iCOS: Option-Implied COS Method

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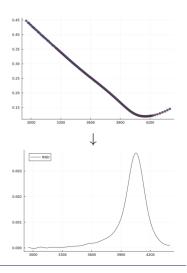
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Introduction

- Option price (panel) data contain valuable information about investor expectations on future asset prices;
- From option prices, one can obtain information on the (risk-neutral) density, moments, volatility, etc;
- > This rich information, however, is often challenging to extract
 - the non-linear nature of option prices and various sources of uncertainty add complexity and computational costs to the analysis

RND from options

- Observed option prices convert to BSIV
- BSIV 'curve-fitting' (parametric or non-parametric)
- Convert the curve back to options
- Obtain the RND by calculating the second-order derivative of a (converted) 'curve'.



In this paper

- We propose a unified non-parametric estimation procedure for the RND, option prices and option sensitivities (option delta);
- Our approach leverages the Fourier-based cosine technique, the COS method, proposed by Fang and Oosterlee (2008), in a model-free way by *implying* information from observed option contracts.
- The proposed estimation method is fully non-parametric and does not require any optimization routines, offering a flexible and computationally appealing alternative to traditional techniques.

- The COS method of Fang and Oosterlee (2008) is the Fourier-based based method for evaluation of option prices;
- Fourier cosine series expansion of the density function *f*(*y*) on an interval [*a*, *b*] ⊂ ℝ:

$$f(y) = \frac{2}{b-a} \sum_{m=0}^{\infty} A_m \cos(u_m y - u_m a),$$
(1)

with $u_m := \frac{m\pi}{b-a}$ and the cosine coefficients A_m :

$$A_m = \int_a^b \cos(u_m y - u_m a) f(y) \mathrm{d}y.$$
⁽²⁾

Fang and Oosterlee (2008): the cosine coefficients A_m can be approximated via the characteristic function (CF), $\phi(u)$, of *y*:

$$A_m = \Re\left\{\int_a^b e^{\mathrm{i} u_m(y-a)}f(y)\mathrm{d} y\right\} \approx \Re\left\{\phi(u_m)e^{-\mathrm{i} u_m a}\right\} =: \widetilde{A}_m.$$

• Define the cosine series coefficients of the payoff function v(y, T) as

$$H_m := \frac{2}{b-a} \int_a^b v(y,T) \cos(u_m y - u_m a) \mathrm{d}y.$$

deterministic function of the payoff function, pre-computed for put and call.

- Consider a European option with a payoff v(y, T) maturing at time *T*;
- ▶ Its price under the risk-neutral measure Q is given by the COS as:

$$v_{0} = e^{-rT} \mathbb{E}^{\mathbb{Q}}[v(y,T)] \stackrel{\scriptscriptstyle (1)}{\approx} e^{-rT} \int_{a}^{b} v(y,T) f(y) dy$$
$$\stackrel{\scriptscriptstyle (2)}{\approx} e^{-rT} \sum_{m=0}^{\infty} \widetilde{A}_{m} H_{m} \stackrel{\scriptscriptstyle (3)}{\approx} e^{-rT} \sum_{m=0}^{N-1} \widetilde{A}_{m} H_{m},$$

where by $\stackrel{\scriptscriptstyle(i)}{\approx}$ we denote the subsequent numerical approximation.

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where by $\stackrel{(i)}{\approx}$ we denote the subsequent numerical approximation.

- ▶ Importantly, it requires a parametric CF, which is unknown a priory;
- However, we observe option prices in the market!

Information in options

Breeden and Litzenberger (1978): call (and put) option on underlying asset price S_T with strike K and time to expiration T:

$$C_0(K) = e^{-rT} \mathbb{E}^{\mathbb{Q}}[\max(S_T - K, 0)] = e^{-rT} \int_K^\infty (x - K) f_S(x) dx,$$

implies risk-neutral density

$$f_{S}(x) = e^{rT} C_{0}''(x) = e^{rT} \left. \frac{\partial^{2} C_{0}(K)}{\partial K^{2}} \right|_{K=x}.$$
(3)

Payoff spanning

Consider a general payoff function v(S_T) for a European-style option
 Value v₀ at time t = 0 for this option is

$$v_0 = e^{-rT} \mathbb{E}^{\mathbb{Q}}[v(S_T)] = \int_0^\infty v(x) f_S(x) \mathrm{d}x.$$

Carr & Madan (2001): replicating portfolio of this contract:

$$v_0 = e^{-rT}v(F_0) + \int_0^\infty v''(K)O_0(K)dK$$

with OTM options $O_0(K) = \min\{C_0(K), P_0(K)\}$ and futures price F_0 .

