

JACOBS LEVY EQUITY MANAGEMENT CENTER

Are Some Clients More Equal Than Others? Evidence of Price Allocation by Delegated Portfolio Managers (with Ryan D. Israelsen)

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MOTIVATION

Management companies have diversified sets of clients: Defined Benefit and Defined Contribution plans, endowments, insurance companies, mutual funds, and high-net-worth individuals

Investment decisions that maximize the value of the management company as a whole *may differ* from those that maximize the values of individual clients

Do management companies *display favoritism* toward some of their clients (or products) at the expense of others?





MOTIVATION (CONT'D)

Which products may be more important?

- "Star products" high profile products positive spillover effect (e.g., Massa (2003), and Nanda, Wang, and Zheng (2004))
- Young products more sensitive to early performance (e.g., Chevalier and Ellison (1997))
- Products / clients with different attention households vs. institutional investors
- Client power direct (size) or indirect (connections)

Possible Benefits?

- Increase in AUM (e.g. "flow convexity" Brown, Harlow and Starks (1996), Chevalier and Ellison (1997), Sirri and Tufano (1998))
- Reputation (e.g., Khorana and Servaes (1999))



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Gaspar, Massa and Matos (2006) - *mutual fund families* - evidence of strategic performance allocation toward "high family value" funds such as *star funds, high fee funds and young funds*

Chaudhuri, Ivkovic and Trzcinka (2013) - *institutional money management products* - evidence of strategic performance allocation toward *star products* and *young products* which varies with *client power*

Goncalves-Pinto and Schmidt (2013) – *co-insurance* - mutual fund families *coordinate internal trades* in order to protect member funds that are suffering heavy redemptions

Compelling Results



There at least **two reasons** why further investigation is needed:

1. Data limitations

- Due to the lack of availability of transaction-level data, the analysis is usually conducted using returns
- Aggregated across time and across securities

2. Mechanism

- Two main channels of performance allocations suggested in the literature → IPO allocations and cross trading
- It is not clear whether:
 - they occur frequently enough to explain the observed transfer of performance
 - there are other, previously unidentified channels



We have data which allow us to *directly test* the existence of such a behavior using *daily trades*

13F management companies trades on behalf of their institutional clients (provided by Ancerno Ltd. / Abel Noser):

- We know that management company X traded for clients A, B, C using brokerage firm Y
- Delegated portfolios \rightarrow management companies make the call
- 1. Provide *direct evidence* consistent with strategic performance allocation
- 2. Introduce *a new mechanism* that was ignored / couldn't be tested



APPROACH (CONT'D)

Management companies' ADV filings:

It is convenient to bunch similar trades across clients - Trading Desk

- Single or multiple brokers depending on trade size
- Same price (SP) or different prices
- Shouldn't expect to find systematic differences between clients

We specifically *target "bunched" trades*

→Trades by the same management company, same day, same stock, same trading direction for more than one client

Why?

- High degree of overlapping trades → an integral part of the daily trading activity (compared to Cross Trading and IPO allocation)
- Control for unobservable variables such as stock picking ability, broker talent and trading desk skills



Manager	DATE	Stock	Client	NumTRD	Num SHR	\$ VOL	PRC
MGR1	1/1/2010	S1	1	1	500	23,510	47.02
MGR1	1/1/2010	S1	2	1	500	23,530	47.06
MGR1	1/1/2010	S1	3	1	500	23,530	47.06
MGR1	1/1/2010	S1	4	1	1,000	47,080	47.08
MGR1	1/1/2010	S1	5	2	<u>2,000</u>	<u>94,120</u>	<u>47.06</u>
Same Price	Benchmark				4,500	211,770	47.06

Client 1's PTV (in %) = <u>[500 * (47.02 - 47.06)</u> *<u>1*-1</u> (\$ 23,510)

= 0.085%





APPROACH (CONT'D)

1. Compare the prices that clients receive to the *Same price benchmark*

- 2. Calculate their hypothetical % gains or losses per \$ trade volume and create monthly *PTV* averages
- 3. Explore whether these differences are systematic
- 4. Test our hypotheses





HYPOTHESES

H1 - Null Hypothesis: there are no *systematic* differences in prices across clients

\rightarrow If rejected?

H2 – the **SPA** hypothesis (**S**trategic **P**erformance **A**llocation or "Favoritism"): Systematic differences across clients are driven by strategic performance allocation

H3 – the **D**ifferent **T**rading **P**ractices (Alternative) Hypothesis: Systematic differences across clients are driven by different trading practices (and not favoritism).





