}(\frac{\mathsf{TurnOver}_{i,t}}{\mathsf{TurnOver}_{i,t-1}}) = \mathsf{In}(\frac{\mathsf{volume}_{i,t}}{\mathsf{MarketCap}_{i,t}}) - \mathsf{In}(\frac{\mathsf{volume}_{i,t-1}}{\mathsf{MarketCap}_{i,t-1}})$$

Koch et al. (2016)

TrunOver

$$\Delta \mathsf{TurnOver} = \mathsf{In}(\frac{\mathsf{TurnOver}_{i,t}}{\mathsf{TurnOver}_{i,t-1}}) = \mathsf{In}(\frac{\mathsf{volume}_{i,t}}{\mathsf{MarketCap}_{i,t}}) - \mathsf{In}(\frac{\mathsf{volume}_{i,t-1}}{\mathsf{MarketCap}_{i,t-1}})$$

Koch et al. (2016)

	Depe	ndent Varia	ıble: ΔTurr	nOver _i
	(1)	(2)	(3)	(4)
Δ TurnOver _{Market}	0.457***	0.351***	0.182***	0.235***
	(4.04)	(10.69)	(3.42)	(4.72)
$\Delta TurnOver_{Industry-i}$	0.220***	0.159***	0.0528	0.117*
•	(4.28)	(4.10)	(1.03)	(2.37)
$\Delta TurnOver_{Group-i}$			0.286***	0.213***
			(6.21)	(5.15)
Portfo. Weight	-	-	MC	MC
Control	No	Yes	No	Yes
Observations	746640	742341	305563	301329
R^2	0.298	0.579	0.460	0.749

Aghajanzadeh, Heidari & Mohseni (TeIAS)

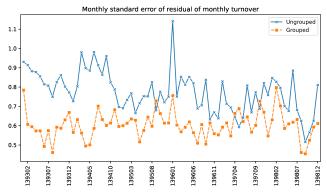
Residual of Monthly Turnover

- $\bullet \ \, \mathsf{Turnover}_{\mathit{i},\mathit{t}} = \alpha_{\mathsf{0}} + \alpha_{\mathsf{1}} \times \mathsf{Turnover}_{\mathit{i},\mathit{avg}} + \alpha_{\mathsf{2}} \times \mathsf{Turnover}_{\mathit{m},\mathit{t}} + \boxed{\varepsilon_{\mathit{i},\mathit{t}}}$
 - Turnover_{i,t}: Monthly Turnover (Average of daily turnovers in each month)
 - Turnover_{i,avg}: Annual average of monthly turnover
 - Turnover_{m,t}: Market's turnover

Grouped	$Firm \times Month$	mean	std	min	25%	50%	75%	max
Ungrouped Grouped	8050 18199	-0.001 0.001	0.822 0.777	-4.789 -4.832		-0.016 -0.033		4.407 4.955

Residual of Monthly Turnover

Grouped	$Group \times Month$	mean	std	min	25%	50%	75%	max
Ungrouped	72	0.776	0.108	0.516	0.694	0.774	0.840	1.140
Grouped	2393	0.604	0.300	0.001	0.413	0.580	0.763	2.797



Low residual standard error

	Deper	ndent Variab	le: Future P	airs's co-mov	/ement
	(1)	(2)	(3)	(4)	(5)
Same Group	0.0180***	0.0189***	0.0151***	0.0131***	0.0351***
•	(8.45)	(8.09)	(5.24)	(5.00)	(5.80)
LowResidualStd		0.00193 (1.99)	0.00159 (1.59)	-0.000678 (-0.84)	
$LowResidualStd \times SameGroup$			0.00981* (2.57)	0.0119** (3.37)	
${\sf Group} \; {\sf Turnover} \; {\sf std}_1$					-0.00119 (-0.64)
${\sf Group\ Turnover\ std}_2$					-0.00292* (-2.28)
${\sf Group\ Turnover\ std}\times{\sf SameGroup}$					-0.0261** (-3.02)
Observations	1665996	1665996	1665996	1665996	1665996
Group Size Effect	No	Yes	Yes	No	Yes
Group Effect	No	No	No	Yes	No
Pair Size FE	Yes	Yes	Yes	Yes	Yes
Sub-sample	Total	Total	Total	Total	Total
Controls	Yes	Yes	Yes	Yes	Yes
R^2	0.00120	0.00150	0.00161	0.00597	0.00173

t statistics in parentheses

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

Ins Imbalance

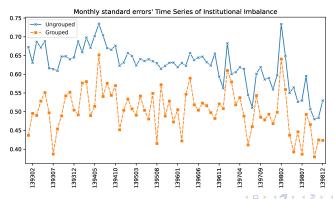
• Seasholes and Wu (2007)

$$Imbalance_{ins} = \frac{Buy_{ins} - Sell_{ins}}{Buy_{ins} + Sell_{ins}}$$

Grouped	$Firm \times Month$	mean	std	min	25%	50%	75%	max
Ungrouped Grouped	20197 12021				-0.474 -0.462	0.016 -0.009	0.479 0.341	1.0 1.0

