## The Dollar Profits to Insider Trading

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#### Motivation I – Common prior

Corporate insiders generate substantial abnormal returns

Study	Estimate of abnormal return
Cicero and Wintoki 2015	2.5% for purchases (1m)
Cohen, Malloy, and Pomorski JF 2012	9.8% annualized value-weighted
Wang, Shin, and Francis JFQA 2012	3.9% (2.1%) for CEO and CFO purchases (sales) (3m)
Jeng, Metrick, and Zeckhauser REStat 2003	6% for purchases (12m)
Lakonishok and Lee RFS 2001	4.8% for buy-sell long-short pf (12m)

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#### Motivation II – Returns ≠ profits

Corporate insiders generate substantial abnormal returns

**BUT** 

➤ Returns (alpha) can differ from extracted economic value (e.g. Berk and van Binsbergen 2015)

#### This paper:

- Returns ≠ profits when trade size and frequency are choice variables
- Can dollar profits tell us more about insider trading?

#### Motivation III – Why care about dollar profits?

#### Theory I: Informed (insider) trading

- Quantities chosen strategically to balance costs and benefits (e.g. Kyle 1985, Huddart et al. 2001, Lenkey 2014)
  - + How much the insider cares about trading profits
  - Insider's concern for adverse selection costs or litigation risk
- These determinants can create a wedge between returns and profits

#### Theory II: Agency

- Insiders' returns as a measure of opportunism (e.g. Ali and Hirshleifer 2017)
- But individual utility more strongly linked to \$ profits than % returns
- Dollar profits = f(abnormal return, trade quantities)
- Joint distribution of returns and trade quantities unknown

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#### Research questions

1. Given their superior information, how much value (dollar profits) do insiders extract?

- 2. What drives dollar profits and trade quantities?
  - a) Do predictors of returns also predict trade quantities and profits?
  - b) Trading intentions
  - c) Monitoring

#### Contribution

- 1. Literature focuses on percentage returns
  - First to analyze trade quantities and dollar profits

- Insider trading as source of private benefits/compensation?
   (Manne 1996, Hue and Noe 2001, Roulstone 2003, Henderson 2011, Denis and Xu 2013, Cziraki et al. 2014)
  - First to use short-swing rule to capture trading intentions
  - Are profits large?

- Can monitoring restrain insider trading?
  - Depends on how do trade quantities and profits respond

#### Preview of findings

# Using \$ profits vs. % returns offers contrasting evidence on a number of important questions

- Typical dollar profits are small
- Informed trading proxies predict returns, but not profits
  - Proxies are negatively correlated with quantities
  - Frequency is first-order determinant of profits
- Sole exception: new proxy of trading intentions based on trading around the short-swing rule threshold
  - Still, even profits of these insiders remain modest
- Different insiders respond differently to increase in monitoring

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#### Insider trading universe

• Insider trading data from Thomson Reuters spanning 1986 to 2013

Transactions	644,643
Buys	148,363
Sells	496,280
Insider-years	263,413
Firm-years	52,602
Unique insiders	92,758
Unique firms	7,643

Aggregate trades by insider-day

#### Calculating dollar profits – main measure

Insider trade



Dollar profit( $t_1,t_2$ ) = return( $t_1,t_2$ ) × value traded

Subtract benchmark, e.g. FF3, to obtain abnormal profit

Abnormal dollar profit( $t_1,t_2$ ) = abnormal return( $t_1,t_2$ ) × value traded

- Use window of (0,20): common in literature
- Potential profit, insider does not necessarily pocket this
- Sample selection: (1) if potential profit is negative, wait for price to adjust, (2) some insiders do not close trades at all

#### Calculating dollar profits – alternative measure



Dollar profit( $t_1, t_2$ ) = return( $t_1, t_2$ ) × value traded

• Return( $t_1, t_2$ ) =  $(p_2 - p_1) / p_1$ 

#### **Properties**

- ✓ Actual profit, insider does pocket this
- $\star$  How to calculate  $p_1$  if sale is preceded by multiple purchases?
- Can insiders profit from price declines?