Step 1 – H1 – Existence of price differences

1.1 Do some clients receive *systematically* better (worse) prices?

1.2 What is the economic magnitude?

Step 2 – SPA and DTP hypotheses

H2:

2.1 What are the *characteristics* of management companies and Clients likely to be involved?

2.2 What are the *direct benefits* to the favoring managers and benefited clients?

H3:

3. Alternative explanations which are consistent with the different trading practices hypothesis



1. Does it happen?

1.1 *Systematic differences* between clients in general, and between clients *within* management companies

1.2 *Strong evidence* of out-of-sample persistence in price allocation for a subset of management companies

1.3 The average magnitude can be as large as 0.50% of \$ trade volume



SUMMARY OF MAIN RESULTS (CONT'D)

2. Characteristics and benefits

2.1 Characteristics:

- Management companies with more *trade opportunities*
- Clients with more *trade opportunities* and *higher attention*

2.2 Benefits:

- Managers' direct benefits \rightarrow *increase in volume* by the favored clients
- Clients' direct benefits \rightarrow trading alpha of 15 bps per month

3. Alternative explanations

Directed Brokerage Arrangements, Price Impact, Trade Commissions, Trading Style, Fill Ratios



ANCERNO's institutional trading data by delegated portfolio managers (no holdings) from 1999 (Q1) -2011(Q3)

Main variables:

 Number of shares, buy/sell indicator, execution price, trade commissions, Cusip and ticker

Other identifiers:

- Client type → Ancerno's clients are mainly *pension plan sponsors* and *mutual fund families*
- Unique client codes, unique management company codes, unique broker codes
- We received *linking codes* from Ancerno which enable us to link institutional clients to their management companies and brokerage firms



A RICH LINK STRUCTURE







Mgr_m - Cnt_c - Day_t - Stock_s → 6,125,500 daily trades in our "bunched" sample (50% (EW) / 25% (VW) of all Cnt-Mgr trades) *these ratios are much higher for the "significant clients"

 $Mgr_m - Day_t - Stock_s \rightarrow 1,938,500$ unique trades ~ an average of 3.20 clients per bunched trade

488 managers (13F), 825 clients, 5,144 Manager-Client pairs \rightarrow multiple links

Avg. of 27 months of bunched trading activity per Mgr-Cnt pair



Monthly cross-sectional statistics \rightarrow time-series averages

Table 1			
Variables	Mean	Median	SD
Cnt-Per-Mgr	5.16	3.47	4.83
Mgr-Per-Cnt	3.45	2.69	2.97
Num-Trd-In-Mon	46.50	19.81	83.81
Diff-Stocks-Shared-In-Mon	21.25	10.66	36.04
Overlap-Ratio	83.84	100.00	27.06
Overlap-Ratio - VW	42.01	35.24	N/A
Num-Partial-Trds-By-Cnt	5.65	1.07	15.95

*Elton, Gruber and Green (2007), and Blocher (2011)



For each Manager-Client pair we calculate the *time series average* of the *monthly PTV series*

Table 2

Frequency	6 and above			
<u>P-value</u>	10% 5% 1%			
Num C-M Pairs	3827	3827	3827	
% Sig Nominal P-values	17.82%	10.72%	4.94%	
% Sig Simulated P-Values	15.56%	10.24%	3.53%	
Num Sig Pos	356	230	77	
Num Sig Neg	240	162	59	
Num Sig Pos-Neg Ratio	1.49	1.42	1.31	

*Randomly reshuffle the clients in each Manager-Day-Stock bunched trade, repeat the calculation and store the simulated p-value \rightarrow 10,000 times



- 1. Economic magnitude *conditioning on the clients' time-in-sample*
- 2. Economic magnitude *conditioning on trades with more opportunities* [> Ave (H-L)] H-L % spread = [(Highest Client Price – Lowest Client Price) / VW Price]

Table 3	Signifi	Significant Positive Clients			<u>ents</u> Signi			egative	Clients
	ALL > Ave H-L		e H-L	ALL		L	> Ave H-L		
Time in Sample	Ave	SD	Ave	SD		Ave	SD	Ave	SD
1-6 months	0.137	0.325	0.278	0.373	<u>ا</u>	-0.121	0.252	-0.306	0.391
7-12 months	0.124	0.254	0.269	0.349		-0.125	0.195	-0.271	0.241
13-24 months	0.068	0.115	0.199	0.229		-0.058	0.076	-0.202	0.178
25-36 months	0.062	0.073	0.189	0.214		-0.080	0.093	-0.209	0.221
37-48 months	0.053	0.054	0.138	0.100		-0.088	0.149	-0.187	0.200
49-60 months	0.059	0.144	0.154	0.154		-0.045	0.044	-0.131	0.139
More than 60 months	0.027	0.035	0.104	0.092		-0.033	0.035	-0.100	0.072