Ins Imbalance std

	$Group \times Month$	mean	std	min	25%	50%	75%	max
Grouped								
Ungrouped	72	0.624	0.054	0.48	0.601	0.631	0.655	0.735
Grouped	2057	0.502	0.251	0.00	0.337	0.503	0.647	1.414



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Low Ins Imbalance Group

	Depe	ndent Varial	ole: Future Pa	airs's co-move	ment
	(1)	(2)	(3)	(4)	(5)
Same Group	0.0180***	0.0186***	0.00982***	0.00931***	0.135***
	(8.45)	(8.25)	(4.63)	(4.60)	(13.96)
Low Imbalance std		0.000899	0.0000151	0.000421	
		(0.98)	(0.02)	(0.45)	
Low Imbalance std × SameGroup			0.0235***	0.0228***	
			(9.05)	(8.46)	
Group Ins Imb std ₁					0.00168
•					(0.25)
Group Ins Imb std ₂					0.00986
					(1.65)
Group Ins Imb std × SameGroup					-0.198***
·					(-12.76)
Observations	1665996	1665996	1665996	1665996	599765
Group Size Effect	No	Yes	Yes	No	Yes
Group Effect	No	No	No	Yes	No
Pair Size FE	Yes	Yes	Yes	Yes	Yes
Sub-sample	Total	Total	Total	Total	Total
Controls	Yes	Yes	Yes	Yes	Yes
R^2	0.00120	0.00151	0.00163	0.00601	0.00357

t statistics in parentheses

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

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Conclusion

- Direct common ownership can affect firms' co-movement
- Firms in the business groups co-move more than other pairs
- Direct common ownership only matters for firms in the business groups
- Firms in the same business group trade together

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 - Synchronicity and firm interlocks
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- If two stocks in pair have n mutual owner, which total market cap divides them equally, the mentioned indexes equal n.
 - Each holder owns 1/n of each firm.
 - Firm's market cap is α_1 and α_2 :
 - So for each holder of firms we have $S_{i,t}^f P_{i,t} = \alpha_i$
 - SQRT

$$\left[\frac{\sum_{f=1}^{n} \sqrt{\alpha_1/n} + \sum_{f=1}^{n} \sqrt{\alpha_2/n}}{\sqrt{\alpha_1} + \sqrt{\alpha_2}}\right]^2 = \left[\frac{\sqrt{n}(\sqrt{\alpha_1} + \sqrt{\alpha_2})}{\sqrt{\alpha_1} + \sqrt{\alpha_2}}\right]^2 = n$$

Quadratic

$$\left[\frac{\sum_{f=1}^{n} (\alpha_1/n)^2 + \sum_{f=1}^{n} (\alpha_2/n)^2}{\alpha_1^2 + \alpha_2^2}\right]^{-1} = \left[\frac{\alpha_1^2 + \alpha_2^2}{n(\alpha_1^2 + \alpha_2^2)}\right]^{-1} = n$$





Anton and Polk (2014)

$$FCAP_{ij,t} = \frac{\sum_{f=1}^{F} (S_{i,t}^{f} P_{i,t} + S_{j,t}^{f} P_{j,t})}{S_{i,t} P_{i,t} + S_{j,t} P_{j,t}}$$

Anton and Polk (2014)

$$FCAP_{ij,t} = \frac{\sum_{f=1}^{F} (S_{i,t}^{f} P_{i,t} + S_{j,t}^{f} P_{j,t})}{S_{i,t} P_{i,t} + S_{j,t} P_{j,t}}$$

SQRT

$$\left[\frac{\sum_{f=1}^{F}(\sqrt{S_{i,t}^{f}P_{i,t}}+\sqrt{S_{j,t}^{f}P_{j,t}})}{\sqrt{S_{i,t}P_{i,t}}+\sqrt{S_{j,t}P_{j,t}}}\right]^{2}$$

