Uniform Confidence for Pricing Kernels

Wolfgang Karl Härdle Yarema Okhrin Weining Wang

Ladislaus von Bortkiewicz Chair of Statistics C.A.S.E.-Center for Applied Statistics and Economics Humboldt-Universität zu Berlin University of Augsburg



Motivation

- Arbitrage free market; riskless bond with rate r
- □ Underlying price process $\{S_t\}$

Price z_t at time t from a payoff $\psi(S_T)$

$$z_{t} = \int_{0}^{\infty} exp(-r\tau)\psi(x)dQ_{S\tau}(x)$$
$$= \int_{0}^{\infty} exp(-r\tau)\psi(x)\frac{q_{t}(x)}{p_{t}(x)}dP_{S\tau}(x)$$

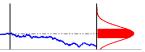


Figure 1: conditional measure at time of maturity T built opon a path of the stochastic process for underlying asset with information up to time t.

Uniform Confidence for PKs

Empirical Pricing Kernel (EPK)

Pricing Kernel (PK) a stochastic discount factor, i.e.

$$\mathcal{K}_{t, au}(x) = exp(-r au) rac{q_t(x)}{p_t(x)}$$

EPK is therefore an estimate of PK:

$$\widehat{\mathcal{K}_{t,\tau}(x)} = exp(-r\tau) \frac{\widehat{q_t(x)}}{\widehat{p_t(x)}}$$

The EPK Paradox

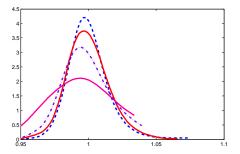


Figure 2: Examples of inter-temporal pricing kernels with maturity 0.00833(3*D*) respectively on 17-Jan-2006 (blue), 18-Apr-2006 (red), 16-May-2006 (magenta), 13-June-2006 (black).

The EPK Paradox

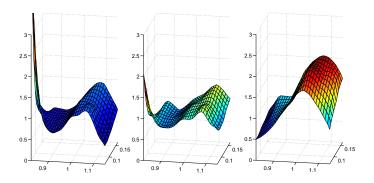


Figure 3: Estimated PK across moneyness and maturity



The EPK Paradox

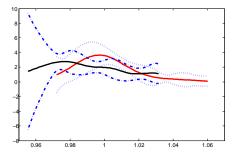


Figure 4: Examples of inter-temporal pricing kernels with various maturities in years: 0.02222 (8D) (red) 0.1(36D) (green) on 12-Jan-2006 and their confidence bands



Aims

- Nonparametric confidence band to test alternatives
- Investigate shapes of EPKs: investor preference
- Understand the dynamics of risk patterns



Outline

- 1. Motivation ✓
- 2. Uniform Confidence Band
- 3. Monte-Carlo Study
- 4. Empirical Data Analysis



Risk Neutral Density (RND) Estimation

The RND may be estimated from option prices, Breeden and Litzenberger (1978):

$$q_t(S_T) = \exp(r\tau) \frac{\partial^2 H_t(k,\tau)}{\partial k^2}|_{k=S_T}$$

with call price function $H_t(k, \tau)$.

Aït-Sahalia and Lo (1998) estimate $H_t(k, \tau)$ nonparametrically and differentiate it twice w.r.t. k.

Call prices (X_i, Y_i) , with fixed τ , we have:

$$Y_i = H(x_i) + \varepsilon_i, i = 1, \ldots, n_q$$

 (X_i, Y_i) s are i.i.d.

Define $L\{y; H(u)\}$ as the conditional density of Y given K = u Local polynomial estimate, $(x \approx u)$:

$$H(u) \approx H(x, u) \stackrel{\text{def}}{=} \sum_{j=0}^{3} H_j(x) (u - x)^j$$

Local likelihood

$$L_{n_q}\{H(x)\} \stackrel{\text{def}}{=} \frac{1}{n_q} \sum_{i=1}^{n_q} K_{h_{n_q}}(x - x_i) \log L\{Y_i; H(x_i, x)\},$$



Data

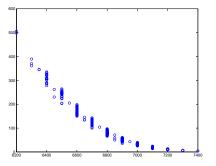


Figure 5: Plot of call prices against strikes k, $n_q = 1000$, $n_p = 500$.



