

# Do Hedge Funds Hedge? New Evidence from Volatility Risk Premia Embedded in VIX Options

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## 1 Motivation

- Background
- Our Paper
- Main Findings
- Main Contributions

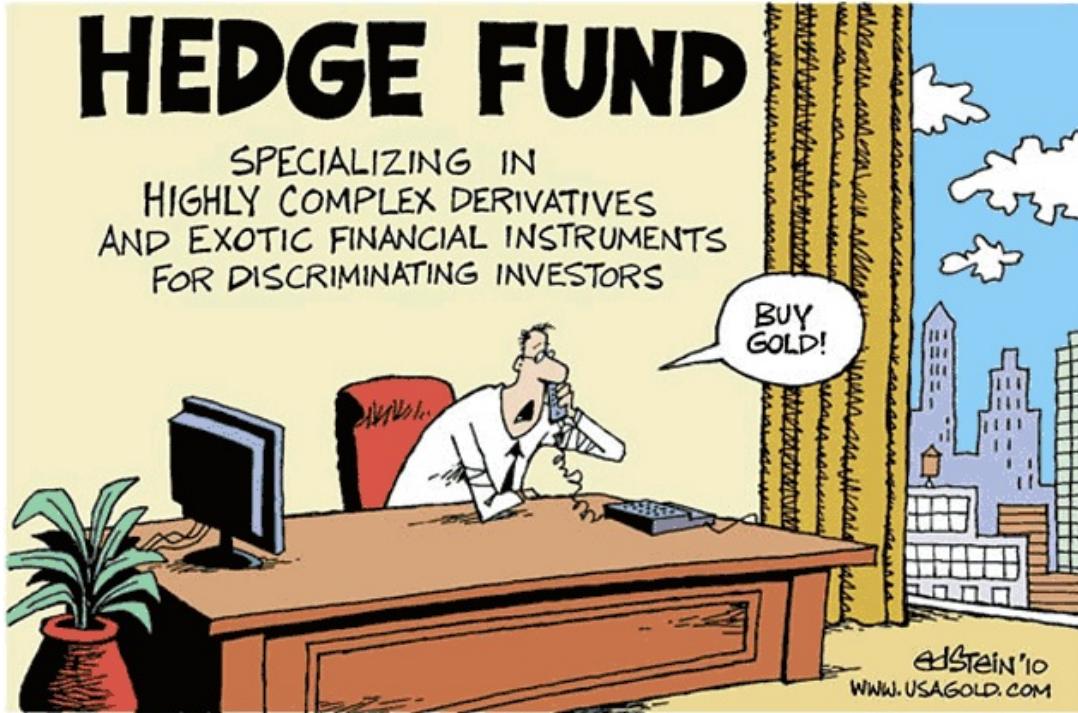
## 2 Methodology

## 3 Implementation

## 4 Results

## 5 Conclusion

## Background



**Figure 1: A culture of secrecy.** Understanding complex hedge funds strategies has become crucial to explain hedge fund performance, managers' skills, and compensation scheme

# Background

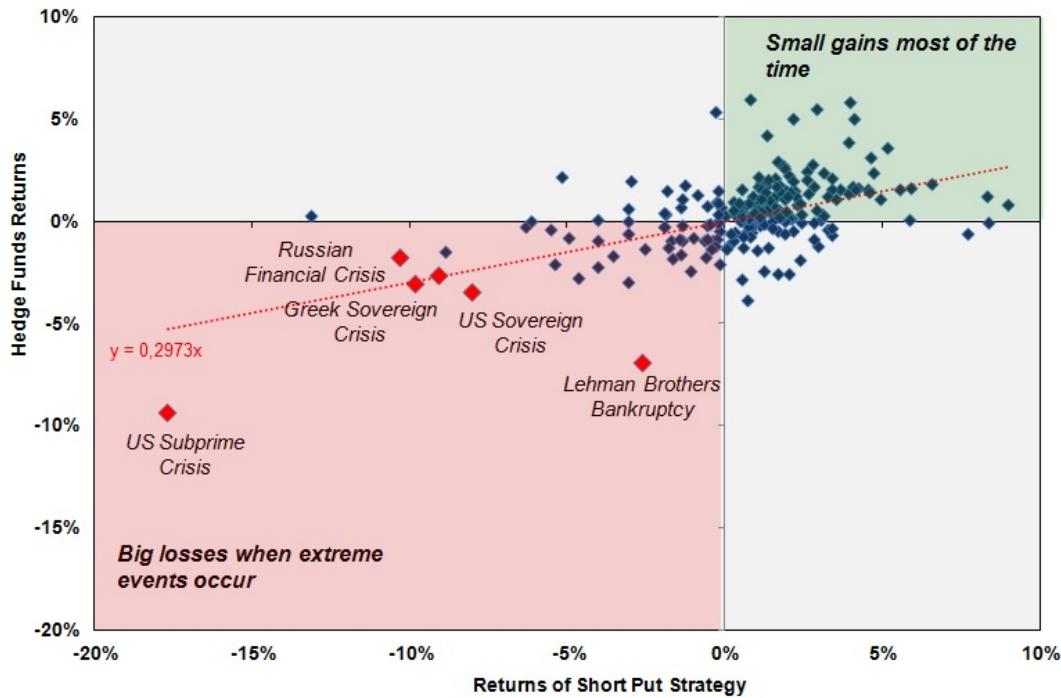


Figure 2: **Penny-by-penny gains before big losses.** This scatterplot exhibits monthly returns between Hedge Fund Global Index and **S&P 500 Put Write Index** over the period from January 1998 to June 2017

# Background

Alpha or Beta?