1. Magnitudes decline with time-in-sample ightarrow

- Incentives to subsidize a favored client are strong when these clients are new
- Avoid exploiting specific clients for extended periods of time
- 2. More opportunities \rightarrow larger transfers



For each *management company* we calculate the difference between the *top and bottom* clients (based on monthly PTV averages) \rightarrow In Sample

Table 4				
Frequency	6 and above			
P-value	10%	5%	1%	
Num Mgrs	361	361	361	
Nominal P-values	42.38%	26.59%	13.29%	
Simulated P-Values	19.94%	14.40%	4.43%	
Num Managers - SimPval	72	52	16	

*Randomly reshuffle the clients in each Manager-Day-Stock bunched trade, repeat the calculation and store the simulated p-value \rightarrow 10,000 times





OUT-OF-SAMPLE PERSISTENCE – TEST 1

For each month *m* and management company *j*, we use rolling months *m*-12 to *m*-1:

- Rank clients into PTV Ranking-Quartiles (Bot, 2, 3, and Top) ٠
- Use the difference between *Top and Bottom Quartiles to* define the *Sig-Mgrs* • and Non-Sig-Mgrs groups
- Re-rank clients into Post-Ranking Quartiles in month m •







 $1 \rightarrow 1$

 $4 \rightarrow 4$

 $1 \rightarrow 2.5$

 $4 \rightarrow 2.5$

Split each Client (within a management company) into *2 equal sub-periods* Determine the significant management companies using the *first period*

Table 6			
	Non-Sig	Sig	Sig HL
MinFreq	6	6	6
Ranking period			
Top Average	0.173	0.174	0.277
Bot Average	-0.121	-0.167	-0.337
Post Ranking period			
Top Average	-0.002	0.069	0.278
T-stat	0.19	4.06	3.63
Bot Average	0.017	-0.097	-0.182
T-stat	1.10	3.94	5.41
Post Top-Bottom Diff	-0.019	0.165	0.460
T-stat	1.02	5.55	5.50
Persistence Ratio Top	-1.1%	39%	100%
Persistence Ratio Bot	-13.8%	58%	54%





Table /	Variables		(4)	(5)
1.6:- MCD-	LnCnt-Per-Mgr	*	0.029	
1 SIG MIGRS			0.54	
0 Non-Sig MGRs	LnMgr-Per-Cnt	*	-0.179	
			3.98	
	LnMgr-Cnt-Shrd-Vol	*	0.020	0.023
			1.80	1.95
	LnOverlap-Ratio	*	0.128	0.086
			2.88	2.21
	LnNum-FF48-Ind	*	0.14	0.13
		بلد	10.40	9.08
	Cnt-Per-Mgr	*		0.10
		*		2.21
	Cht-Per-Wigr2			-0.01
	Mar Por Cot	*		2.15
	Mgi-rei-Citt			3.09
	Mar-Per-Cnt2	*		-0.01
	Wight Chief Chiez			3 57

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Fama-MacBeth Probit Models (153 Mon) at the Mgr-Month level



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PREDICTED PROBABILITIES OF BEING IN THE SIGNIFICANT MANAGER GROUP

0.18 0.16 0.14 **Predicted Probability** 0.12 0.1 0.08 0.06 0.04 0.02 0 1 3 11 13 15 17 19 21 23 25 27 29 31 33 35 37 39 5 9 Number of Clients per Manager

Number of Clients per Manager

We set the **control variables to their means** and **vary our variable of interest** based on the sample range





PREDICTED PROBABILITIES OF BEING IN THE SIGNIFICANT MANAGER GROUP

Number of Managers per Client



We set the **control variables to their means** and **vary our variable of interest** based on the sample range





