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### Innovation and Informed Trading: Evidence from Industry ETFs

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## The rise of ETFs...

- Few recent financial innovations have had the impact of ETFs
  - More than 5,000 exchange-traded products, assets now exceeding those of hedge funds, etc.
- This "disruptive innovation" has had far reaching effects
  - A more liquid, lower-cost alternative to mutual funds
  - A vehicle to access unavailable asset classes
- In this paper we investigate another role the expanded ability to hedge



# Industry ETFs

- We focus on the role of industry ETFs and how this innovation affects informed trading and the efficiency of the market.
  - Innovations that facilitate risk-sharing can facilitate informed trading (Dow, 1998; Goldstein, Li, and Yang, 2014; et al)
  - While futures or index ETFs can be used to hedge market risk, industry ETFs can better hedge industry risk
    - Investors bet on firm-specific information: long stocks/short ETFs
  - We can use the time-series variation in inception dates to investigate their economic impact.

# Short interest on industry ETFs

15 Most Shorted ETFs (excluding inverse/leveraged)

Ticker	Fund	Short Interest %
XRT	SPDR S&P Retail ETF	507.99
SMH	VanEck Vectors Semiconductor ETF	176.58
UNG	United States Natural Gas Fund LP	106.57
ХОР	SPDR S&P Oil & Gas Exploration & Production ETF	89.41
VXX	iPath S&P 500 VIX Short-Term Futures ETN	80.95
FXE	CurrencyShares Euro Trust	78.56
XBI	SPDR S&P BIOTECH ETF	59.43
IMED	PureFunds ETFx HealthTech ETF	58.63
DWLV	PowerShares DWA Momentum & Low Volatility Rotation Portfolio	57.34
FXY	CurrencyShares Japanese Yen Trust	55.53
CTNN	iPath Pure Beta Cotton ETN	50.00
EWW	iShares MSCI Mexico Capped ETF	48.07
IYR	iShares U.S. Real Estate ETF	47.92
ERY	Direxion Daily Energy Bear 3X Shares	46.39
OIH	VanEck Vectors Oil Services ETF	42.75

Source: Bloomberg; data as of July 17, 2017

- Industry ETFs have large short interest
- Short interest does not have to imply a bearish outlook
  - Bloomberg recently argued "...The funds' shorts don't necessarily indicate bearish sentiment, but rather are used to hedge out part of the market in order to isolate a long position..."



## "long the stock/short the ETF"

- We hypothesize that informed traders use industry ETFs to hedge their long positions on firms with positive firm-specific information
- Two implications
  - Short interest on the industry ETF should generate a temporary price impact leading to a price reversal on ETFs
    - Short interest could positively predict the return of the industry ETF
  - This hedging let informed traders trade more aggressively, which could make the market for underlying stocks more efficient

# Our empirical work

- We test for return predictability using a Fama-MacBeth approach.
  - We know positive short interest predicts negative performance at the stock level is this also true for the industry ETF?
- We test for the informed trading effects using earnings announcements.
  - If more aggressive informed trading, then the market should react less to positive earnings surprises.

# A vast literature(s)

- Innovations facilitate risk-sharing or facilitate speculative trading — Allen and Gale(1994); Dow(1998); Simsek (2013)
- EFTs as financial innovations Chen (1995); Cong and Xu (2016); Madhavan (2016); Bhattacharya and O'Hara (2017)
- Empirical ETFs Ben-David et al (2014); Wermers and Xue (2015); Glosten et al (2017); Israel et al (2017)
- Short-selling Diether et al (2009); Battalio and Shultz (2011); Grundy et al (2012); Comerton-Forde, et al (2016)

