

Option-Implied Crash Index

Jun Pan

**Shanghai Advanced Institute of Finance (SAIF)
Shanghai Jiao Tong University**

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Joint work with Junxiong Gao from SAIF, Shanghai Jiao Tong University

- The seminal work of Black, Scholes, and Merton:
 - ▶ Pure diffusion: Black and Scholes (1973), Merton (1973).
 - ▶ Jump diffusion: Merton (1976).
 - ▶ Pure jump: Cox and Ross (1976).
- Models incorporating stochastic volatility and jump risk:
 - ▶ Stochastic volatility:
Hull and White (1987), Stein and Stein (1991), Heston (1993).
 - ▶ Jump-diffusion models:
Bakshi, Cao, and Chen (1998), Bates (2000), Duffie, Pan and Singleton (2000).
- Empirical evidence:
 - ▶ Using cross-sectional options: Bakshi, Cao, and Chen (1998).
 - ▶ Using joint-times series of options and stock prices: Pan (2002).

Models, Data, and Risk Measures

- The complete-market setting of Black, Merton, and Scholes:
 - ▶ Model parameter: the diffusion coefficient σ .
 - ▶ Application: the option-implied volatility index (VIX).
- The jump-diffusion models with the jump component of Merton (1976):
 - ▶ Model parameter: the mean jump size μ .
 - ▶ Application: the option-implied crash index (CIX).
- The non-parametric approach of Breeden and Litzenberger (1978):
 - ▶ Ait-Sahalia and Lo (1998): the state-price density.
 - ▶ CBOE and Goldman Sachs (2003): the updated VIX index.
 - ▶ Bakshi, Kapadia, and Madan (2003): option-implied skew.

The Stochastic Volatility Model with Jump (SVJ)

- The data generating process:

$$dS_t = \left(r_t - q_t + \eta^s V_t + \lambda V_t (\mu - \mu^*) \right) S_t dt + \sqrt{V_t} S_t dW_t^{(1)} + dZ_t - \mu S_t \lambda V_t dt$$
$$dV_t = \kappa_v (\bar{v} - V_t) dt + \sigma_v \sqrt{V_t} \left(\rho dW_t^{(1)} + \sqrt{1 - \rho^2} dW_t^{(2)} \right)$$

- The risk-neutral dynamics:

$$dS_t = (r_t - q_t) S_t dt + \sqrt{V_t} S_t dW_t^{(Q1)} + dZ_t^Q - \mu^* S_t \lambda V_t dt$$
$$dV_t = \left(\kappa_v (\bar{v} - V_t) + \eta^v V_t \right) dt + \sigma_v \sqrt{V_t} \left(\rho dW_t^{(Q1)} + \sqrt{1 - \rho^2} dW_t^{(Q2)} \right)$$

- The market prices of risks:

- ▶ The diffusion risk premium in index returns: $\eta^s V_t$.
- ▶ The jump risk premium in index returns: $\lambda V_t (\mu - \mu^*)$.
- ▶ The volatility risk premium: $\eta^v V_t$.

Option Pricing Under the SVJ Model

- The time- t price of a European-style call option:

$$C_t^{\text{SVJ}} = \mathbb{E}_t^Q \left[\exp \left(- \int_t^T r_u \, du \right) (S_T - K)^+ \right] = S_t f \left(V_t, \vartheta, r_t, q_t, \tau, \frac{K}{S_t} \right)$$

- ▶ K is the strike price and $T = t + \tau$ is the expiration date.
 - ▶ The latent state variable: V_t .
 - ▶ The model parameters: $\vartheta = (\kappa_v, \bar{v}, \sigma_v, \rho, \lambda, \mu, \sigma_J, \eta^s, \eta^v, \mu^*)$.
- Via put/call parity, the time- t price of a European-style put option:

$$P_t^{\text{SVJ}} = S_t f \left(V_t, \vartheta, r_t, q_t, \tau, \frac{K}{S_t} \right) - S_t e^{-q\tau} + K e^{-r\tau} .$$