Solutions:

$$\widehat{\mathbf{H}(\mathbf{x})} \stackrel{\text{def}}{=} \operatorname{argmax}_{H} L_{n_q} \{ H(x) \},$$

where

$$\widehat{\mathbf{H}(\mathbf{x})} = \{\widehat{H_0(x)}, \widehat{H_1(x)}, \widehat{H_2(x)}, \widehat{H_3(x)}\}^\top$$

Estimate for $q_t(x)$:

$$\widehat{q_t(x)} \propto 2! \widehat{H_2(x)}$$

The kernel density estimate for $p_t(x)$ is based on historical $\{S_t\}$:

$$\widehat{p_t(x)} = n_p^{-1} \sum_{j=1}^{n_p} K_{h_{n_p}}(x - S_j)$$

Uniform Convergence

Theorem

Under regularity conditions, for all x in an interval J, we have a.s.,

$$\sup_{x \in J} |\widehat{\mathcal{K}_{t,\tau}(x)} - \mathcal{K}_{t,\tau}(x)| = \mathcal{O}[\max\{(n_p h_{n_p}/\log n_p)^{-0.5}, h_{n_p}^2, h_{n_q}^3, h_{n_q}^{-2}\{n_q h_{n_q}/\log n_q\}^{-0.5}\}]$$

Uniform Confidence Band

Theorem

Under regularity conditions,

$$\mathcal{K}_{nt,\tau}(x) \stackrel{\mathrm{def}}{=} n_q^{1/2} h_{n_q}^{5/2} \{\widehat{\mathcal{K}_{t,\tau}(x)} - \mathcal{K}_{t,\tau}(x)\} \operatorname{Var}\{\widehat{\mathcal{K}_{t,\tau}(x)}\}^{-1/2}.$$

We have:

$$P \left\{ (-2\log h_{n_q})^{1/2} \left\{ \sup_{x \in J} |\mathcal{K}_{nt,\tau}(x)| - c_{nt} \right\} < z \right\} \\ \longrightarrow \exp\{-2\exp(-z)\},$$

where
$$c_{nt} = (-2 \log h_{n_a})^{1/2} + (-2 \log h_{n_a})^{-1/2} \{x_{\alpha} + \log(C/2\pi)\}$$

Uniform Confidence for PKs -



Uniform Confidence Band

Thus, a $(1-\alpha)100\%$ confidence band for pricing kernel $\mathcal{K}_{t,\tau}$ is:

$$[f(x): \sup_{x} \{|\widehat{\mathcal{K}_{t,\tau}(x)} - f(x)|\widehat{\mathsf{Var}}(\widehat{\mathcal{K}_{t,\tau}(x)})^{-1/2}\} \leq L_{\alpha}]$$

where

$$L_{\alpha} = 2!(n_q h_{n_q}^5)^{-1/2} c_{nt}$$

and

$$x_{\alpha} = -\log\{-1/2\log(1-\alpha)\}$$

Extension on τ

Let $\mathfrak x$ be the possible set of maturities, the extension of our results over τ :

$$[f_{t,\tau}(x): \sup_{x\in \mathcal{E}, \tau\in\mathfrak{x}}\{|\widehat{\mathcal{K}_{t,\tau}(x)} - f_{t,\tau}(x)|\widehat{\mathsf{Var}}(\widehat{\mathcal{K}_{t,\tau}(x)})^{-1/2}\} \leq L_\alpha].$$

In the BS setup, the evolution of bands over time, for fixed τ_1 $(g(\tau_1 - \tau_2) = \mathcal{K}_{t,\tau_1}(x)/\mathcal{K}_{t,\tau_2}(x)))$

$$\begin{split} &[f_{t,\tau_2}:\widehat{g}(\tau_1-\tau_2)\{-L_{\alpha}\widehat{\mathsf{Var}}(\widehat{\mathcal{K}_{t,\tau_1}(x)})^{1/2}+\widehat{\mathcal{K}_{t,\tau_1}(x)}\}\leq f_{t,\tau_2}(x)\leq \\ &\widehat{g}(\tau_1-\tau_2)\{L_{\alpha}\widehat{\mathsf{Var}}(\widehat{\mathcal{K}_{t,\tau_1}(x)})^{1/2}+\widehat{\mathcal{K}_{t,\tau_1}(x)}\}], \end{split}$$

Bootstrap

Theorem

Under regularity conditions

$$[f_{t,\tau}: \sup_{x \in E} \{|\widehat{\mathcal{K}_{t,\tau}(x)} - f_{t,\tau}(x)| \operatorname{Var}(\widehat{\mathcal{K}_{t,\tau}})^{-1/2}\} \leq L_{\alpha}^*]$$

where the bound L_{α}^{*} satisfies

$$P^*(-\{U_{n_q}(x)^{-1}H_{n_q}^{-1}A_{n_q}^*(x)/B_{t,\tau}(x)^{-1}N(x)^{-1}M^*(x)N(x)^{-1}\}_{3,3} \le L_{\alpha}^*) = 1 - \alpha$$

1 1

A Monte-Carlo Study

For $q_t(x)$, generate data from BS model, interest rate r=0.04, $S_t=6500$, $k\in[6200,7400]$, $\tau=1$ M, $\varepsilon_i\in U[0,6]$, $\sigma=0.1878$.