Quadratic

$$\left[\frac{\sum_{f=1}^{F}[(S_{i,t}^{f}P_{i,t})^{2}+(S_{j,t}^{f}P_{j,t})^{2}]}{(S_{i,t}P_{i,t})^{2}+(S_{j,t}P_{j,t})^{2}}]^{-1}$$

Anton and Polk (2014)

$$FCAP_{ij,t} = \frac{\sum_{f=1}^{F} (S_{i,t}^{f} P_{i,t} + S_{j,t}^{f} P_{j,t})}{S_{i,t}P_{i,t} + S_{j,t}P_{j,t}}$$

SQRT

Quadratic

$$\frac{\left[\frac{\sum_{f=1}^{F}(\sqrt{S_{i,t}^{f}P_{i,t}}+\sqrt{S_{j,t}^{f}P_{j,t}})}{\sqrt{S_{i,t}P_{i,t}}+\sqrt{S_{j,t}P_{j,t}}}\right]^{2}}{\sqrt{S_{i,t}P_{i,t}}+\sqrt{S_{j,t}P_{j,t}}}]^{2}$$

$$\left[\frac{\sum_{f=1}^{F}[(S_{i,t}^{f}P_{i,t})^{2}+(S_{j,t}^{f}P_{j,t})^{2}]}{(S_{i,t}P_{i,t})^{2}+(S_{j,t}P_{j,t})^{2}}\right]^{-1}$$

Intuition

If for a pair of stocks with n mutual owners, all owners have even shares of each firm's market cap, then the proposed indexes will be equal to n. $\frac{1}{2}$

Example



Example



For better observation, assume that

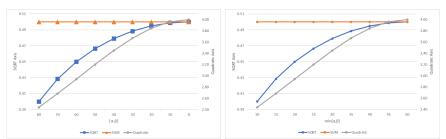
- $\alpha + \beta = 100$
- both firm have equal market cap

Example



For better observation, assume that

- $\alpha + \beta = 100$
- both firm have equal market cap



Comparison of three methods for calculating common ownership

Example of three common owner



Firm X

Example of three common owner

Common owner 1

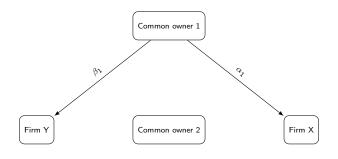
 $\mathsf{Firm}\ \mathsf{Y}$

Common owner 2

Firm X

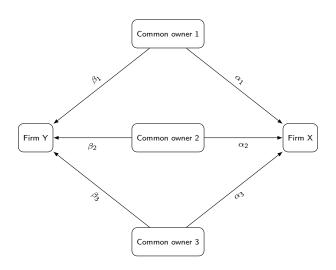
Common owner 3

Example of three common owner



Common owner 3

Example of three common owner



Example of three common owner

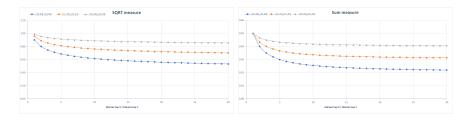
Ownership	Type I	Type II	Type III	Type IV	Type V	Type VI	Type VII
α_1	1/3	20	10	20	10	5	1
eta_1	1/3	10	10	20	10	5	1
α_2	1/3	10	80	20	10	5	1
β_2	1/3	20	80	20	10	5	1
α_3	1/3	70	10	20	10	5	1
eta_3	1/3	70	10	20	10	5	1
SQRT	3	2.56	2.33	1.8	0.9	0.45	0.09
SUM	1	1	1	0.6	0.3	0.15	0.03
Quadratic	3	1.85	1.52	8.33	33.33	133.33	3333.33

Comparison

- For better comparison we relax previous assumptions:
 - Two Firms with different market caps.

	$(\alpha_1,\beta_1),(\alpha_2,\beta_2)$							
	(10,40)	(10,40)	(15,35)	,(15,35)	(20,30),(20,30)			
$\frac{MarketCap_{X}}{MarketCap_{y}}$	SQRT	SUM	SQRT	SUM	SQRT	SUM		
1	0.90	0.50	0.96	0.50	0.99	0.50		
2	0.80	0.40	0.89	0.43	0.96	0.47		
3	0.75	0.35	0.85	0.40	0.94	0.45		
4	0.71	0.32	0.83	0.38	0.92	0.44		
5	0.69	0.30	0.81	0.37	0.91	0.43		
6	0.67	0.29	0.80	0.36	0.91	0.43		
7	0.65	0.28	0.79	0.35	0.90	0.43		
8	0.64	0.27	0.78	0.34	0.90	0.42		
9	0.63	0.26	0.77	0.34	0.89	0.42		
10	0.62	0.25	0.76	0.34	0.89	0.42		

Comparison



Comparison of two methods for calculating common ownership

Conclusion

We use the SQRT measure because it has an acceptable variation and has fair values at a lower level of aggregate common ownership.