*“Hedge funds may have strategies that yield payoffs similar to those of a company selling earthquake insurance; that is, most of the time the insurance company makes no payouts and has a nice profit, but from time to time disaster strikes and the insurance company makes large losses that may exceed its cumulative profits from good times. (...) Others have described hedge fund strategies as picking up pennies in front of a steamroller.”*

*“A hedge fund that implements a strategy akin to selling earthquake insurance and whose risk is not captured well by commonly used risk factors will have a significant positive alpha—until the quake hits.”*

R. Stulz, *Journal of Economic Perspectives*, 2007

# Background

## Alpha or Beta?

” Exotic risks are also referred to as “advanced”, “**alternative**”, or “**smart beta** in the literature (e.g., Carhart et al. 2014). In our taxonomy, we separate premium-bearing risks into those that are generally available through liquid, low-cost, and transparent investment vehicles such as index mutual funds or ETFs (**traditional beta**) from those that can typically only be obtained through **hedge funds** (**exotic beta**). ”

V. Agarwal, C. Green, and H. Ren, *Journal of Financial Economics*, 2017

# Our Paper

- HF are not just selling put options but this example is interesting
- The PNL of the strategy is (at least) impacted by two factors
  - ▶ When PRICE drops, Put Premium increases, PNL is NEGATIVE
  - ▶ When VOL spikes, Put Premium increases, PNL is NEGATIVE
- Under the standard Black-Scholes framework, the two factors are not treated in the same manner
  - ▶ (log)-PRICE Variations are stochastic (return **distribution**)
  - ▶ (log)-VOL is a **parameter** (potentially time-varying)
- How to take into account the stochastic nature of the VOL?

# Our Paper

- Generally, HF use derivatives to "hedge" some risks, and then are exposed to VOL (**distribution**)
- What is the impact of the stochastic nature of VOL (**distribution**) on:
  - ▶ Portfolio/Investment decisions ?
  - ▶ In turn, the cross-section of HF returns ?
- Already been investigated in Agarwal, Arisoy, Naik (JFE, 2017) paper  
**BUT**
  - ▶ They implicitly assume that the (log)-VOL is **normally distributed**
  - ▶ They build an investment strategy to get an exposure to the **VOL** of the (log)-VOL distribution
  - ▶ They use this factor in an asset pricing exercise
- Can we extend the study to higher-order moments (**SKEW, KURT**) of the (log)-VOL distribution ?

# Our Paper

We solve several technical issues:

## ① The data issue:

- ▶ How to filter from VIX option market data some information about the **implied distribution** of VOL ?
- ▶ How to filter from VIX spot returns a proxy for the **historical distribution** of VOL ?

## ② The investable factor issue:

- ▶ How to build an investable strategy exposed to **VOL** of VOL ?
- ▶ How to build an investable strategy exposed to **SKEW** of VOL ?
- ▶ How to build an investable strategy exposed to **KURT** of VOL ?

# Main Findings

## 1 - HF are particularly VOL sensitive

HF are exposed to VOL risk, and the **3 VOL risk premia** are instrumental determinants of HF performance

## 2 - VOL sensitivity substantially arises from VOL of VOL

HF are particularly negatively exposed to **VOL of VOL**. This is especially true for **Relative Value** and **Equity Hedge** during crises when uncertainty is high

## 3 - HF are also exposed to higher-order moments of VOL

**Relative Value** and **Directional** are positively exposed to **SKEW of VOL**. **Global Macro** are positively exposed to **SKEW of VOL**. **Merger Arb** are positively exposed to **KURT of VOL** during crises

# First Contribution - Evidence from HF Performance

- Volatility is instrumental in both **time-series** and **cross-section**
- Does **HF alpha** arise from volatility strategies ?
- Use of **multiple option-based** dynamic trading strategies

## Existing literature

- Asness et al. (2001), Geman et al. (2003), Agarwal and Naik (2004), Stulz (2007), Patton (2009), Agarwal, Arisoy, and Naik (2015), Agarwal, Green, and Ren (2017)
- Extract VOL risks in HF from **nontradable** risk measures, or **standard** and **fragmentary** option-based strategies

## Our contribution

- Extract VOL risks from **multiple** tradable **VOL risk premia**
- As **decomposed** into **VOL, SKEW and KURT of VOL** strategies
- Shows they are instrumental **determinants** in HF performance

# Second Contribution - Evidence for Risk Premia Strats

- Most HF styles sell crash insurance, but VOL risk exposures across HF styles are **complex**
- Depend on **specific** trading strategies they use to "hedge" risk exposures