Option-implied cosine coefficients

Transformation $y = \log \frac{S_T}{x}$, $a = \log \frac{\alpha}{x}$ and $b = \log \frac{\beta}{x}$ with x > 0, and notice

$$A_{m} = \int_{a}^{b} \cos\left(u_{m}y - u_{m}a\right) f(y) dy \qquad (4)$$
$$= \int_{\alpha}^{\beta} \cos\left(u_{m}\log\frac{S_{T}}{\alpha}\right) f_{S}(S_{T}) dS_{T} = \mathbb{E}^{\mathbb{Q}} \left[\underbrace{\cos\left(u_{m}\log\frac{S_{T}}{\alpha}\right) \mathbf{1}_{\{\alpha \leq S_{T} \leq \beta\}}}_{=:v(S_{T})}\right],$$

which can be spanned as a portfolio of options with $v(S_T)$:

$$e^{-rT}A_m = \underbrace{e^{-rT}\cos\left(u_m\log\frac{F}{\alpha}\right) + \int_{\alpha}^{\beta}\psi_m(K)O_0(K)dK}_{=:D_m} + \underbrace{(-1)^mC'_K(\beta) - P'_K(\alpha)}_{=:b_m}.$$

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Option-implied cosine coefficients

- $C'_K(\beta)$ and $P'_K(\alpha)$ are call and put price derivatives w.r.t. strike *K* and $\psi_m(x) := v''(x)$
- The option-implied cosine coefficients D_m are exact and completely model-free;
- But, require a continuum of options with strikes $K \in [\alpha, \beta]$;
- D_m can be estimated by \widehat{D}_m given *n* observed OTM option prices $O(K_i)$, i = 1, ..., n with $\alpha := K_1 < ... < K_n =: \beta$.

▶ In fact,
$$\widehat{D}_m \xrightarrow{P} D_m$$
 as $n \to \infty$.

Option-implied call price

The call price $C_0(x)$ with the strike price x: $\alpha \le x \le \beta$ can be decomposed as

$$C_0(x) = C_0^{[\alpha,\beta]}(x) + C_0^{(\beta,\infty)}(x) = C_0^{[\alpha,\beta]}(x) + (x-\beta)C'_K(\beta) + C_0(\beta),$$

C₀^[α,β] is the value of the call on the interval [α, β];
 ⇒ can be evaluated exactly using the COS expansion:

$$C_0(x) = \sum_{m=0}^{\infty} D_m H_m(x) + C_0(\beta) + Z_c(x)C'_K(\beta) + Z_p(x)P'_K(\alpha).$$
(5)

Z_c(x) and *Z_p(x)* some deterministic functions of *x*;
 (5) is exact but circular ⇒ useful for interpolation/approximation;

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In practice

1. Given a finite number of option prices, estimate \hat{D}_m , m = 1, ..., N;

2. estimate
$$\widehat{\overline{C}}(x) := \sum_{m=0}^{N} \widehat{D}_m H_m(x)$$
 for $x \in [\alpha, \beta]$;

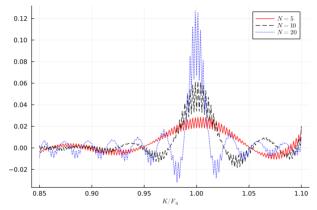
3. regress
$$C(K_i) - \widehat{\overline{C}}(K_i) - C(\beta)$$
 on $Z(K_i) := (1, Z_c(K_i), Z_p(K_i))$ to get
 $\widehat{\theta} := (\widehat{\theta}, \widehat{C}'_K(\beta), \widehat{P}'_K(\beta))';$

4. obtain $\widehat{C}(x) := \widehat{\overline{C}}(x) + C(\beta) + Z(x)\widehat{\theta}$ for any strike $x \in [\alpha, \beta]$.