## **Summary statistics**

Variable	Mean	Sd	p10	p50	p90
Yr frequency	2.8	5.4	1	1	5
Yr value traded	1,845	8,984	14	232	3,483
Abnormal return (%)	0.9	10.8	-10.9	0.6	12.9
Abnormal profit	4.2	85	-31	0.141	41
Yr abnormal profit	12	182	-46	0.464	76
Yr abnormal round-trip profit	125	981	-106	5	354
Profits/compensation (%)	1.1	11.1	-4.1	0.1	6.7
Firm-year level abnormal profit	61	621	-189	3.349	397
Insider-lifetime abnormal profi	35	413	-72	1.095	157

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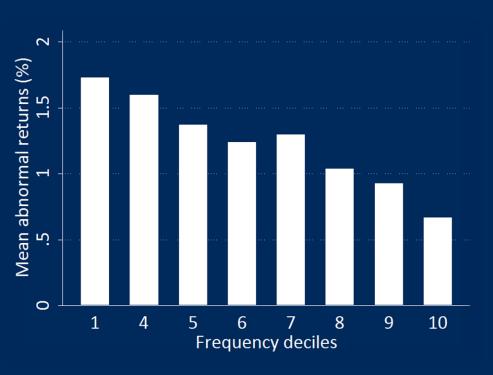
Insider-lifetime frequency

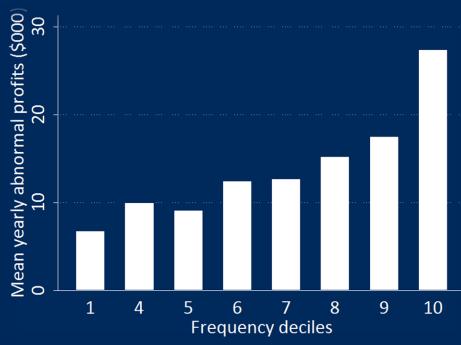
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## Infrequent traders: high returns, but low profits

Returns and yearly profits by frequency deciles





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#### Known predictors of percentage returns

- <u>Buy:</u> purchases more likely to be information-driven, sales may be motivated by diversification or liquidity (e.g. Jeng, Metrick, and Zeckhauser 2003)
- Opportunistic: trades deviating from routine trading patterns are more informative (Cohen, Malloy, and Pomorski 2012)
- <u>CFO</u>: CFO trades more informative (Wang, Shin, and Francis 2015)
- **Executive:** Trades by insiders closer to decision-making are more informative (e.g., Ravina and Sapienza 2010)

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## Returns vs. profits: informed trading proxies

	Abnormal	Trade	Trade	Abnormal	Yr abnormal 1	Yr abnormal
	return	frequency	value	profit	profit	round-trip profits
	(1)	(2)	(3)	(4)	(5)	(6)
Buy only (d)	0.920***	-0.479***	-83.712***	0.405	-3.402***	68.041
	(0.10)	(0.05)	(15.05)	(0.58)	(1.01)	(44.59)
Opportunistic (d)	0.507***	-5.657***	26.220	2.083***	-2.396	-74.232
	(0.10)	(0.32)	(47.37)	(0.74)	(3.80)	(86.68)
Infrequent (d)	0.333***	-4.322***	-286.435***	-1.135**	-21.834***	-246.999***
	(0.06)	(0.10)	(23.73)	(0.48)	(1.88)	(29.69)
CFO (d)	0.353***	-0.698***	-163.598***	0.220	-3.813***	-73.347***
	(0.08)	(0.04)	(18.75)	(0.49)	(1.15)	(19.19)
Executive (d)	0.197***	-0.770***	-54.086***	0.644	-2.588***	22.985
	(0.05)	(0.04)	(18.79)	(0.41)	(0.99)	(16.88)