## Existing literature

- Ait-Sahalia et al. (2001), Alireza (2005), Chang et al. (2013), Bondarenko (2014), Schneider et al. (2015), Al Wakil (2016)
- Evidences on **profitable** divergence trading strategies to monetize compensation for higher-order risks

## Our contribution

- Evidences that **divergence swaps** are widely traded by HF
- Shows **VOL, SKEW, KURT of VOL** risk premia are mimicking portfolios for VOL risk premia harvested by HF

## 1 Motivation

## 2 Methodology

- Multifactor Models
- Volatility Risk Premia
- Mimicking Portfolios

## 3 Implementation

## 4 Results

## 5 Conclusion

# Multifactor Models

- HF returns exhibit **option-like payoffs** replicable by option-based factors
- Examples of option-based strategies: short puts (Agarwal and Naik, 2004), VIX lookback straddles (Agarwal, Arisoy, and Naik, 2015), OTM puts (Agarwal, Green, and Ren, 2017)
- We use the same approach with new factors related to the volatility distribution:

$$R_{it} = \underbrace{\alpha}_{\text{Hedge fund alpha}} + \underbrace{\sum_{j=1}^m \beta_{jt} F_{jt}}_{\text{Control variables: Fung-Hsieh factors (2004)}} + \underbrace{\sum_{k=1}^n \beta_{kt} F_{kt}}_{\text{VOL risk premia: tradable, option-implied}}$$

where  $R_{it}$  is returns of  $i$ -th HF strategy

# Volatility Risk Premia

- Existing literature usually captures VOL risk from a **unique** exposure: doesn't reflect the full extent of volatility strategies [see e.g. Bondarenko (2004), Carr and Wu (2009), Bollerslev et al. (2009)]

## VOL (of VOL) risk premium

- Widely traded to exploit pricing discrepancies between risk-neutral and realized volatilities (of volatility):

$$VoV_{t,t+\tau,T} \equiv E_t^P [\sigma_{t,t-\tau,T}] - E_t^Q [\sigma_{t,t+\tau}]$$

where **VoV** is the difference between ex-post realized volatility and ex-ante risk-neutral expectation of future volatility

- Correspond to the payoff of a **volswap** on volatility

# Volatility Risk Premia

Let  $IV_{t,T}$  volatility smirk at time  $t$  for maturity  $T$  and moneyness  $\xi$ . Assume three-dimensional representation of  $IV$ :

$$IV_{t,T}(\xi) = \begin{cases} \gamma_{0,t,T} & \text{Black - Scholes : flat smile} \\ \gamma_{0,t,T}[1 + \gamma_{1,t,T}\xi] & \text{Skewed IV smile} \\ \gamma_{0,t,T}[1 + \gamma_{1,t,T}\xi + \gamma_{2,t,T}\xi^2] & \text{Smirked IV smile} \end{cases}$$

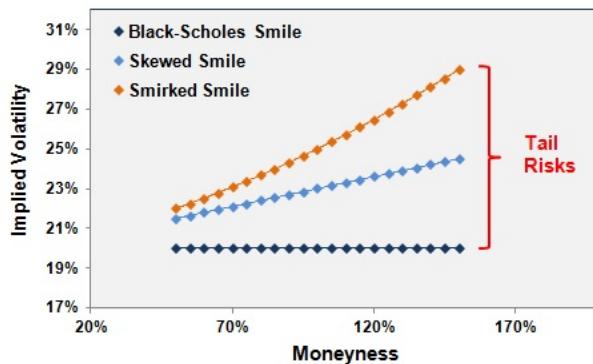


Figure 3: Smile dynamics contains market price of risk. VOL risks are embedded in level  $\gamma_{0,t,T}$ , slope  $\gamma_{1,t,T}$ , and curvature  $\gamma_{2,t,T}$

# Mimicking Portfolios

(Using Zhang-Xiang (2008) asymptotic approximations)

$$\begin{aligned}\gamma_{0,t,T} &\approx \left[ 1 - \frac{1}{24} (\textcolor{blue}{RNKurt}_{t,T} + 3) \right] \textcolor{blue}{RNVol}_{t,T}, \\ \gamma_{1,t,T} &\approx \frac{1}{6} \textcolor{blue}{RNSkew}_{t,T}, \\ \gamma_{2,t,T} &\approx \frac{1}{24} [\textcolor{blue}{RNKurt}_{t,T} + 3]\end{aligned}$$

Extending volswaps to **higher-order risks** gives:

$$\left\{ \begin{array}{ll} VoV_{t,T} = RDVol_t - RNVol_{t,T} & \text{Volatility Factor} \\ SoV_{t,T} = RDSkew_t - RNSkew_{t,T} & \text{Skewness Factor} \\ KoV_{t,T} = RDKurt_t - RNKurt_{t,T} & \text{Kurtosis Factor} \end{array} \right.$$