Option-implied call price

- The call price estimator C
 ^(x) is model-free and eliminates approximation errors (1) and (2);
- In contrast to the COS method, the choice of the interval [a, b] is data-driven;

Portfolio weights for ATM call option



The illustration is based on the simulated Black-Scholes model with strike prices between 85% and 110% of the spot price with equidistant increments and 201 option contracts.

Assumptions

- 1. The smallest and largest strike prices are fixed at $\alpha = K_1$ and $\beta = K_n$, and the strike prices between them are equidistant, i.e. $\Delta K := K_i - K_{i-1} = \frac{\beta - \alpha}{n-1}, i = 2, ..., n.$
- 2. Option prices are observed with an additive error term:

$$O(K_i) = O_0(K_i) + \varepsilon_i, \quad i = 1, \dots, n,$$

where ε_i are mean-zero, conditionally independent, and heteroskedastic across strikes.

3. The RND of future prices $f_S(s) \in C^p([\alpha, \beta])$ with p > 1.

Option-implied call

Proposition

Under Assumptions 1–3, the computationally feasible option-implied call price estimator $\widehat{C}(x)$ with a strike price $x \in [\alpha, \beta]$ is such that as $n \to \infty$ and $N \to \infty$ with $Nn^{-1/2} \to 0$ $\frac{\widehat{C}(x) - C_0(x)}{\sigma_c(x)} \xrightarrow{d} \mathcal{N}(0, 1)$,

with a closed-form expression for the variance $\sigma_c^2(x)$.

Risk-Neutral Density

Proposition

Under Assumptions 1–3, the computationally feasible option-implied RND estimator of the future log price $y = \log S_T$:

$$\widehat{f}(y) = \nu_f \sum_{m=0}^{N'} \left(\widehat{D}_m + (-1)^m \widehat{C}'_K(\beta) - \widehat{P}'_K(\alpha) \right) \cos\left(u_m y - u_m \log \alpha\right)$$
(6)

with $v_f := 2e^{rT} \log (\alpha / \beta)$ and is such that for any fixed y as $n \to \infty$ and $N \to \infty$ with $Nn^{-1/6} \to 0$ $\frac{\widehat{f}(y) - f(y)}{v_f \sigma_f(y)} \xrightarrow{d} \mathcal{N}(0, 1)$,

where $\sigma_f^2(y)$ is the variance term.

Option-implied delta

The implied, model-free delta for a European call can also be obtained via a portfolio of options:

$$\Delta(x) = \frac{\partial C_0(x)}{\partial S_0} = -\frac{1}{S_0} \sum_{m=1}^{\infty} u_m B_m H_m(x) + \frac{1}{S_0} \left(C_0(\beta) - \beta C'_K(\beta) \right).$$

- Coefficients $B_m := e^{-rT} \mathbb{E}^{\mathbb{Q}} \left[\sin \left(u_m \log \frac{S_T}{\alpha} \right) \mathbf{1}_{\{\alpha \leq S_T \leq \beta\}} \right]$ are spanned in a similar way as A_m ;
- The non-parametric estimator for the option's Delta can also be derived similarly:

$$\widehat{\Delta}(x) = -\frac{1}{S_0} \sum_{m=1}^N u_m \widehat{B}_m H_m(x) + \frac{1}{S_0} \left(C(\beta) - \beta \widehat{C}'_K(\beta) \right); \tag{7}$$

An analogous asymptotic result holds for the delta estimator $\widehat{\Delta}(x)$.

iCOS

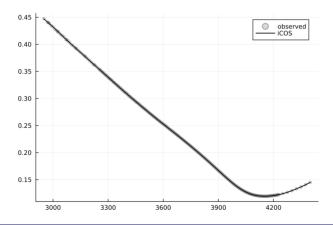
- ▶ iCOS combines the COS machinery with the *option-implied* information;
- non-parametric and based on portfolio replication argument
 - does not do 'curve-fitting' and does not use Breeden and Litzenberger (1978);
 - (almost) optimization-free;
- uses all available information
 - kernel smoothing and local regression models use local information controlled by bandwidth parameter;
- the simulation results indicate good finite performance (more);
- choice of N can be adaptive more.