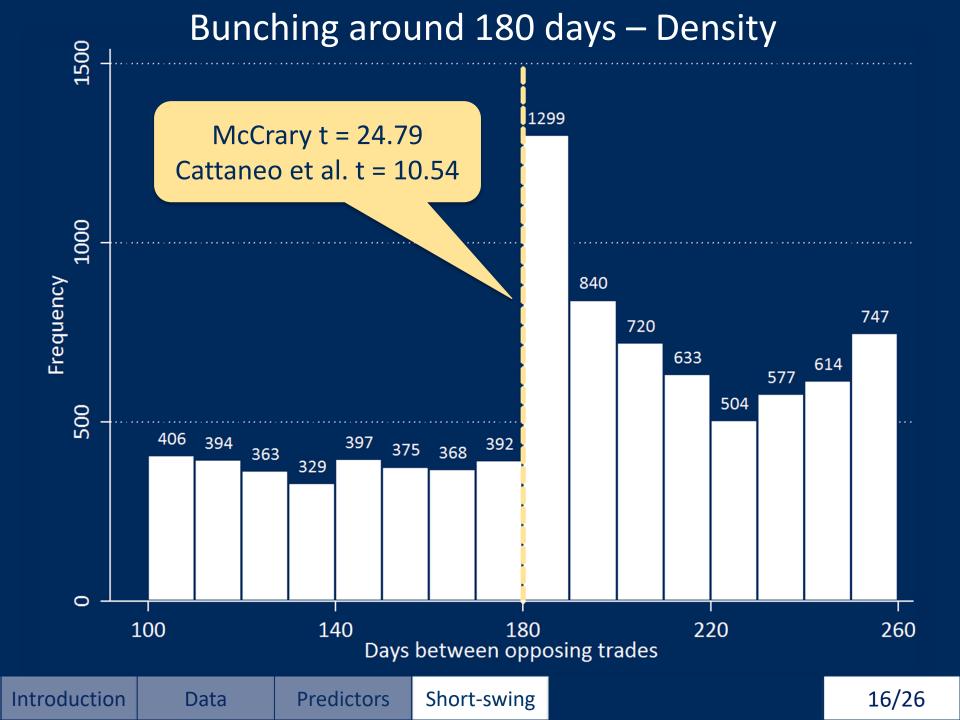
Each cell contains results from a separate regression, with controls and FE Returns of infrequent traders 37% higher, but profits more than 100% lower.

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#### Revealed preferences using the short-swing rule

- Short-swing rule: Round-trip profits within less than 6 months have to be returned to issuer
  - Section 16(b) of the Securities Exchange Act of 1934
- Exploit potential discontinuity to infer trading motives
  - Bunching around points that feature discontinuities to elicit behavioral responses (Bach and Metzger 2018, Goncharov, Ioannidou and Schmalz 2018)
  - Manipulability: assignment variable is discrete choice opposite of RD
- Study insider behavior around the 6-month threshold
  - Null: If insiders did not care about keeping profits, distribution around 180 days between opposite trades should be continuous
  - Close trade just after expiration → likely driven by profit-seeking

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## Bunching around 180 days – Tests

	McC	Crary (2008	Cattaneo et al. (2017)					
Cutoff (days)	Log density	s.e.	t	t				
	Short-swing rule							
180	1.042	0.042	24.789	10.537				
181	1.029	0.042	24.470	12.139				
182	1.011	0.042	24.179	12.479				
		Placebo ci	utoffs					
30	-0.650	0.033	-19.442	-2.999				
60	-0.337	0.039	-8.638	-1.920				
90	0.106	0.050	2.136	1.678				
100	-0.120	0.053	-2.238	-0.641				
365	-0.004	0.042	-0.097	-0.411				
730	0.070	0.061	1.147	0.437				