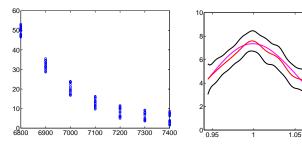


Figure 6: (Left) H against k (Right) Plot of confidence bands (black), estimated value, the Black Scholes SPD (magenta) of the EPK, $h_{n_q} = 0.085$, $\alpha = 0.05$, $n_q = 300$.

Uniform Confidence for PKs

A Monte-Carlo Study

For historical density, simulate data from Geometric Brownian Motion, with $\mu=0.23,~\sigma=0.1878.$

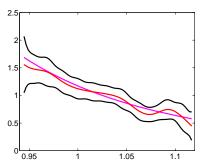


Figure 7: Plot of confidence bands, estimated value, the Black Scholes EPK (magenta), $h_{n_q} = 0.060$, $\alpha = 0.05$, $n_q = 500$, $n_p = 600$.

Coverage Probability

Case	(n =)300	450	600
$(\tau =)3$	0.9063(2.402) 0.8964(2.438)	0.9144(2.204) 0.9056(2.134)	` ,

Table 1: Cov. prob. (area) of the uniform confidence band for $q_t(x)$ at $\alpha=5\%$ with $\sigma=0.1878$, sim = 500

Case	(n =)300	450	600
3	0.7820(2.5434)	0.7980(2.4978)	0.8020(2.4131)
6	0.8602(2.4987)	0.8749(2.4307)	0.8900(2.4131)

Table 2: Same for EPK at $\alpha=10\%$



Empirical Data Analysis

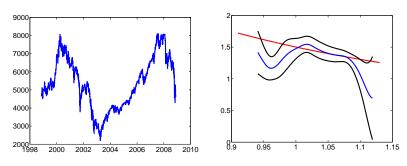


Figure 8: (Left) Plot of DAX index (Right) Plot of confidence bands (black), EPK by Black Scholes fitting, nonparametric EPK, $h_{n_q} = 0.075$, $\alpha = 0.05$, $n_p = 506$, $n_q = 715$.



Empirical Data Analysis

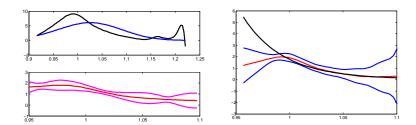


Figure 9: (Left) Plot of q and p (upper panel) Plot EPK and its bands, 2006 Feb 28th (lower panel) (Right) Plot of confidence bands (blue), EPK by Black Scholes fitting (black), EPK, 2006, April, 24th.

Empirical Data Analysis

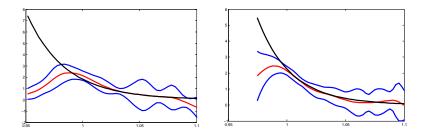


Figure 10: (Left) Plot of confidence bands (blue), EPK by Black Scholes fitting (black), EPK, 2006, July, 24th. (Right) Same for 2006, Aug, 18th.

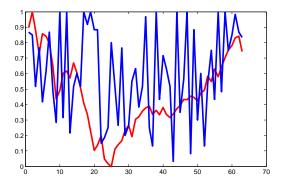


Figure 11: Plot of estimation of the BS EPK covered in band, DAX price (red) $\tau=2\mathrm{M.}(200001\text{-}200006)$

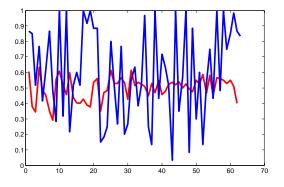


Figure 12: Plot of estimation of the BS EPK covered in band (blue), DAX price difference (red) $\tau=2$ M.(200001-200006)

Conclusions

- Uniform confidence bands tell us about risk patterns
- Smoothing of EPK is best done via IVS
- Bootstrap does not improve coverage probability significantly
- oxdot BS for au=0.5M is mostly rejected
- Bootstrap improvement possible for robust smoothers



Uniform Confidence for Pricing Kernels

Wolfgang Karl Härdle Yarema Okhrin Weining Wang

Ladislaus von Bortkiewicz Chair of Statistics C.A.S.E.-Center for Applied Statistics and Economics Humboldt-Universität zu Berlin University of Augsburg



Bibliography

Aït-Sahalia, Y., and A. W. Lo.

Nonparametric Estimation of State-Price Densities Implicit in

Financial Asset Prices

Journal of Finance, 1998, 53, 499-547

Aït-Sahalia, Y., and A. W. Lo.

Nonparametric Risk Management and Implied Risk Aversion

Journal of Econometrics, 2000, 94, 9-51

Breeden, D. T. and Litzenberger, R. H. Prices of state-contingent claims implicit in option prices Journal of Business, 1978, 51, 621-651



Bibliography — 5-2

Bibliography

Detlefsen, K., Härdle, W., and Moro, R.