- SPX options traded on April 1, 2021 with maturity T = 29 days;
- Focus on option with strikes from interval [α, β] = [2950, 4400], while forward F₀ = \$4008.5;
- Use mid-quote option prices as the input for the estimation procedure;
- The number of options n = 239 and Fourier terms N = 23.

Empirical Application

S&P 500 options

Figure 1: SPX options fit displayed on BSIV space



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Figure 2: Pricing errors for call estimates $\widehat{C}(K) - C(K)$

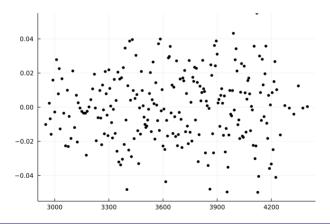
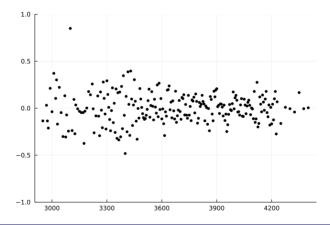
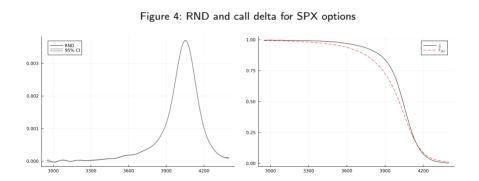


Figure 3: Pricing errors for call estimates $\widehat{C}(K)-C(K)$ relative to the half-spread



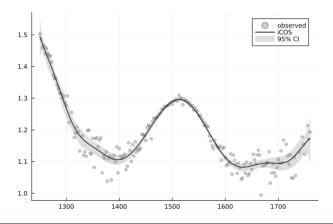


Amazon options

- Equity options on Amazon with a very short time-to-maturity of 1 day;
- Earning Announcement Day (EAD) of April 26, 2018 (before the announcement);
- Atypical W-shaped implied volatility curve and bimodal RND
 much harder to capture with standard curve-fitting methods

Amazon options

Figure 5: AMZN options fit displayed on BSIV space



Amazon options

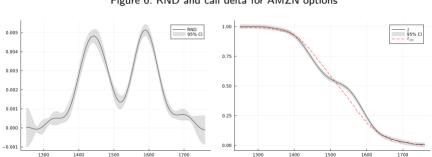


Figure 6: RND and call delta for AMZN options

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Conclusion

- Develop a new non-parametric estimation procedure for the RND, option prices and deltas;
- Based on the COS method with the option-implied information;
- It is model-free, non-parametric and does not require optimization procedures;
- Simulation results suggest good finite-sample performance;
- Application to SPX options and AMZN equity option on EAD.

Simulation setup **back**

The 'double-jump' stochastic volatility model of Duffie, Pan, and Singleton (2000):

$$d \log S_t = (r - \frac{1}{2}v_t - \mu\lambda)dt + \sqrt{v_t}dW_{1,t} + J_t dN_t, dv_t = \kappa(\bar{v} - v_t)dt + \sigma\sqrt{v_t}dW_{2,t} + J_t^v dN_t.$$

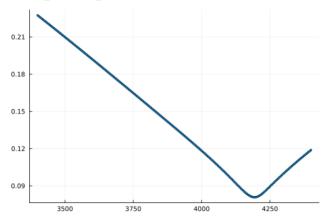
- T = 30 days and the number of options n = 201;
- Add errors into the simulated option prices:

$$O(K_i) = O_0(K_i) + 0.025 \cdot \epsilon, \quad i = 1, ..., n,$$

where ϵ is an i.i.d. standard normal random variable.

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Simulated option prices



Simulated option prices from the SVCJ model displayed on BSIV. The option prices are simulated using the COS method given the analytic solution for the CCF and a large number of expansion terms N = 1024 with $[a, b] = [-4\sqrt{T}, 4\sqrt{T}]$.

Simulation setup

- The number of expansion terms is N = 25.
- Compare with the kernel smoothing method with the following bandwidth:

$$h = \frac{c}{\log n} n^{-\frac{1}{2p+1}}$$

with p = 2 and different constant values c > 0.