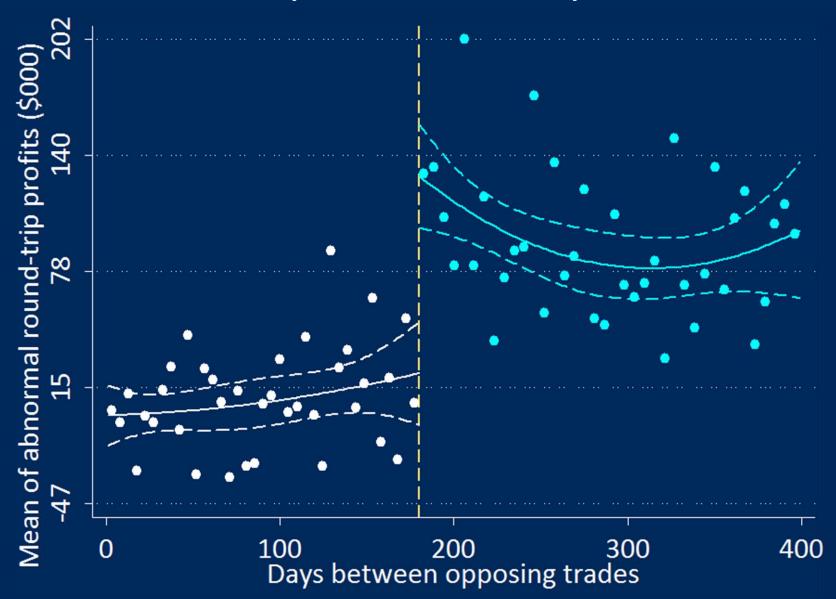
Data

## Bunching around 180 days – Subsamples

Subsample	Before	After	Factor	χ2	p-value
Firm size					
Small firms	84	186	2.2	6.33	0.01
Large firms	72	255	3.5		
Market-to-book value					
Low M/B	74	265	3.6	3.69	0.03
High M/B	93	236	2.5		
Insider wealth					
Low wealth	13	31	2.4	3.10	0.05
High wealth	4	28	7.0		
Insider type					
Other insiders	181	451	2.5	8.89	0.00
Executives	219	772	3.5		
Analyst forecast error					
Low error	9	15	1.7	3.50	0.04
High error	5	27	5.4		
Analyst forecast dispersion					
Low dispersion	8	7	0.9	7.97	0.00
High dispersion	5	29	5.8		

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#### Discontinuity around 180 days - Profits



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#### Mechanical relation?

- Stark differences between trades closed 181-200 days and other trades
- Is this all mechanical?
  - Suppose insider happens to get lucky
  - This is why she closes the trade just after expiration
- Disentangling intentions from a mechanical relation:
  - Oboos short-swing trading predict future behavior?
  - Inconsistent with mechanical interpretation

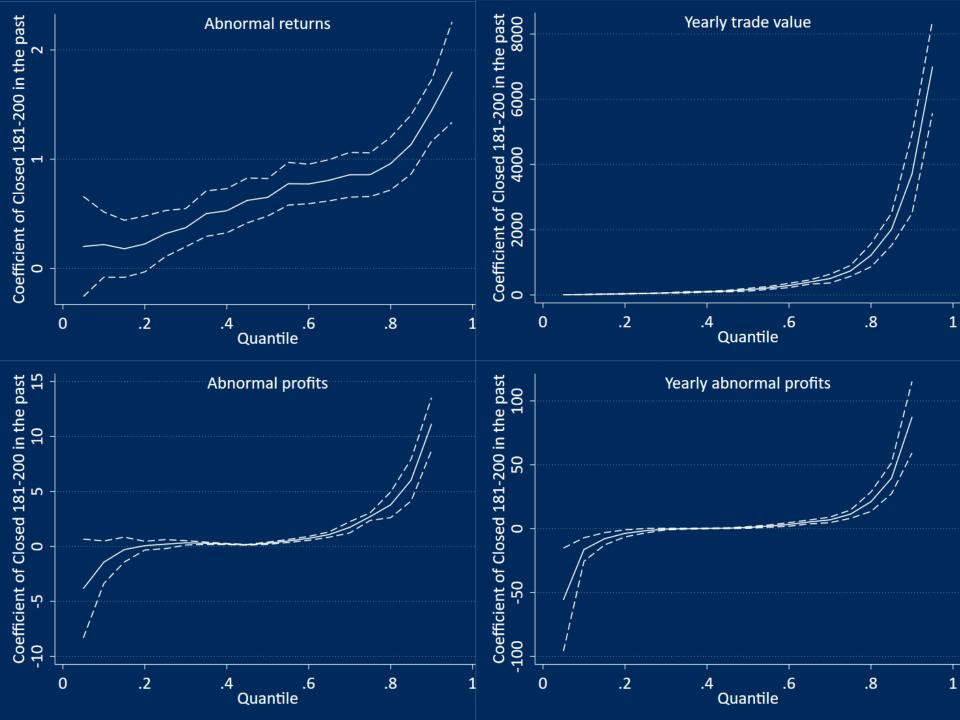
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## "Short-swing closers" make higher future profits