# Data and Sample

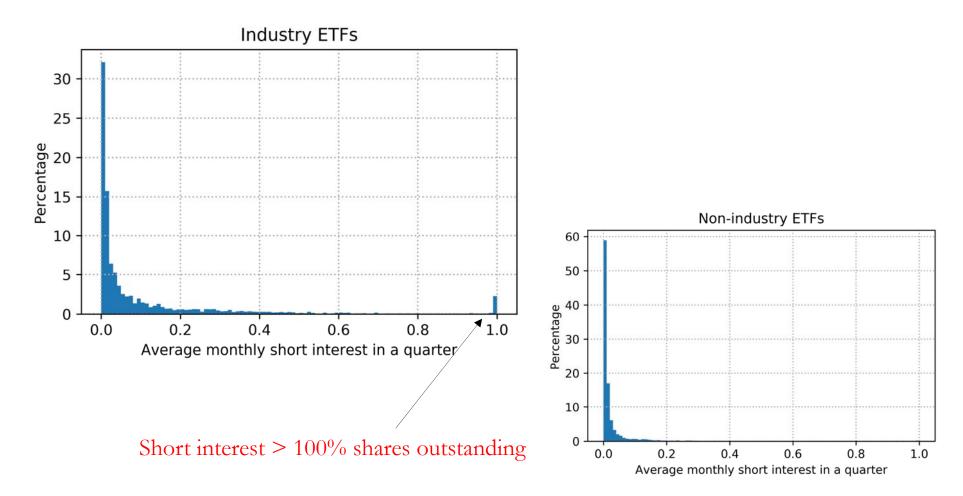
#### • The ETF list

- CRSP Mutual Fund Database + ETFDB
- Equity only, US only 449 ETFs

#### • Extract industry ETFs based on holdings

- Thompson Reuters 13F data + CRSP for holdings
- We require the dominant industry holdings exceed one-third of ETFs holdings and the ETF must have at least 30 stocks
- Data on price and volume are from CRSP, short interest is from COMPUSTAT
- Final sample 116 industry ETFs (based on FF12 industry classification)
- **Earnings data** is constructed following Livnat and Mendenhall (2006), and  $SUE_{i,t} = \frac{EPS_{i,t} EPS_{i,t-4}}{\sigma_{i,t}}$
- Funds data
  - Hedge funds list is constructed following Griffen and Xu (2009)
  - Intuitional holdings based on Thompson Reuters 13F

### Industry vs. Non-industry ETFs



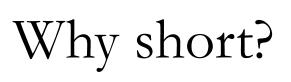
## Industry vs. Non-industry ETFs

	Mean	Std.	5%	25%	50%	75%	95%
SIR	0.118	0.211	0.001	0.007	0.026	0.117	0.607
Price	53.842	32.056	18.104	30.167	48.465	68.034	110.976
Volume (in shares)	422847.684	2111222.625	764.333	5382.917	19368.167	75731.833	1963009.083
TNA (in \$ millions)	1051.615	2156.886	11.417	91.817	331.200	918.467	5295.167

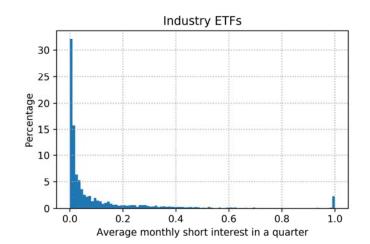
#### Panel A: Industry ETFs

#### Panel B: Non-industry ETFs

	Mean	Std.	5%	25%	50%	75%	95%
SIR	0.041	0.118	0.000	0.003	0.008	0.021	0.191
Price	56.290	35.659	15.884	28.185	48.853	75.056	123.672
Volume (in shares)	313927.645	2855881.357	538.017	2304.917	7388.500	32629.167	380683.450
TNA (in \$ millions)	2309.432	9704.713	9.467	50.733	179.333	923.600	10578.307



• Speculation hypothesis



- Investors are betting against a particular industry
  - Suggests bad news on the member stocks
- Hedging hypothesis
  - Informed investors short an industry to hedge their long position on a particular underlying stock
    - Suggests good news on the member stocks

### Positive earnings ratio

- We construct a measure that captures the earnings performance of the underlying firms in each ETF.
- Step 1: we define positive earnings if a firm's SUE is in the top 25% of the sample.
- Step 2: We compute the ratio of underlying firms in the ETF with positive earnings.

# Hedging or Speculating?

• We ran the following regression:

 $PosSUE_{i,t} = \alpha_i + \alpha_t + \beta_1 SIR_{i,t-1} + controls + \epsilon_{i,t},$ 

 $\begin{aligned} &PosSUE_{i,t} \\ &= \alpha_i + \alpha_t + \beta_1 SIR_{i,t-1} + \beta_2 DummyIndet f_i \times SIR_{i,t-1} + controls + \epsilon_{i,t}. \end{aligned}$ 

- $DummyIndetf_i$  is 1 if the ETF is an industry ETF.
- We control for year, quarter, and ETF fixed effects.

## Industry ETFs are different

Dependent variable: PosSUE_(i,t)	Ind. ETFs	Non-ind. ETFs	All
$\frac{10550E_{(i,t-1)}}{SIR (i,t-1)}$	0.0324*	-0.0235*	-0.0253*
	(1.6687)	(-1.6823)	(-1.7972)
DummyIndetf x	-	-	0.0581**
SIR (i,t-1)	-	-	(2.481)
log(TNA)	0.0067**	0.0024**	0.0037***
	(2.416)	(2.2304)	(3.0643)
Year F.E.	Y	Y	Y
Quarter F.E.	Y	Y	Y
ETF F.E.	Y	Y	Y
Num.Obs.	4079	9413	13492
R-squared	0.4868	0.6592	0.5855