Volatility risk premia can be replicated by **distinct** delta-vega-neutral **portfolios** of European call and put options:

$$\begin{aligned}\Pi_{Kurt} &= C(S_t, K_2) + \frac{\nu_{C_2}}{\nu_{P_2}} P(S_t, K_2) - C(S_t, K_3) - \frac{\nu_{C_3}}{\nu_{P_1}} P(S_t, K_1) \\ &\quad - \left[ \Delta_{C_3} + \frac{\nu_{C_3}}{\nu_{P_1}} \Delta_{P_1} - \Delta_{C_2} - \frac{\nu_{C_2}}{\nu_{P_2}} \Delta_{P_2} \right] S_t\end{aligned}$$

1 Motivation

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# Implementation

## Risk-neutral estimators: $RNVol$ , $RNSkew$ , and $RNKurt$

- From market **VIX option prices**, provided by OptionMetrics
- **Model-free** approach of Bakshi, Kapadia, and Madan (2003)

## Realized estimators: $RDVol$ , $RDSkew$ , and $RDKurt$

- From **high-frequency VIX spot prices**, provided by Bloomberg
- **Two-Scales Realized** measure of Ait-Sahalia, Mykland and Zhang (2005)
- **Bias-corrected** measure using subsampling and averaging to control for market microstructure noise

## VOL risk premia: $VoV$ , $SoV$ , and $KoV$

- On a **daily** frequency, over **2008-2013**
- Risk premia are highly **ephemeral**: monthly calculations are inappropriate
- Constraint: length of **high-frequency** dataset

# Implementation

	30 Days Maturity			60 Days Maturity			90 Days Maturity		
	VoV	SoV	KoV	VoV	SoV	KoV	VoV	SoV	KoV
<b>Panel A: Levels of Option-Implied Risk Premia</b>									
Nb. Observations	300	300	300	267	267	267	237	237	237
<b>Mean</b>	<b>-0,193</b>	<b>-0,718</b>	<b>-6,027</b>	<b>-0,411</b>	<b>-0,747</b>	<b>-5,818</b>	<b>-0,487</b>	<b>-0,652</b>	<b>-5,707</b>
Median	-0,242	-0,849	-2,680	-0,447	-0,774	-2,264	-0,515	-0,766	-1,852
Max	1,361	3,031	4,855	1,033	2,645	5,358	0,971	2,507	4,723
Min	-1,280	-4,555	-35,560	-1,205	-5,522	-33,474	-1,258	-4,382	-33,029
Standard Dev.	0,397	1,437	8,073	0,354	1,502	8,056	0,388	1,450	8,420
Skewness	0,736	0,047	-1,437	1,011	-0,157	-1,392	0,662	-0,038	-1,467
Kurtosis	4,017	2,457	4,322	5,091	2,673	4,273	3,559	2,541	4,435
LBQ Test	79,14*	0,01	0,12	40,11*	0,05	0,57	37,21*	0	0
ADF Test	-8,57**	-14**	-12,07**	-6,33**	-13,16**	-10,98**	-5,52**	-12,87**	-11,01**
Student Test	-8,41**	-8,66**	-12,93**	-18,94**	-8,13**	-11,8**	-19,35**	-6,92**	-10,44**
<b>Panel B: First Differences of Option-Implied Risk Premia</b>									
Nb. Observations	299	299	299	266	266	266	236	236	236
Mean	0,000	-0,004	0,025	-0,216	-0,100	0,006	-0,312	-0,013	0,308
Median	-0,016	-0,032	0,055	-0,224	-0,103	0,359	-0,330	-0,223	0,091
Max	1,166	5,826	30,593	1,432	5,615	28,863	1,400	4,908	36,231
Min	-1,443	-5,247	-35,141	-2,003	-6,249	-33,166	-1,966	-5,965	-33,478
Standard Dev.	0,393	2,029	11,543	0,523	2,013	11,314	0,519	1,915	11,922
Skewness	0,051	-0,017	-0,059	-0,030	-0,126	-0,211	0,058	-0,030	-0,062
Kurtosis	3,687	2,687	3,528	3,784	3,179	3,638	3,593	2,930	3,716
LBQ Test	49,88*	61,45*	96,49*	52,91*	0,77	0,58	42,15*	0,89	0,9
ADF Test	-26,52**	-28,05**	-32,7**	-9,03**	-17,1**	-15,53**	-7,95**	-16,3**	-16,26**
Student Test	0	-0,03	0,04	-6,75**	-0,81	0,01	-9,24**	-0,1	0,4