Model Estimation

- We use the joint time-series of the S&P 500 index and options (Pan 2002).
- The moment conditions are constructed on the two state variables:
 - ▶ Daily index returns: $y_t = \ln S_t - \ln S_{t-1} - r_t - q_t$
 - ▶ Time- t volatility V_t , inferred from an ATM call, given model parameters ϑ ,

$$C_t^{\text{Market}} = C_t^{\text{SVJ}} = S_t f \left(V_t, \vartheta, r_t, q_t, \tau, \frac{K}{S_t} \right).$$

- The model parameters ϑ and the latent V_t are simultaneously estimated,
 - ▶ using moment conditions based on

$$\begin{aligned} \varepsilon_t^{y1} &= y_t - M_1(V_{t-1}, \vartheta) & \varepsilon_t^{v1} &= V_t - M_5(V_{t-1}, \vartheta) \\ \varepsilon_t^{y2} &= y_t^2 - M_2(V_{t-1}, \vartheta) & \varepsilon_t^{v2} &= V_t^2 - M_6(V_{t-1}, \vartheta) \\ \varepsilon_t^{y3} &= y_t^3 - M_3(V_{t-1}, \vartheta) & \varepsilon_t^{yv} &= y_t V_t - M_7(V_{t-1}, \vartheta) \\ \varepsilon_t^{y4} &= y_t^4 - M_4(V_{t-1}, \vartheta), \end{aligned}$$

- ▶ and pricing errors of 30-day ITM calls and 90-day ATM calls.

Estimation Results: Model Parameters

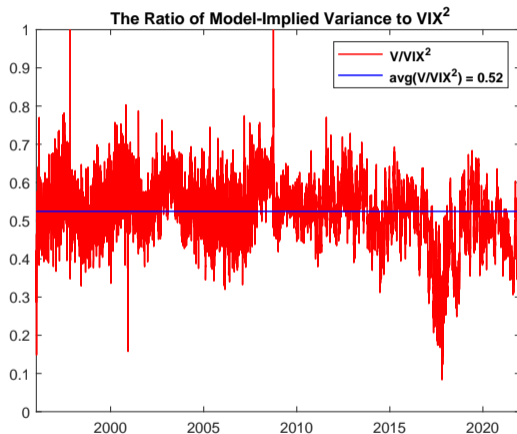
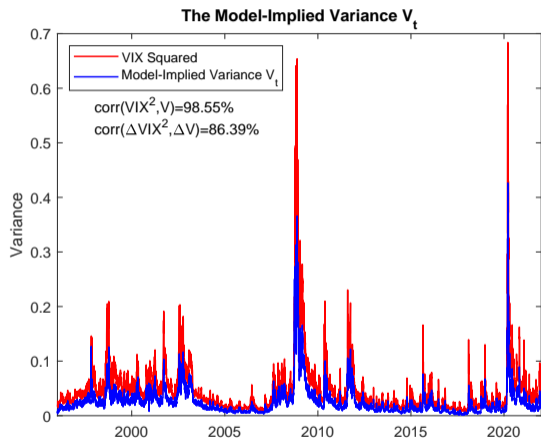
Panel A: SVJ estimation using 1996-2021 daily price data

k_v	\bar{v}	σ_v	ρ	λ	μ	σ_J	η_s	η_v	μ^*
5.88	0.0108	0.29	-0.52	23.9	-1.00%	2.97%	3.10	3.19	-16.16%
[45.12]	[8.70]	[15.54]	[-15.11]	[13.71]	[-0.14]	[5.83]	[1.51]	[16.65]	[-37.32]

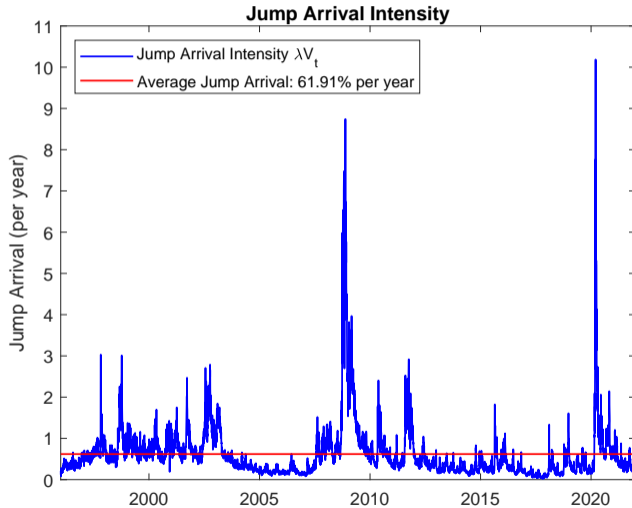
Panel B: SVJ fitting the joint moments of y_t and V_t