# Sample excluding 2007-2008

Dependent variable: PosSUE_(i,t)	Ind. ETFs	Non-ind. ETFs	All
$\frac{1055 \text{ CH}_{(i,t)}}{SIR (i,t-1)}$	0.0436*	-0.0124	-0.0174
	(1.8814)	(-1.0757)	(-1.4781)
DummyIndetf x	-	-	0.0625**
SIR(i,t-1)	-	-	(2.3364)
log(TNA)	0.0099***	0.0043***	0.0058***
	(2.7909)	(3.2452)	(3.8403)
Year F.E.	Y	Y	Y
Quarter F.E.	Y	Y	Y
ETF F.E.	Y	Y	Y
Num.Obs.	3580	8244	11824
R-squared	0.5003	0.6817	0.6054

# Predicable returns and IETF short interest

- We have shown that:
  - Industry ETFs have greater extreme short interest than other ETFs
    - "long the stock/short the ETF" strategy
- One asset pricing implication:

Extreme short interest should create a temporary price impact in the IETF — leading to predictable IETF returns

# Fama-MacBeth regression of returns on short interest ratios

<b>Dependent variable:</b> <b>Ret_(t+1)</b>	Industry ETFs		Member firms (of IETFs)		
$\Delta SIR_t$	0.019**	0.024***	-0.064**	-0.080***	
	(2.27)	(2.65)	(-2.32)	(-2.83)	
Intercept	0.013***	0.010***	0.009**	0.007*	
	(3.55)	(3.25)	(2.46)	(1.71)	
Controls	Ν	Y	Ν	Y	
Sample period		- 2006.12, - 2016.11	1999.03	- 2016.11	

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# Can you profit from this?

- We construct a long-short portfolio based on IETF monthly  $\Delta$ SIR deciles
  - We long the ETF in highest decile and short the ETF in the lowest decile.
- Result?
  - It generates a statistically significant monthly alpha of approx. 30 basis points
  - Applying a similar approach to stocks yields a **negative 33 basis points**.

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# Industry ETFs and market efficiency

- If informed traders use industry ETFs to hedge, then this may allow more aggressive trading
  - Suggests that information is impounded faster so this should reduce the market reaction to positive earnings surprises
- We calculated the cumulative abnormal daily return from days -1 to1 around the earning announcement based on the Fama-French 3 factor model.
- We ran

 $\begin{aligned} & CAR_{i,t} \\ &= \alpha_i + \alpha_t + \theta_1 SUE_{i,t} + \theta_2 DummyIndetfown + \theta_3 DummyIndetfown \\ &\times SUE_{i,t} + controls + \epsilon_{i,t}. \end{aligned}$ 

#### Market reaction to positive SUE is reduced

Dependent variable:	All	Negative SUE	Positive SUE
$CAR_{(i,t)}$		C	
SUE (i,t)	0.2002***	0.0944***	0.0043
	(30.8998)	(9.6612)	(0.3632)
DummyIndetfown	0.007***	0.0045***	0.0153***
	(9.6669)	(3.0769)	(9.765)
<i>DummyIndetfown x SUE_(i,t)</i>	-0.0551***	-0.014	-0.0464**
	(-5.0421)	(-0.8049)	(-2.4792)
log(MktCap)	-0.0014***	0.0011***	-0.0072***
	(-7.3373)	(2.9473)	(-16.2583)
BM	0.0032***	0.0036***	0.0039***
	(7.3859)	(4.8421)	(4.9137)
TR	-0.566***	-0.5242***	-0.8457***
	(-10.0016)	(-5.4694)	(-8.2553)
MOM	-0.0012	-0.0071***	-0.0021
	(-1.3935)	(-4.9633)	(-1.5745)
EarnPerst	0.0	-0.0016	0.0018
	(0.0429)	(-1.3259)	(1.3534)
NumEst	0.0004***	0.0007***	0.0007***
	(7.5835)	(5.7893)	(6.0641)
Industry F.E.	Y	Y	Y
Month F.E.	Y	Y	Y
Year F.E.	Y	Y	Y
Num.Obs.	291599	72715	72922
R-squared	0.0163	0.0116	0.0155

# The channel ...

- Hedge funds could be a channel for this effect if they are using IETFs to hedge.
- To investigate this, we calculated aggregate abnormal hedge fund holdings on firm *i* in the quarter before the earnings announcement.
- We ran:

$$\begin{split} & HfAhdngRatio_{i,t} \\ &= \alpha_i + \alpha_t + \theta_1 SUE_{i,t} + \theta_2 DummyIndetfown + \theta_3 DummyIndetfown \\ &\times SUE_{i,t} + controls + \epsilon_{i,t}. \end{split}$$