Common Ownership measure

	Dependent Variable: Future Monthly Correlation of 4F+Industry Residuals									
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Common Ownership Measure	0.00177***	0.00150**	0.00133**	0.00102	0.000936	0.000663	0.000536	0.000377	-0.0000197	-0.0000113
	(3.93)	(2.90)	(2.76)	(1.87)	(1.90)	(1.17)	(1.06)	(0.65)	(-0.04)	(-0.02)
Same Group			0.0156***	0.0157***	0.00774***	0.00813***	0.00575*	0.00624**	0.00503*	0.00549*
			(7.32)	(7.44)	(3.61)	(3.71)	(2.62)	(2.81)	(2.11)	(2.27)
Common Ownership Measure × SameGroup					0.0103***	0.00935***	0.0110***	0.00992***	0.0119***	0.0107***
					(7.76)	(6.72)	(7.47)	(6.49)	(7.94)	(6.97)
SameIndustry							-0.000364	-0.000312	0.000286	0.000339
,							(-0.21)	(-0.19)	(0.17)	(0.21)
SameSize							0.0133***	0.0135***	0.0131***	0.0132***
							(4.48)	(4.56)	(4.61)	(4.68)
SameBookToMarket							0.00772***	0.00772***	0.00893***	0.00893***
							(4.55)	(4.58)	(5.05)	(5.09)
CrossOwnership							0.0280*	0.0260	0.0303*	0.0283*
,							(2.07)	(1.93)	(2.27)	(2.14)
Observations	1665996	1665996	1665996	1665996	1665996	1665996	1665996	1665996	1665996	1665996
Group FE	No	No	No	No	No	No	No	No	Yes	Yes
Measurement	Sum	Quadratic	Sum	Quadratic	Sum	Quadratic	Sum	Quadratic	Sum	Quadratic
R ²	0.000171	0.000170	0.000348	0.000349	0.000443	0.000437	0.000898	0.000898	0.00575	0.00575

t statistics in parentheses

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

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Main Effect

Common-ownership and comovement effect

[Anton and Polk (2014)]

Stocks sharing many common investors tend to comove more strongly with each other in the future than otherwise similar stocks.

• Common-ownership and liquidity demand

[Koch et al. (2016), Pastor and Stambaugh (2003), Acharya and Pedersen (2005)] Commonality in stock liquidity is likely driven by correlated trading among a given stock's investors. Commonality in liquidity is important because it can influence expected returns

Trading needs and comovement

[Greenwood and Thesmar (2011)]

If the investors of mutual funds have correlated trading needs, the stocks that are held by mutual funds can comove even without any portfolio overlap of the funds themselves

Stock price synchronicity and poor corporate governance

[Boubaker et al. (2014), Khanna and Thomas (2009), Morck et al. (2000)] Stock price synchronicity has been attributed to poor corporate governance and a lack of firm-level transparency. On the other hand, better law protection encourages informed trading, which facilitates the incorporation of firm-specific

Stock price synchronicity has been attributed to poor corporate governance and a lack of firm-level transparency. On the other hand, better law protection encourages informed trading, which facilitates the incorporation of firm-specific information into stock prices, leading to lower synchronicity



Synchronicity and firm interlocks

JFE-2009-Khanna

- Three types of network
 - Equity network
 - 2 Director network
 - Owner network
- Dependent variables