Simulation results. Call prices

Table 1: Monte Carlo results for the call prices under the SVCJ model

	K/F_0	0.86	0.9	0.95	1.0	1.05	1.09
$C_0(K)$		560.66	402.23	210.81	54.13	0.61	0.14
iCOS	MC bias	-0.00075	0.00039	-0.00036	3.0e-5	-0.00207	-0.00055
	MC std	0.0102	0.00969	0.00941	0.00926	0.009	0.00933
	As. std	0.00981	0.00944	0.009	0.00909	0.00906	0.00893
KS, <i>c</i> = 0.2	MC bias	-0.07909	-0.00276	-0.00383	-0.10338	0.35918	-0.04309
	MC std	0.00442	0.00427	0.0044	0.00433	0.00637	0.00369
KS, <i>c</i> = 0.1	MC bias	-0.01325	2.0e-5	-0.00027	-0.02496	0.11357	-0.00912
	MC std	0.00621	0.00582	0.00603	0.00589	0.00661	0.00594
KS, <i>c</i> = 0.05	MC bias	-0.00078	0.0005	6.0e-5	-0.00627	0.03224	-0.00178
	MC std	0.00828	0.00811	0.00835	0.00831	0.00839	0.00834
KS, <i>c</i> = 0.03	MC bias	-0.00066	0.00068	0.00024	-0.00249	0.01229	-0.00133
	MC std	0.01043	0.01051	0.01063	0.01061	0.01068	0.01082

Simulation results. RND (back)

	K/F_0	0.86	0.9	0.95	1.0	1.05	1.09
$f(\log K)$		0.13	0.49	2.76	11.37	3.11	0.03
iCOS	MC bias MC std As. std	$0.0211 \\ 0.1503 \\ 0.1462$	0.0032 0.1138 0.1099	0.0058 0.0994 0.095	-0.0066 0.0972 0.0933	0.0711 0.0954 0.091	0.0216 0.0937 0.0854
KS, <i>c</i> = 0.2	MC bias MC std	0.002 0.0033	$0.0511 \\ 0.005$	0.0169 0.006	0.319 0.0702	-0.2341 0.0075	0.0161 0.0043
KS, <i>c</i> = 0.1	MC bias MC std	-0.0999 0.0211	-0.0373 0.0263	-0.1284 0.0286	-0.2761 0.0316	-0.1874 0.0327	-0.0584 0.0285
KS, $c = 0.05$	MC bias MC std	-0.0209 0.1364	-0.0203 0.1469	-0.0609 0.1541	-0.1186 0.1627	-0.1087 0.1743	-0.0186 0.1844
KS, $c = 0.03$	MC bias MC std	0.0063 0.469	-0.0174 0.4929	-0.0296 0.5117	$0.0006 \\ 0.5418$	-0.0703 0.5904	-0.0091 0.6125

Table 2: Monte Carlo results for the RND under the SVCJ model

Cosine coefficients

The cosine coefficients D_m are estimated as

$$\widehat{D}_m := e^{-rT} \cos\left(u_m \log \frac{F}{\alpha}\right) + \sum_{i=1}^n w_i \psi_m(K_i) O(K_i) \Delta_n,\tag{8}$$

where w_i are the coefficients of a chosen numerical integration method.

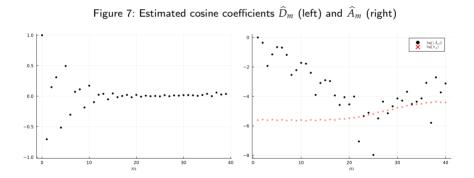
Proposition

Under Assumptions 1–2, $\mathbb{E}\left[\widehat{D}_m - D_m\right] = \zeta_{m,n}^D$, where $\zeta_{m,n}^D$ is the discretization error with the order depending on the chosen numerical integration scheme, and as $n \to \infty$

$$\frac{D_m - D_m}{\sigma_D(m)} \xrightarrow{d} \mathcal{N}(0, 1),$$

with $\sigma_D^2(m) = \sum_{i=1}^n w_i^2 \psi_m^2(K_i) \sigma_i^2 \Delta_n^2$.

S&P 500 options (back)



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