	Abnormal	Trade	Trade	Abnormal	Yr abnormal	Yr abnormal
T6A: All observations	return	frequency	value	profit	profit	round-trip profit
	(1)	<u>(2)</u>	<u>(3)</u>	(4)	(5)	(6)
Closed 181-200 in the past	0.372*	1.966***	130.477***	3.932**	22.442***	120.908**
	(0.20)	(0.24)	(43.75)	(1.66)	(5.59)	(48.41)
Controls	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE, Year FE	Yes	Yes	Yes	Yes	Yes	Yes
N	643,558	262,218	643,558	643,558	262,218	30,002
	Abnormal	Trade	Trade	Abnormal	Yr abnormal	Yr abnormal
T6C: Only round-trips	return	frequency	value	profit	profit	round-trip profit
	(1)	(2)	(3)	(4)	(5)	(6)
Closed 181-200 in the past	0.297	3.039***	164.634***	2.443	33.035*	305.330**
	(0.41)	(0.65)	(55.65)	(2.74)	(17.06)	(136.91)
Controls	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE, Year FE	Yes	Yes	Yes	Yes	Yes	Yes
N	58,174	23,526	58,174	58,174	23,526	7,781

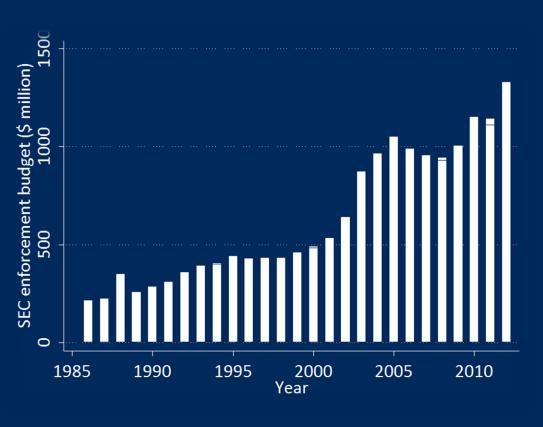
Short-swing trading predicts higher <u>future</u> returns and profits

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#### SEC enforcement budget as monitoring proxy

How do abnormal returns and profits respond to variation in monitoring intensity?



- Resource-based measure of enforcement (Del Guercio, Odders-White, and Ready 2017)
- Determined through political process, not by amount of insider trading
- Produces variation in attention by regulator/monitoring

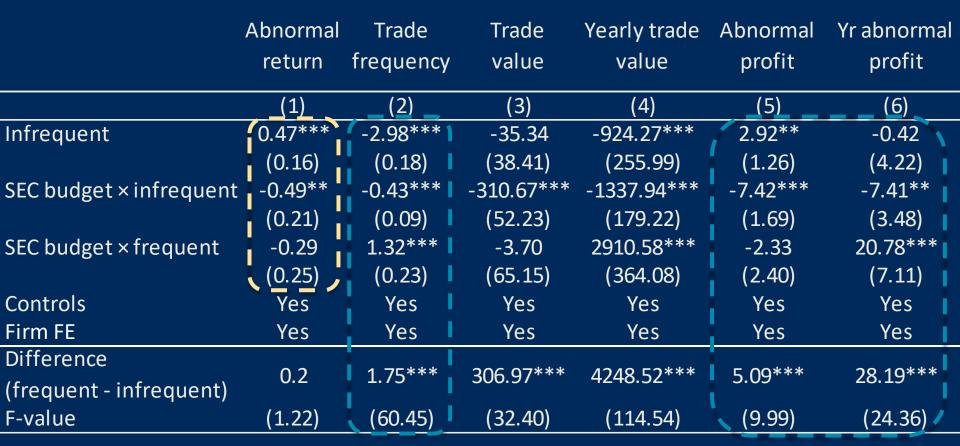
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## SEC enforcement budget

	Abnormal return	Trade frequency	Trade value	Abnormal profit	Yr abnormal profit	Yr abnormal round-trip profit
	(1)	(2)	(3)	(4)	(5)	(6)
SEC budget	-0.430**	-0.180*	-191.795***	-5.599***	-2.339	25.120
	(0.21)	(0.10)	(50.64)	(1.83)	(3.80)	(47.71)
Controls	Yes	Yes	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes

Returns, volume, and per-trade profits decrease with monitoring, but yearly profits do not

#### SEC budget: frequent vs. infrequent traders



#### When SEC enforcement intensity is high:

- Infrequent traders trade less, their trades are less profitable
- Frequent traders trade (even) more, realize higher profits

#### Summary

\$ profits and % returns offer contrasting evidence on insider trading!

- Typical insider trading profits are small
- Informed trading proxies predict returns, but not profits
  - Proxies are negatively correlated with quantities
  - Frequency is first-order determinant of profits
- Our novel proxy of trading intentions predicts profits
  - Even insiders identified by this measure make modest profits
- Different insiders respond differently to monitoring

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