Dependent variable:	All	Negative SUE	Positive SUE
HfAhdngRatio_(i,t)		C	
SUE (i,t)	0.1345	0.0428	0.2352
,	(1.5702)	(0.3039)	(1.3629)
DummyIndetfown	-0.0363*	0.0164	-0.1041***
	(-1.9484)	(0.4293)	(-2.8134)
DummyIndetfown x SUE_(i,t)	0.0735	-0.0181	0.931**
	(0.3437)	(-0.0554)	(2.1608)
log(MktCap)	-0.0122***	0.0108	0.0226***
	(-3.1918)	(1.3304)	(2.9731)
BM	0.0197*	-0.0169	-0.007
	(1.8528)	(-0.9672)	(-0.4304)
TR	32.3415***	23.336***	31.3782***
	(23.6757)	(9.7707)	(13.2446)
MOM	0.034*	0.1052***	0.0972***
	(1.9494)	(3.5782)	(4.0837)
EarnPerst	-0.0672***	0.0074	-0.0444
	(-4.7204)	(0.2544)	(-1.5082)
NumEst	-0.007***	-0.0035	-0.0124***
	(-5.2105)	(-1.1806)	(-4.0509)
Industry F.E.	Y	Y	Y
Month F.E.	Y	Y	Y
Year F.E.	Y	Y	Y
Num.Obs.	291620	72722	72932
R-squared	0.035	0.0397	0.0342

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#### A placebo test — mutual fund holdings

Dependent variable:	All	Negative SUE	Positive SUE
MfAhdngRatio_(i,t)		C	
$SUE_{(i,t)}$	2.5139***	4.0446***	-4.2241***
	(10.6551)	(10.0637)	(-10.0624)
DummyIndetfown	-0.9834***	-0.5686***	-1.1736***
	(-14.9465)	(-5.2105)	(-11.6335)
<i>DummyIndetfown x SUE_(i,t)</i>	-0.6845	0.6863	0.8297
	(-1.1713)	(0.7723)	(0.815)
log(MktCap)	0.3333***	0.3589***	0.4869***
	(19.7558)	(13.521)	(19.4775)
BM	-0.5808***	-0.3701***	-0.381***
	(-16.2149)	(-7.523)	(-8.6381)
TR	71.5907***	-15.8043**	82.3471***
	(16.9328)	(-2.442)	(12.9173)
МОМ	2.4397***	2.38***	1.5836***
	(34.256)	(25.9914)	(20.9999)
EarnPerst	0.1002**	-0.0479	0.1877**
	(2.1033)	(-0.5969)	(2.2023)
NumEst	-0.0776***	-0.0396***	-0.0844***
	(-16.9889)	(-4.6167)	(-9.7255)
Industry F.E.	Y	Y	Y
Month F.E.	Y	Y	Y
Year F.E.	Y	Y	Y
Num.Obs.	291620	72722	72932
R-squared	0.0896	0.0694	0.0898



#### Extensions

- Are hedge funds really hedging?
  - Hedge funds often act as arbitrageurs in the ETF creation/redemption process. Could this be the reason we find a correlation between SIR and hedge fund holdings
    - One implication of the ETF arbitrage is that this would suggest a hedge fund increases holdings on all constituents of the ETF.
    - The correlation should be universal among all constituents:  $\beta_1 > 0$ .
- We ran:

$$\begin{split} HfAhdngRatio_{i,j,t} &= \alpha_i + \alpha_t + \beta_1 SIR_{j,t-1} + \beta_2 DummyPosSUE_{i,t} \times SIR_{j,t-1} + controls \\ &+ \epsilon_{i,j,t} \end{split}$$

# Are hedge funds hedging?

		Not
Dependent variable: HfAhdngRatio_(i,j,t)	Coef.	
<i>SIR_(j,t-1)</i>	0.1048	significant
	(0.9544)	
DummyPosSUE_(i,t) x SIR_(j,t-1)	0.282**	
	(2.0229)	
log(MktCap)	-0.0548***	
	(-8.6449)	
Industry F.E.	Y	Positive correlation
Year F.E.	Y	between the SIR
Quarter F.E.	Y	
ETF F.E.	Y	and hedge fund
Num.Obs.	343613	holdings on firms
R-squared	0.0167	with positive
		earnings surprises

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### Conclusion

- Industry ETFs appear to be a valuable innovation by facilitating the hedging of industry risk
  - Our paper provides strong evidence of industry ETFs enabling the "long the stock/short the industry ETF" strategy
    - Short interest in industry ETFs is a different animal than short interest in stocks and even other ETFs
    - It makes the market more informationally efficient
    - It appears to be implemented by hedge funds
    - It results in return predictability
    - It is asymmetric reflecting that while shorting costs of ETFS are low, shorting costs of stocks are not.



#### Thank you!