y_t	y_t^2	y_t^3	y_t^4	V_t	V_t^2	$y_t V_t$
20.14	0.35	1.15	-0.20	0.03	0.01	1.10
[0.26]	[0.09]	[0.21]	[-0.06]	[0.04]	[0.02]	[0.20]

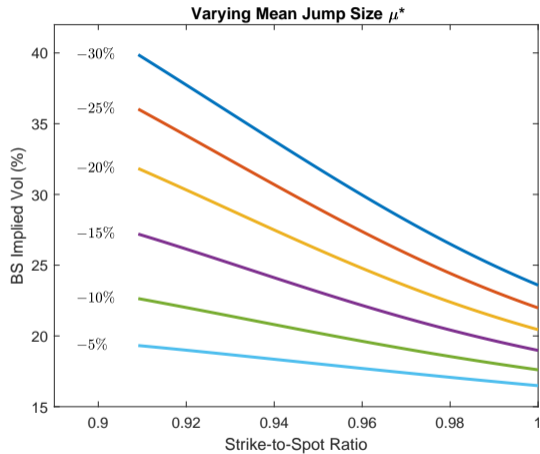
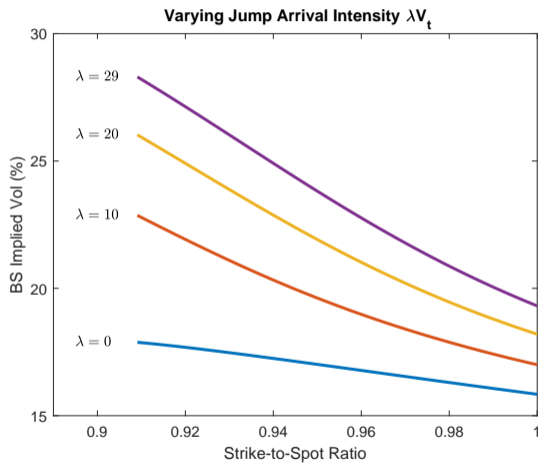
Estimation Results: The Latent State Variable V_t



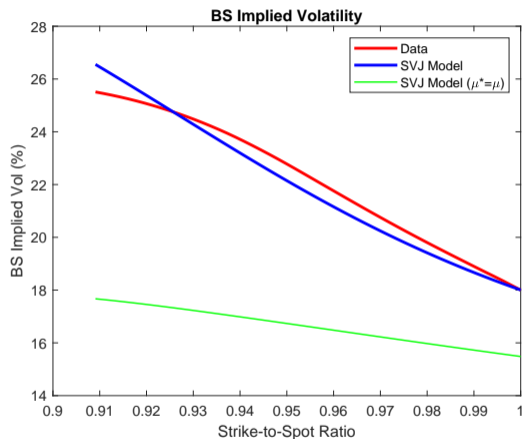
Estimation Results: Model-Implied Jump Arrival Intensity



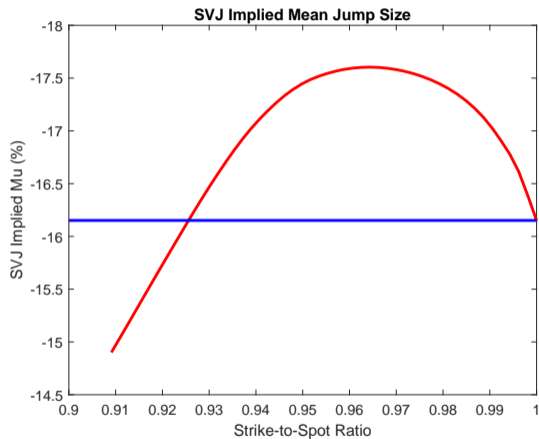
Estimation Results: Model-Implied Volatility Surface