Table 1: VOL risk premia embedded in VIX options

# Implementation

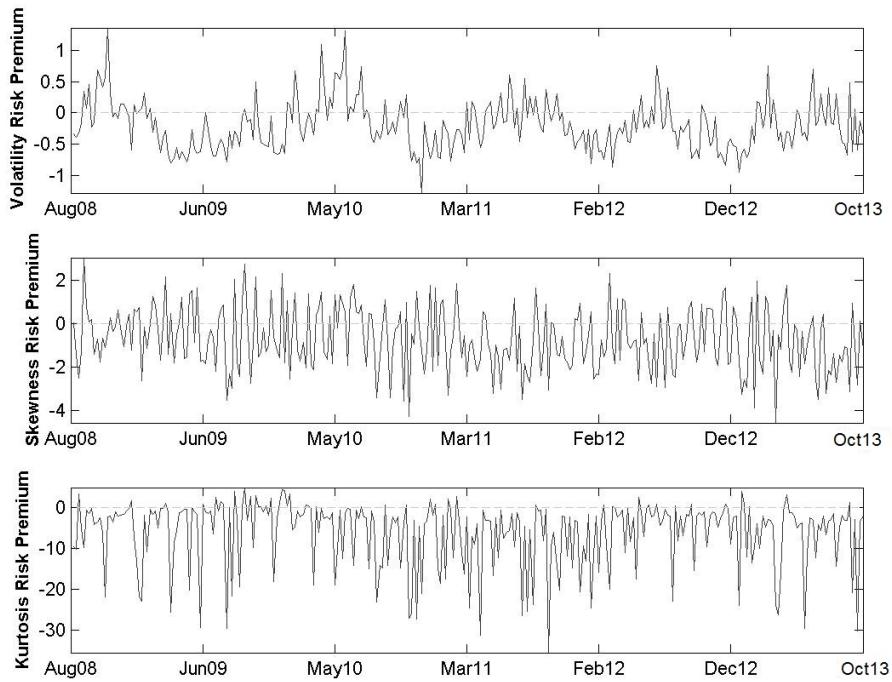


Figure 4: **VOL risk premia embedded in VIX options**

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  - Data
  - First Result - HF are VOL sensitive
  - Second Result - VOL sensitivity arises from VOL of VOL
  - Third Result - HF exposed to higher-order moments
- 5 Conclusion

# Data

## Multiple datasets

- **HF indices**, on a **daily** frequency, from HFR: Global, Directional, Equity Hedge, Macro, Merger Arbitrage, and Relative Value
- **Fung-Hsieh** seven factors on a **daily** frequency, i.e. MKT-RF, SMB, term spread  $\Delta$ , credit spread  $\Delta$ , and trend-following factors
- **VOL risk premia**, priced on a **daily** frequency, over 2008-2013

## Our approach

- Over 2008-2013, we regress 5 **daily HF indices** on Fung-Hsieh factors and the 3 volatility risk premia
- To understand HF performance on a **daily** frequency

# First Result - HF are VOL sensitive

	Factor Beta Quantiles	1 [Low]	2	3 [High]	High - Low
<b>Panel A: VOL of VOL Premia in the Cross-Section</b>					
Average VOL of VOL Beta	-0,216%	-0,026%	0,091%		0,307%
	[-18,35]	[+3,37]	[11,47]		[28,88]
Average Excess Return	0,0057	0,0051	-0,1113		-0,117***
	[0,16]	[0,15]	[+2,4]		[-2,38]
<b>Panel B: SKEW OF VOL Premia in the Cross-Section</b>					
Average SKEW OF VOL Beta	-0,019%	0,035%	0,059%		0,078%
	[-6,49]	[14,33]	[18,28]		[16,7]
Average Excess Return	0,001	-0,024	0,008		0,008
	[0,01]	[+0,6]	[0,18]		[0,15]
<b>Panel C: KURT OF VOL Premia in the Cross-Section</b>					
Average KURT OF VOL Beta	-0,003%	0,005%	0,009%		0,012%
	[-9,1]	[17,3]	[21,91]		[31,74]
Average Excess Return	0,0028	-0,0185	-0,0832		-0,086***
	[0,07]	[+0,52]	[+1,79]		[-1,86]

Table 2: Quantile portfolios sorted by VOL risk premia betas

## First Result - HF are VOL sensitive

After controlling for Fung-Hsieh factor loadings, we form **3 quantile portfolios** (high, medium, low beta) of cross-sectional HF index returns, sorted on loadings of each VOL premium:

- HF that significantly load on VOL OF VOL (KURT OF VOL) premia substantially **outperform** low-beta funds by nearly 11.7% (8.6%) per year
- This finding sheds light to what extent HF alpha arises actually from volatility strategies
- Nevertheless, HF that significantly bear SKEW OF VOL premia were not compensated by excess returns