Using deterended weekly return for calculation

- 1 Pairwise returns synchronicity = $\frac{\sum_{t} (n_{i,j,t}^{up} n_{i,j,t}^{down})}{T_{i,j}}$
- 2 Correlation = $\frac{Cov(i,j)}{\sqrt{Var(i).Var(j)}}$
- Tobit estimation of

$$f_{i,j}^d = \alpha I_{i,j} + \beta (1 * N_{i,j}) + \gamma Ind_{i,j} + \varepsilon_{i,j}$$

being in the same director network has a significant effect

Large controlling shareholder and stock price synchronicity JBF-2014-Boubaker

Stock price synchronicity:

$$SYNCH = \log(\frac{R_{i,t}^2}{1 - R_{i,t}^2})$$

where $R_{i,t}^2$ is the R-squared value from

$$\textit{RET}_{\textit{i},\textit{w}} = \alpha + \beta_1 \textit{MKRET}_{\textit{w}-1} + \beta_2 \textit{MKRET}_{\textit{w}} + \beta_3 \textit{INDRET}_{\textit{i},\textit{w}-1} + \beta_4 \textit{INDRET}_{\textit{i},\textit{w}} + \varepsilon_{\textit{i},\textit{w}}$$

OLS estimation of

$$\begin{aligned} \textit{SYNCH}_{i,t} &= \beta_0 + \beta_1 \textit{Excess}_{i,t} + \beta_2 \textit{UCF}_{i,t} + \sum_k \beta_k \textit{Control}_{i,t}^k \\ &+ \textit{IndustryDummies} + \textit{YearDummies} + \varepsilon_{i,t} \end{aligned}$$

- Stock price synchronicity increases with excess control
- Firms with substantial excess control are more likely to experience stock price crashes

Connected Stocks

JF-2014-Anton Polk

- Common active mutual fund owners
- Measuring Common Ownership

•
$$FCAP_{ij,t} = \frac{\sum_{f=1}^{F} (S_{i,t}^{f} P_{i,t} + S_{j,t}^{f} P_{j,t})}{S_{i,t} P_{i,t} + S_{j,t} P_{j,t}}$$

- ullet Using normalized rank-transformed as $FCAP_{ij,t}^*$
- $\rho_{ij,t}$: within-month realized correlation of each stock pair's daily four-factor returns

$$\rho_{ij,t+1} = a + b_f \times \textit{FCAPF}^*_{ij,t} + \sum_{k=1}^{n} \textit{CONTROL}_{ij,t,k} + \varepsilon_{ij,t+1}$$

Estimate these regressions monthly and report the time-series average as in Fama-MacBeth

Commonownership measurements

Model-based measures

- lacktriangledown $\mathrm{HJL}_I^A(A,B) = \sum_{i\in I^A,B} rac{lpha_{i,B}}{lpha_{i,A}+lpha_{i,B}}$ Harford et al. (2011)
 - Bi-directional
 - Pair-level measure of common ownership
 - Its potential impact on managerial incentives
 - Measure not necessarily increases when the relative ownership increases
 - Accounts only for an investor's relative holdings
- lacktriangledown MHHI $=\sum_{j}\sum_{k}s_{j}s_{k}rac{\sum_{i}\mu_{ij}
 u_{ik}}{\sum_{i}\mu_{ij}
 u_{ij}}$ Azar et al. (2018)
 - Capture a specific type of externality
 - Measured at the industry level
 - Assumes that investors are fully informed about the externalities
- \bullet GGL^A $(A,B) = \sum_{i=1}^{I} \alpha_{i,A} g(\beta_{i,A}) \alpha_{i,B}$ Gilje et al. (2020)
 - Bi-directional
 - Less information
 - Not sensitive to the scope
 - Measure increases when the relative ownership of firm A increases



Commonownership measurements

Ad hoc common ownership measures

- $Overlap_{Count}(A, B) = \sum_{i \in I^{A,B}} 1$ He and Huang (2017),He et al. (2019)
- $Overlap_{Min}(A,B) = \sum_{i \in I^{A,B}} min\{\alpha_{i,A},\alpha_{i,B}\}$ Newham et al. (2018)
- Overlap_{AP}(A, B) = $\sum_{i \in I^{A,B}} \alpha_{i,A} \frac{\bar{\nu}_A}{\bar{\nu}_A + \bar{\nu}_B} + \alpha_{i,B} \frac{\bar{\nu}_B}{\bar{\nu}_A + \bar{\nu}_B}$ Anton and Polk (2014)
- $Overlap_{HL}(A,B) = \sum_{i \in I^{A,B}} \alpha_{i,A} \times \sum_{i \in I^{A,B}} \alpha_{i,B}$ Hansen and Lott Jr (1996) , Freeman (2019)
- Unappealing properties
 - Unclear is whether any of these measures represents an economically meaningful measure of common ownership's impact on managerial incentives.
 - Both Overlap_{Count} and Overlap_{AP} are invariant to the decomposition of ownership between the two firms, which leads to some unappealing properties.