Option-Implied Crash Surface



Vol Surface, Data vs Model



Option-Implied Crash Surface

Construct the Option-Implied Crash Index (CIX)

- Infer the SVJ model implied $\mu_t^I(\tau, K)$ from an OTM put with market price P_t^{Market} ,

$$P_t^{\text{Market}} = P_t^{\text{SVJ}} = S_t f \left(V_t, \vartheta^\perp, \mu_t^I(\tau, K), r, q, \tau, \frac{K}{S_t} \right) - S_t e^{-q\tau} + K e^{-r\tau},$$

where we fix V_t and all other parameters ϑ^\perp to the estimation results.

- This parallels the Black-Scholes implied $\sigma_t^I(\tau, K)$,

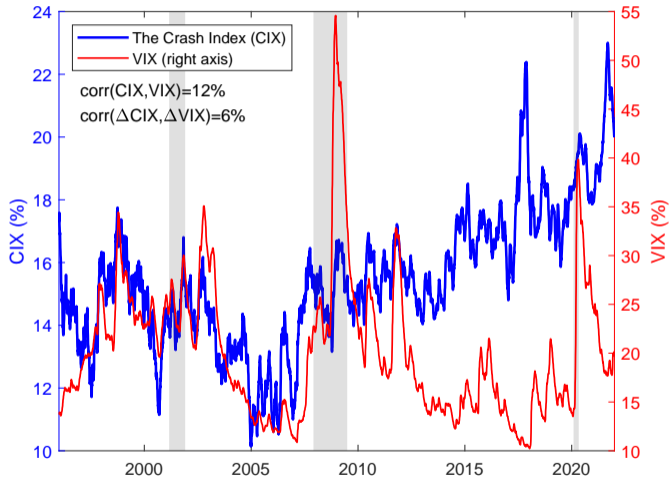
$$P_t^{\text{Market}} = P_t^{\text{BS}} = S_t f \left(\sigma_t^I(\tau, K), r, q, \tau, \frac{K}{S_t} \right) - S_t e^{-q\tau} + K e^{-r\tau}.$$

- Following VIX, we construct CIX by interpolating around the 30-day to expiration,

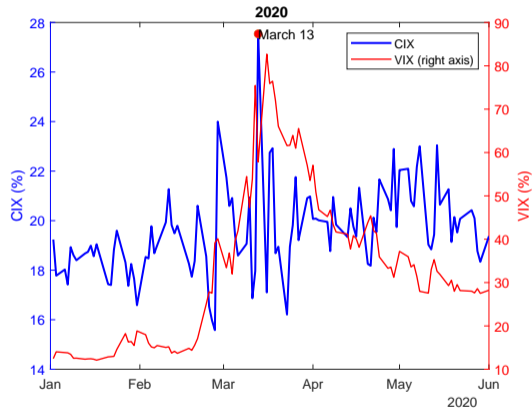
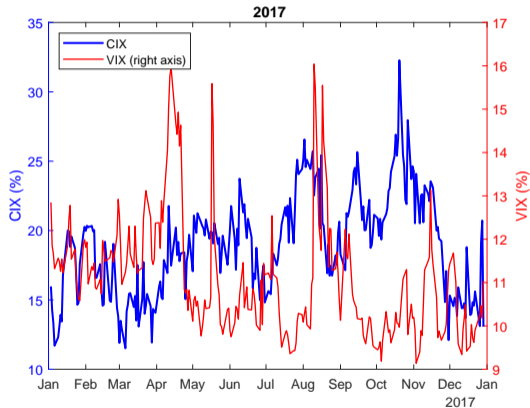
$$-\text{CIX}_t = \mu_t^I(\tau_1) \frac{\tau_2 - 30}{\tau_2 - \tau_1} + \mu_t^I(\tau_2) \frac{30 - \tau_1}{\tau_2 - \tau_1},$$

using $\tau_1 \leq 30 < \tau_2$ and $\mu_t^I(\tau)$ for each τ is averaged across all $K/S \in [0.93, 0.97]$.