CHARACTERISTICS OF SIGNIFICANT CLIENTS

Fama-MacBeth Probit Models (153 Mon) at the Cnt-Mgr-Month level + MGR DUM

Table 8				
	Posi	itive	Nega	ative
Variables	(1)	(4)	(6)	(9)
LnCnt-Trd-Relative-Vol	-0.103	-0.138	-0.072	-0.102
	3.37	4.06	2.07	2.14
Mgr-Per-Cnt	0.080	0.076	-0.053	-0.074
	5.07	4.77	3.31	4.27
Mgr-Per-Cnt2	-0.005	-0.005	0.003	0.004
	5.19	4.86	3.00	3.85
LnOverlap-Ratio	0.296	0.348	0.010	0.090
	3.34	3.74	0.13	1.28
*		*		*
*		*		*
HBAS		2.956		8.734
		1.96		3.50
SD		3.986		8.135
		3.83		3.31
*		*		*
Mgr Dummies	YES	YES	YES	YES
SMP	76,953	76,953	57,754	57,754
N	153	153	153	153

1 Sig CNTs 0 Non-Sig CNTs

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PREDICTED PROBABILITIES OF BEING A SIGNIFICANT POSITIVE OR SIGNIFICANT NEGATIVE CLIENT

Number of Managers per Client



Positive Clients

We set the **control variables to their means** and **vary our variable of interest** based on the sample range



Negative Clients



BENEFITS

Management Companies – Increase in Volume

Using our sub-period analysis:

- A significant increase of **15%-30%** in trading volume for top clients within the significant management companies
- Insignificant differences for the non-significant managers group

Clients – Positive Trading Alpha

For each Client-Manager pair we calculate a monthly "*Trade Gain*" measure, and find:

- *Positive and significant* difference in performance of 0.15% per month (t-stat of 1.98) between the positive-significant clients and their counterparts
- *Small and insignificant* difference in performance of -0.02% per month (t-stat of 0.27) between the negative-significant clients and their counterparts



"Directed brokerage arrangements" - clients may direct the manager to execute their trades with specific brokers → may not be able to deliver the best execution price

2,478,678 unique $Mgr_m - Cnt_c - Day_t - Stock_s$ (BKR_b) "bunched" trades ~ 40% of the sample



Figure 4.A Out-of-Sample Ranking Quartiles





"Dynamic compensation schemes" - management companies may take trade commissions into account

 \rightarrow Clients who pay higher commissions may be compensated through better execution prices

Compare the Trade Commissions (in %) and PTV averages

For each management company \rightarrow

Rank Com Ave \rightarrow PTV Ave

Rank PTV Ave \rightarrow Com Ave

Та	b	e	1	1
		_	_	_

	ALL Mgr		Sig	Mgr
Groups	Com	Ptv	Com	Ptv
Com 1 - Bot	0.077	0.007	0.113	0.002
Com 2	0.141	0.001	0.172	0.007
Com 3 - Top	0.295	0.008	0.289	0.005
Top - Bottom <i>t</i> -statistic	0.218	0.001 0.21	0.176	0.004 0.45

	ALL Mgr		SigN	Лgr
Groups	Ptv	Com	Ptv	Com
Тор	0.124	0.140	0.128	0.140
Bot	-0.086	0.140	-0.108	0.142
Tom-Bot	0.210	0.000	0.236	-0.001
t-statistic	5.77	0.04	6.69	0.21





Client heterogeneity within management company might lead to different execution practices

1. Trade Size and Price Impact:

Clients who are allocated larger quantities may *mechanically* be allocated worse prices (be last in line)

 \rightarrow Conditioning on the management company – the correlation between *trade size* and *execution price* is not significant

2. Different portfolios / Unique Strategies:

Clients whose overall portfolios *differ* may receive *different attention* within a given bunched trade

Use Daniel, Grinblatt, Titman and Wermers (1997) ranking scores and Anand, Irvine, Puckett and Venkataraman (2013) Trading Style measure

→We *do not find* statistically significant differences between *significant and non-significant* clients (within a management company)



3. Different Fill Ratios

Different execution practices may lead to different fill rates For example:

- Client 1 may have a 90% fill rate on *day 0* and 10% on the following day
- Client 2 may have a 50% fill rate on *day 0* and 50% on the following day

 \rightarrow Such a difference may suggest that the clients are different in their trading needs

We *do not find* statistically significant differences between *significant and non-significant* clients (within a management company)



CONCLUSION

Contribution

- 1. Use *trade-level data* to provide *direct evidence* of price allocation by a subset of delegated portfolio managers
- 2. Reveal a *new mechanism* that was ignored/couldn't be tested before

Overall

- 1. Strong evidence which indicates that there are systematic differences across clients for a subset of management companies
- Magnitudes are economically significant can be as large as 0.50% of \$ trade volume → lower bound
- 3. Explore the characteristics of the management companies and clients likely to be involved, provide evidence of the benefits
- 4. Rule out alternative explanations

Future Research

Other channels - e.g., trade allocation