# Second Result - VOL sensitivity arises from VOL of VOL

Investment Style	Nb. Obs.	Intercept	MKT-RF	SMB	dTERM	dCREDIT	PTFSBD	PTFSFX	PTFSCOM	VoV	SeV	KoV	R-Square / Adj. R-Square
Global	275	0.0000 [-0.19]	0.088*** [11.48]							-0.0012*** [-4.26]	0.0002** [2.09]	0.0000 [0.51]	0.41
		-0.0231 [-0.63]	0.0872*** [10.45]	0.0064 [0.37]	-0.0052 [-0.4]	-0.0114 [-0.63]	0.0024 [1.6]	4.3693 [0.63]	-0.0001** [-2]	<b>-0.0011*** [-3.4]</b>	0.0001* [1.8]	0.0000 [0.36]	0.41
Directionnal	275	0.0003 [1.22]	0.1979*** [15.11]							-0.0009* [-1.85]	0.0004*** [3.34]	0.0000 [1.5]	0.50
		-0.0407 [-0.66]	0.1908*** [13.52]	-0.0623** [-2.13]	-0.0027 [-0.12]	-0.0206 [-0.67]	0.0027 [1.06]	1.2206 [0.1]	-0.0002** [-2.24]	-0.0007 [-1.36]	0.0004*** [2.68]	0.0000 [1.48]	0.52
Equity Hedge	275	-0.0001 [-0.22]	0.1973*** [15.54]							-0.0013*** [-2.96]	0.0001 [1.06]	0.0000 [0.49]	0.54
		0.0096 [0.16]	0.1906*** [13.88]	0.0084 [0.29]	0.0153 [0.71]	0.0051 [0.17]	0.0044* [1.79]	16.4733 [1.43]	-0.0001* [-1.66]	<b>-0.0013*** [-2.63]</b>	0.0001 [0.52]	0.0000 [0.57]	0.53
Macro	275	-0.0009*** [-3.2]	-0.0321** [-2.23]							-0.0011** [-2.23]	-0.0004** [-2.5]	0.0000 [-0.01]	0.05
		-0.0991 [-1.54]	-0.0059 [-0.4]	0.0866*** [2.83]	-0.0691*** [-3]	-0.0495 [-1.54]	0.0018 [0.69]	-17.9721 [-1.46]	-0.0001 [-1.27]	-0.0005 [-0.95]	-0.0002 [-1.35]	0.0000 [-0.82]	0.18
Merger Arbitrage	275	0.0002 [1.21]	0.0979*** [11.3]							-0.0001 [-0.47]	0.0001 [0.62]	0.0000 [0.88]	0.36
		0.1272*** [3.12]	0.0901*** [9.68]	-0.0272 [-1.41]	0.0497*** [3.42]	0.0633*** [3.11]	-0.0001 [-0.06]	-8.6250 [-1.11]	0.0000 [-0.7]	-0.0002 [-0.53]	0.0000 [0.5]	0.0000 [0.98]	0.35
Relative Value	275	0.0004** [2.01]	0.0424*** [3.88]							-0.0009* [-2.4]	0.0005*** [4.14]	0.0000 [0.06]	0.12
		0.0338 [0.65]	0.0314*** [2.65]	-0.0229 [-0.93]	0.0267 [1.44]	0.0168 [0.65]	0.0021 [1]	2.7385 [0.28]	-0.0001 [-0.96]	<b>-0.0011** [-2.25]</b>	0.0004*** [3.47]	0.0000 [0.27]	0.10

Table 3: HF sensitivity to VOL of VOL over 2008-2013

# Second Result - VOL sensitivity arises from VOL of VOL

Investment Style	Nb. Obs.	Intercept	MKT-RF	SMB	dTERM	dCREDIT	PTFSBD	PTFSFX	PTFSCOM	VoV	SoV	KoV	R-Square/ Adj. R-Square
Global	71	0.0007 [1,51]	0.0939*** [6,54]							-0.003*** [-3,41]	0.0004 [1,65]	0.0000 [0,59]	0.49 0.46
		0.0571 [0,74]	0.1044*** [5,3]	-0.0100 [-0,24]	0.0136 [0,46]	0.0288 [0,75]	0.0116* [1,91]	1.6329 [0,08]	-0.0002 [-1,58]	<b>-0.0032***</b> <b>[ -3,01]</b>	0.0004 [1,46]	0.0000 [0,47]	0.53 0.46
Directionnal	71	0.0012 [1,53]	0.2318*** [9,49]							-0.0020 [-1,33]	0.0011** [2,47]	0.0000 [0,73]	0.60 0.58
		0.1378 [1,08]	0.207*** [6,35]	-0.0845 [-1,24]	0.0725 [1,49]	0.0684 [1,07]	0.0146 [1,45]	-32.9534 [-0,97]	-0.0004** [-2,23]	-0.0015 [-0,84]	0.0009* [1,91]	0.0000 [0,4]	0.66 0.60
Equity Hedge	71	0.0010 [1,4]	0.207*** [9,47]							-0.0034** [-2,59]	0.0005 [1,26]	0.0000 [0,68]	0.62 0.60
		0.1467 [1,26]	0.2203*** [7,4]	-0.0033 [-0,05]	0.0464 [1,05]	0.0741 [1,27]	0.0182* [1,98]	5.4085 [0,17]	-0.0002 [-1,26]	<b>-0.0039**</b> <b>[ -2,46]</b>	0.0005 [1,08]	0.0000 [0,69]	0.66 0.60
Macro	71	-0.0005 [-0,73]	-0.0637*** [-3,12]							-0.0028** [-2,26]	-0.0011*** [-3]	0.0001 [1,1]	0.24 0.19
		-0.0696 [-0,69]	-0.0372 [-1,45]	-0.0290 [-0,54]	-0.0646* [-1,69]	-0.0359 [-0,72]	-0.0027 [-0,34]	-40.9889 [-1,53]	-0.0002 [-1,45]	-0.0017 [-1,22]	-0.0007* [-1,89]	0.0000 [0,14]	0.41 0.32
Merger Arbitrage	71	0.0012** [2,64]	0.1244*** [8,39]							-0.0009 [-0,95]	-0.0002 [-0,66]	0.0001** [2,3]	0.57 0.54
		0.2653*** [3,62]	0.1139*** [6,1]	-0.0527 [-1,35]	0.0978** [3,52]	0.1321*** [3,62]	0.0058 [1,01]	-19.6823 [-1,01]	-0.0001 [-0,78]	-0.0011 [-1,11]	-0.0002 [-0,81]	0.0001** [2,28]	0.67 0.61
Relative Value	71	0.0012* [1,68]	0.064*** [2,8]							-0.0033** [-2,36]	0.0011*** [2,75]	0.0000 [-0,11]	0.23 0.18
		0.0937 [0,78]	0.0662** [2,16]	0.0199 [0,31]	0.0477 [1,04]	0.0481 [0,8]	0.0191** [2,02]	31.5392 [0,98]	-0.0001 [-0,82]	<b>-0.0042**</b> <b>[ -2,52]</b>	0.0009** [2,02]	0.0000 [0,32]	0.32 0.21