Time-Series of CIX vs VIX



Time-Series of CIX vs VIX



Validating CIX using Non-Parametric Skew

- Since 2003, the CBOE VIX is computed via

$$e^{-r\tau} \mathbb{E}_t^Q (R(t, \tau)^2) = \int_{S_t}^{\infty} \frac{2 \left(1 - \ln \left(\frac{K}{S_t}\right)\right)}{K^2} C_t(\tau, K) dK + \int_0^{S_t} \frac{2 \left(1 + \ln \left(\frac{S_t}{K}\right)\right)}{K^2} P_t(\tau, K) dK.$$

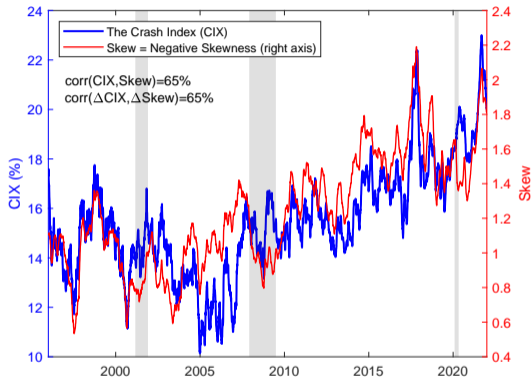
where $R(t, \tau) = \ln S_{t+\tau} - \ln S_t$.

- Bakshi, Kapadia, and Madan (2003) computes the risk-neutral skewness via

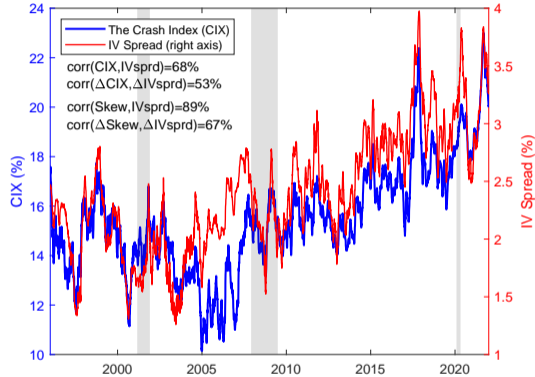
$$e^{-r\tau} \mathbb{E}_t^Q (R(t, \tau)^3) = \int_{S_t}^{\infty} \frac{6 \ln \left(\frac{K}{S_t}\right) - 3 \left(\ln \left(\frac{K}{S_t}\right)\right)^2}{K^2} C_t(\tau, K) dK - \int_0^{S_t} \frac{6 \ln \left(\frac{S_t}{K}\right) + 3 \left(\ln \left(\frac{S_t}{K}\right)\right)^2}{K^2} P_t(\tau, K) dK.$$

- ▶ Crashes: important in generating negative skewness.
- ▶ The non-parametric skewness can be used to validate the information content of our CIX index.

Time-Series of CIX, Skew and IV Spread

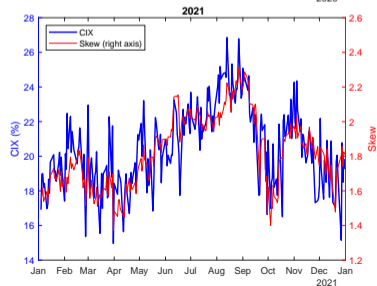
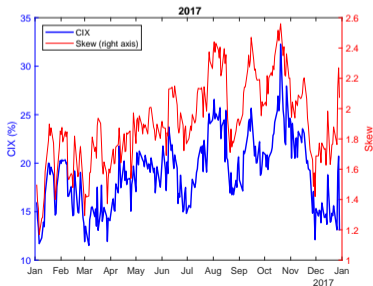
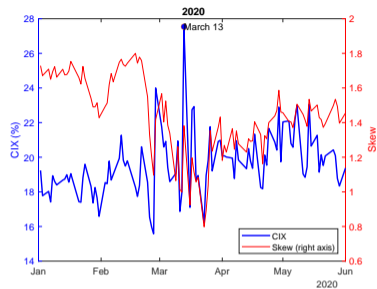
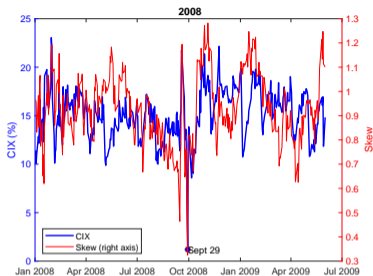