Empirical Pricing Kernels and Investor Preference

Mathematical Methods in Economics and Finance, forthcoming

Engle, R. F. and Rosenberg, J. V.
Empirical pricing kernels
Journal of Financial Economics, 2002, 64, 341-372

Claeskens, G. and Van Keilegom, I.

Bootstrap Confidence Bands for Regression Curves and Their

Derivatives

The Annals of Statistics, 2003, 31, 1852-1884



Bibliography — 5-3

Bibliography

Giacomini, E., Handel, M. and Härdle, W. Time Dependent Relative Risk Aversion Risk Assessment: Decisions in Banking and Finance, Bol, G., Rachev, S., Würth, R., Springer, 2008, 15-46

Härdle, W., Krätschmer, V. and Moro, R.

A Microeconomic Explanation of the EPK Paradox

Journal of Financial Econometrics, submitted on 18.09.09

Härdle, W. and Hlavka, Z.

Dynamics of state price densities

Journal of Econometrics, 2009, 150, 1-15



Bibliography — 5-4

Bibliography

Jackwerth, J.C.

Risk Aversion from Option Prices and Realized Returns The Review of Financial Studies, 2000, 13, 433-451

Leland, H. E.

Who should buy portfolio insurance? Journal of Finance, 1980, 32, 581-594

Rookley, C.

Fully exploiting the information content of intra day option quotes: Applications in option pricing and risk management University of Arizona working paper, 1997



Rookley (1997)

Let C_{it} be the price of the i^{th} option at time t and K_{it} its strike price, and define the rescaled call option $c = C/S_t$ in terms of moneyness $M = S_t/K$ s.t.

$$c_{it} = c\{M_{it}; \sigma(M_{it})\} = \Phi(d_1) - \frac{e^{-r\tau}\Phi(d_2)}{M_{it}}$$

$$d_1 = \frac{\log(M_{it}) + \left\{r_t + \frac{1}{2}\sigma(M_{it})^2\right\}\tau}{\sigma(M_{it})\sqrt{\tau}}$$

$$d_2 = d_1 - \sigma(M_{it})\sqrt{\tau}$$

The RND is then

$$q(\cdot) = e^{r\tau} \frac{\partial^2 C}{\partial K^2} = e^{r\tau} S \frac{\partial^2 C}{\partial K^2}$$

with

$$\frac{\partial^2 c}{\partial K^2} = \frac{\mathrm{d}^2 c}{\mathrm{d}M^2} \left(\frac{M}{K}\right)^2 + 2\frac{\mathrm{d}c}{\mathrm{d}M} \frac{M}{K^2}$$

and

$$\frac{d^{2}c}{dM^{2}} = \Phi'(d_{1}) \left\{ \frac{d^{2}d_{1}}{dM^{2}} - d_{1} \left(\frac{dd_{1}}{dM} \right)^{2} \right\}
- \frac{e^{-r\tau}\Phi'(d_{2})}{M} \left\{ \frac{d^{2}d_{2}}{dM^{2}} - \frac{2}{M} \frac{dd_{2}}{dM} - d_{2} \left(\frac{dd_{2}}{dM} \right)^{2} \right\}
- \frac{2e^{-r\tau}\Phi(d_{2})}{M^{3}}$$

With
$$V = \sigma(M)$$
, $V' = \frac{\partial \sigma(M)}{\partial M}$, $V'' = \frac{\partial^2 \sigma(M)}{\partial M^2}$

$$\frac{d^{2}d_{1}}{dM^{2}} = -\frac{1}{M*V(M)\sqrt{\tau}} \left\{ \frac{1}{M} + \frac{V'(M)}{V(M)} \right\}
+V''(M) \left\{ \frac{\sqrt{\tau}}{2} - \frac{\log(M) + r\tau}{V(M)^{2}\sqrt{\tau}} \right\}
+V'(M) \left\{ 2V'(M) \frac{\log(M) + r\tau}{V(M)^{3}\sqrt{\tau}} \right\}
-\frac{1}{M*V(M)^{2}\sqrt{\tau}} \right\}$$

$$\frac{d^{2}d_{2}}{dM^{2}} = -\frac{1}{M*V(M)\sqrt{\tau}} \left\{ \frac{1}{M} + \frac{V'(M)}{V(M)} \right\}
-V''(M) \left\{ \frac{\sqrt{\tau}}{2} + \frac{\log(M) + r\tau}{V(M)^{2}\sqrt{\tau}} \right\}
+V'(M) \left\{ 2V'(M) \frac{\log(M) + r\tau}{V(M)^{3}\sqrt{\tau}} \right\}
-\frac{1}{M*V(M)^{2}\sqrt{\tau}} \right\}$$