Table 4: HF sensitivity to VOL of VOL over crisis periods

## Second Result - VOL sensitivity arises from VOL of VOL

At **HF investment style** level, after controlling for Fung-Hsieh factors, we estimate time-series OLS regressions:

- A one-standard deviation increase in **VOL of VOL** premia is associated with a substantial decline in aggregate HF returns of **25.2%** annually
- **Relative Value** and **Equity Hedge** are the most negatively exposed strategies to **VOL of VOL**, particularly during crises when volatility swap returns are the highest
- Relative Value are usually considered as the last insurer against tail risks, executing risk transfer from financial institutions
- Whereas Equity Hedge usually overlay hedge long positions; the 2 payoff return profiles equal to buying a call hedged by **selling realized volatility**

# Third Result - HF exposed to higher-order moments

Investment Style	Nb. Obs.	Intercept	MKT-RF	SMB	dTERM	dCREDIT	PTFSBD	PTFSFX	PTFSCOM	VoV	SoV	KoV	R-Square/ Adj. R-Square
Global	275	0.0000 [-0.19] -0.0231 [-0.63]	0.088*** [11.48] 0.0872*** [10.45]	0.0064 [0.37]	-0.0052 [-0.4]	-0.0114 [-0.63]	0.0024 [1.6]	4.3693 [0.63]	-0.0001** [-2]	-0.0012*** [-4.26] -0.001*** [-3.4]	0.0002** [2.09] 0.0001* [1.8]	0.0000 [0.51] 0.0000 [0.36]	0.41 0.41 0.43 0.41
Directionnal	275	0.0003 [1.22] -0.0407 [-0.66]	0.1979*** [15.11] 0.1908*** [13.52]	-0.0623*** [-2.13]	-0.0027 [-0.12]	-0.0206 [-0.67]	0.0027 [1.06]	1.2206 [0.1]	-0.0002** [-2.24]	-0.0009* [-1.85] -0.0007 [-1.36]	0.0004*** [3.34] <b>0,0004***</b> <b>[2,68]</b>	0.0000 [1.5] 0.0000 [1.48]	0.50 0.49 0.52 0.50
Equity Hedge	275	-0.0001 [-0.22] 0.0096 [0.16]	0.1973*** [15.54] 0.1906*** [13.88]	0.0084 [0.29]	0.0153 [0.71]	0.0051 [0.17]	0.0044* [1.79]	16.4733 [1.43]	-0.0001* [-1.66]	-0.0013*** [-2.96] -0.0013*** [-2.63]	0.0001 [1.06] 0.0001 [0.52]	0.0000 [0.49] 0.0000 [0.57]	0.54 0.53 0.55 0.54
Macro	275	-0.0009*** [-3.2] -0.0991 [-1.54]	-0.0321** [-2.23] 0.0866*** [2.83]	-0.0691*** [-3]	-0.0495 [-1.54]	0.0018 [0.69]	-17.9721 [-1.46]	-0.0001 [-1.27]	-0.0011** [-2.23] -0.0005 [-0.95]	-0.0004** [-2.5] <b>-0,0002</b> <b>[1,35]</b>	0.0000 [-0.01] 0.0000 [-0.82]	0.0000 [0.01] 0.0000 [0.15]	0.05 0.03 0.18 0.15
Merger Arbitrage	275	0.0002 [1.21] 0.1272*** [3.12]	0.0979*** [11.3] 0.0901*** [9.68]	-0.0272 [-1.41]	0.0497*** [3.42]	0.0633*** [3.11]	-0.0001 [-0.06]	-8.6250 [-1.11]	0.0000 [-0.7]	-0.0001 [-0.47] -0.0002 [-0.53]	0.0001 [0.62] 0.0000 [0.5]	0.0000 [0.88] 0.0000 [0.98]	0.36 0.35 0.39 0.37
Relative Value	275	0.0004** [2.01] 0.0338 [0.65]	0.0424*** [3.88] 0.0314*** [2.65]	-0.0229 [-0.93]	0.0267 [1.44]	0.0168 [0.65]	0.0021 [1]	2.7385 [0.28]	-0.0001 [-0.96]	-0.0009** [-2.4] <b>-0,001**</b> <b>[2,25]</b>	0.0005*** [4.14] <b>0,0004***</b> <b>[3,47]</b>	0.0000 [0.06] 0.0000 [0.27]	0.12 0.10 0.15 0.11