CIX and Skewness

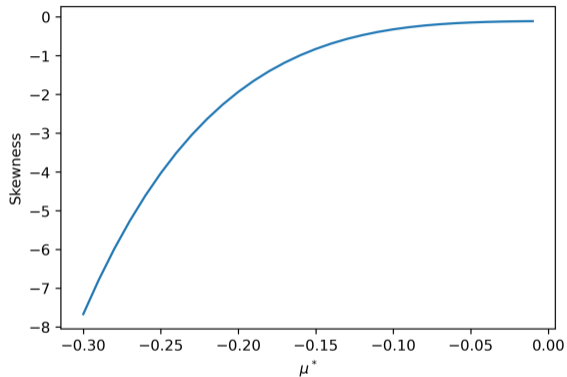


CIX and IV Spread

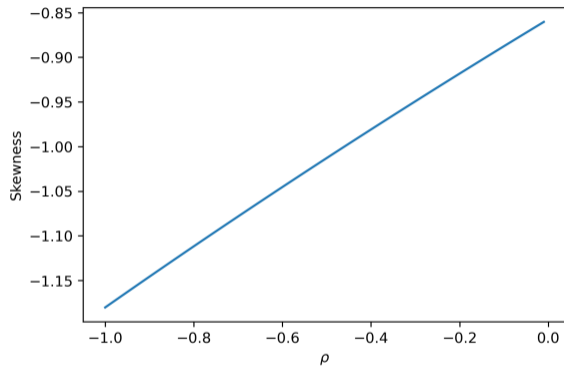
Time-Series of CIX and Skewness



Skewness under the SVJ Model

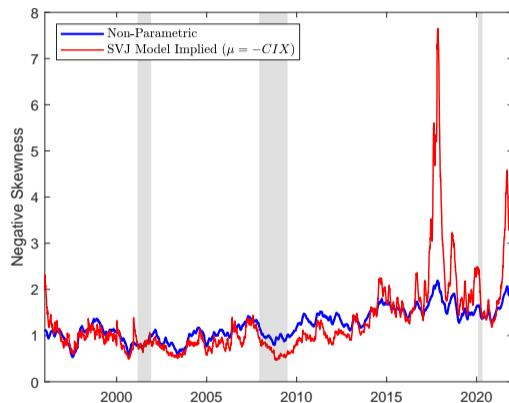
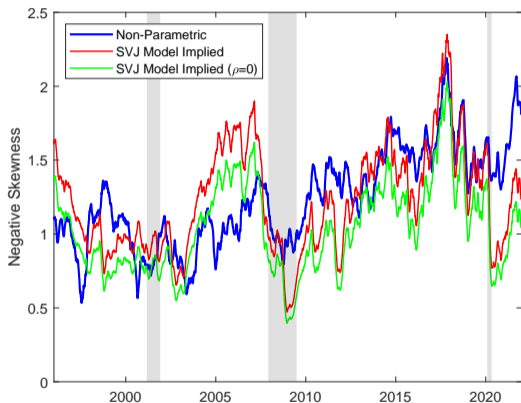


SVJ Skewness with Varying μ^*



SVJ Skewness with Varying ρ

Time-Series of Skewness, Non-Parametric vs the SVJ Model



Explaining the Crash Index (CIX)