Table 5: HF sensitivity to SKEW/KURT of VOL over 2008-2013

# Third Result - HF exposed to higher-order moments

Investment Style	Nb. Obs.	Intercept	MKT-RF	SMB	dTERM	dCREDIT	PTFSBD	PTFSFX	PTFS COM	VoV	SoV	KoV	R-Square/ Adj. R-Square
Global	71	0.0007 [1,51]	0.0939*** [6,54]							-0.003*** [-3,41]	0.0004 [1,65]	0.0000 [0,59]	0.49 0.46
		0.0571 [0,74]	0.1044*** [5,3]	-0.0100 [-0,24]	0.0136 [0,46]	0.0288 [0,75]	0.0116* [1,91]	1,6329 [0,08]	-0.0002 [-1,58]	-0.0032*** [-3,01]	0.0004 [1,46]	0.0000 [0,47]	0.53 0.46
Directionnal	71	0.0012 [1,53]	0.2318*** [9,49]							-0.0020 [-1,33]	0.0011** [2,47]	0.0000 [0,73]	0.60 0.58
		0.1378 [1,08]	0.207*** [6,35]	-0.0845 [-1,24]	0.0725 [1,49]	0.0684 [1,07]	0.0146 [1,45]	-32,9534 [-0,97]	-0.0004** [-2,23]	-0.0015 [-0,84]	0,0009** [1,91]	0.0000 [0,4]	0.66 0.60
Equity Hedge	71	0.0010 [1,4]	0.207*** [9,47]							-0.0034** [-2,59]	0.0005 [1,26]	0.0000 [0,68]	0.62 0.60
		0.1467 [1,26]	0.2203*** [7,4]	-0.0033 [-0,05]	0.0464 [1,05]	0.0741 [1,27]	0.0182* [1,98]	5,4085 [0,17]	-0.0002 [-1,26]	-0.0039** [-2,46]	0.0005 [1,08]	0.0000 [0,69]	0.66 0.60
Macro	71	-0.0005 [-0,73]	-0.0637*** [-3,12]							-0.0028** [-2,26]	-0.0011*** [-3]	0.0001 [1,1]	0.24 0.19
		-0.0696 [-0,69]	-0.0372 [-1,45]	-0.0290 [-0,54]	-0.0646* [-1,69]	-0.0359 [-0,72]	-0.0027 [-0,34]	-40,9889 [-1,53]	-0.0002 [-1,45]	-0.0017 [-1,22]	-0,0007** [-1,89]	0.0000 [0,14]	0.0000 [0,32]
Merger Arbitrage	71	0.0012** [2,64]	0.1244*** [8,39]							-0.0009 [-0,95]	-0.0002 [-0,66]	0.0001** [2,3]	0.57 0.54
		0.2653*** [3,62]	0.1139*** [6,1]	-0.0527 [-1,35]	0.0978*** [3,52]	0.1321*** [3,62]	0.0058 [1,01]	-19,6823 [-1,01]	-0.0001 [-0,78]	-0.0011 [-1,11]	-0.0002 [-0,81]	0,0001*** [2,28]	0.67 0.61
Relative Value	71	0.0012* [1,68]	0.064*** [2,8]							-0.0033** [-2,36]	0.0011*** [2,75]	0.0000 [0,11]	0.23 0.18
		0.0937 [0,78]	0.0662** [2,16]	0.0199 [0,31]	0.0477 [1,04]	0.0481 [0,8]	0.0191** [2,02]	31,5392 [0,98]	-0.0001 [-0,82]	-0,0042** [-2,52]	0,0009** [2,02]	0.0000 [0,32]	0.32 0.21

Table 6: HF sensitivity to SKEW/KURT of VOL over crisis periods

## Third Result - HF exposed to higher-order moments

At **HF investment style** level, after controlling for Fung-Hsieh factors, we estimate time-series OLS regressions:

- **Relative Value** and **Directional** are the most positively exposed to SKEW of VOL, since they are usually skewness-buyers strategies.
- **Relative Value** are not completely insurance-sellers strategies, since they partially hedge VOL of VOL exposure by buying SKEW of VOL.
- Whereas **Global Macro** are usually negatively exposed to SKEW of VOL

Relative Value and Directional profit from VOL of VOL: Relative Value is long gamma, and trend-followers aim to buying max lookback straddles. It equals to **buying realized skewness**. Alternatively, Global Macro takes contrarian bets on tail risks, i.e. **selling realized SKEW** during crises: convergence trades based on long term macro-trends

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# Conclusion

- Provides a new evidence for **volatility risk** in HF performance
- Showing it is an instrumental determinant in both **time-series** and **cross-section**
- To what extent **HF alpha** actually arises from volatility exposure
- Decompose HF volatility risk from **multiple** and **tradable** option-based strategies
- Decomposed into **volatility, skewness, and kurtosis** risk premia
- Indeed, divergent swaps are now **widely** used by HF managers
- Most HF styles **sell** volatility
- But VOL risk exposures across HF are widely **distinguishable**, depending on specific strategies

*Thank you*