	Dependent Variable: ΔCIX_t (%)							
ΔVIX_t	6.37 [3.30]					11.38 [7.63]	10.61 [7.07]	10.50 [7.19]
$\Delta IV\ sprd_t$		59.42 [29.39]				32.32 [14.49]	32.16 [14.14]	32.12 [14.10]
$\Delta Skew_t$			64.60 [31.82]			46.37 [15.37]	46.14 [15.18]	46.31 [15.25]
$\Delta(P/C)_t$				5.53 [3.66]		3.25 [2.92]	3.15 [2.85]	2.99 [2.71]
$\Delta Noise_t$							2.58 [1.66]	2.51 [2.30]
$\Delta Term_t$							-2.70 [-1.95]	-4.00 [-3.32]
ΔTED_t							-1.53 [-1.09]	-0.38 [-0.33]
$\Delta Dsprd_t$							1.03 [0.68]	-0.31 [-0.24]
CIX_{t-1}							-0.35 [-0.34]	-0.39 [-0.38]
VIX_{t-1}							0.01 [0.01]	0.29 [0.27]
Ret_{t-1}							-2.65 [-2.67]	-2.49 [-2.52]
$1_{3rd\ Friday}$							1.72 [2.38]	1.41 [1.95]
adj R^2 (%)	0.39	35.29	41.72	0.29	50.03	0.10	50.10	50.26

Return Predictability Using Option-Implied Risk Measures

Average Daily SPX Returns (bps)									
following large increases					following large reductions				
Prob(Tail)	CIX	VIX	Skew	IVSprd	Prob(Tail)	CIX	VIX	Skew	IVSprd
2%	-47.41 [-2.89]	59.88 [2.03]	-29.08 [-2.03]	-23.87 [-1.47]	2%	20.28 [1.70]	-4.99 [-0.26]	25.22 [1.46]	35.70 [1.92]
5%	-28.96 [-3.39]	29.73 [2.14]	-24.84 [-3.27]	-12.63 [-1.42]	5%	19.55 [2.47]	-3.39 [-0.34]	21.32 [2.33]	30.17 [2.99]
10%	-18.29 [-3.44]	17.52 [2.12]	-13.31 [-2.56]	-9.76 [-1.66]	10%	11.60 [2.24]	4.25 [0.71]	22.08 [3.93]	21.59 [3.52]
15%	-12.16 [-2.89]	14.04 [2.36]	-12.23 [-3.10]	-6.53 [-1.46]	15%	8.32 [2.00]	3.72 [0.83]	19.13 [4.49]	18.99 [4.01]
20%	-10.41 [-2.90]	12.06 [2.52]	-10.42 [-3.09]	-4.77 [-1.26]	20%	6.64 [1.92]	2.97 [0.82]	15.77 [4.40]	17.29 [4.45]

Return Predictability Using Option-Implied Risk Measures

	Dependent Variable: Day $t + 1$ SPX Returns (bps)						
$1_{\Delta CIX_t}$	-24.65 [-4.39]			-20.20 [-4.27]	-23.34 [-2.90]		-3.59 [-2.34]
$1_{\Delta VIX_t}$		15.14 [2.32]		22.62 [3.08]			
$1_{\Delta SKEW_t}$			-19.11 [-3.71]		-13.11 [-2.01]		
$1_{\Delta CIX_t} \times 1_{\Delta VIX_t}$				-36.98 [-1.51]			
$1_{\Delta CIX_t} \times 1_{\Delta SKEW_t}$					9.16 [0.68]		
ΔCIX_t						-6.29 [-4.20]	-4.11 [-2.44]
constant	6.36 [4.76]	2.37 [1.75]	5.81 [4.46]	4.25 [3.07]	7.07 [5.13]	3.21 [3.08]	4.40 [3.89]
$R^2(\%)$	0.37	0.14	0.22	0.65	0.43	0.40	0.48

Conclusions

- We construct an option-implied Crash Index (CIX) by exploring the pricing difference between the OTM puts and ATM options, benchmarked against the SVJ model.
- The construction of our CIX index is analogous to that of the VIX index:
 - ▶ The mean jump size μ of the SVJ model implied by OTM puts.
 - ▶ The volatility parameter σ of the Black-Scholes model implied by ATM options.
- Empirically, we find that
 - ▶ The CIX index is closely related to the non-parametric option-implied skewness and positively correlated with the put/call volume ratio.
 - ▶ Post 2008, the CIX index has increased significantly.
 - ▶ Large increases in CIX are followed by large negative SPX returns.
 - ▶ By contrast, large increases in VIX are followed by large positive SPX returns.