

# Fourier Space Time-stepping Framework for Option Pricing with Lévy Models

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Sydney Financial Mathematics Workshop  
September 14, 2009

# Overview

Develop a framework for numerical pricing of financial derivatives that

- Precise, fast and rapidly convergent
- Applicable to pricing of a wide range of European and path-dependent, single- and multi-asset, vanilla and exotic derivatives
- Efficiently handles path-independent and discretely monitored derivatives
- Generically handles various spot-price models and option payoffs

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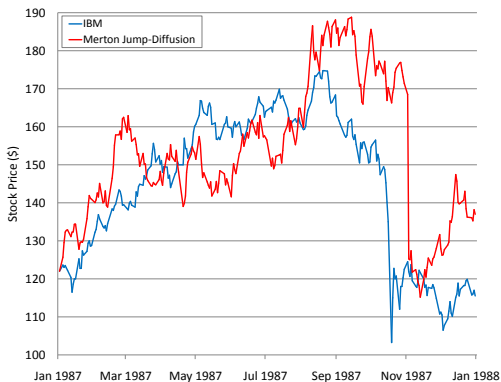
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## The Approach

- Consider the PIDE for the option price
- Transform the PIDE into ODE in Fourier space and solve the ODE analytically
- Utilize FFT to efficiently switch between real and Fourier spaces

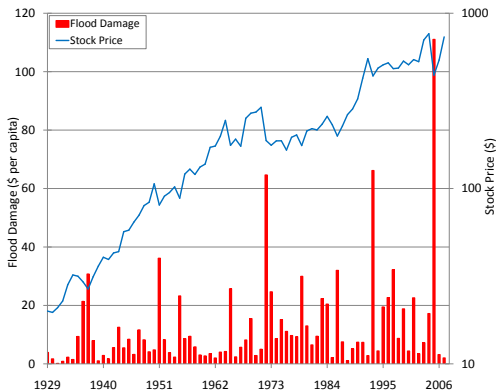
# Equity Derivatives



## Independent-increment exponential Lévy models

- Geometric Brownian motion
- Merton/Kou jump-diffusion
- Exponential Lévy models (VG, NIG, CGMY)

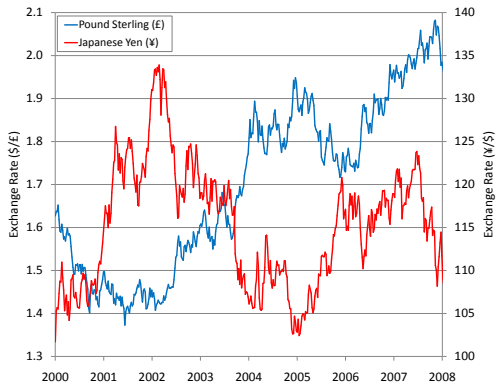
# Insurance Derivatives



Joint evolution of insurance losses and asset prices model

- Losses are random, driven by Poisson process
- Jumps in asset price is driven by loss arrival

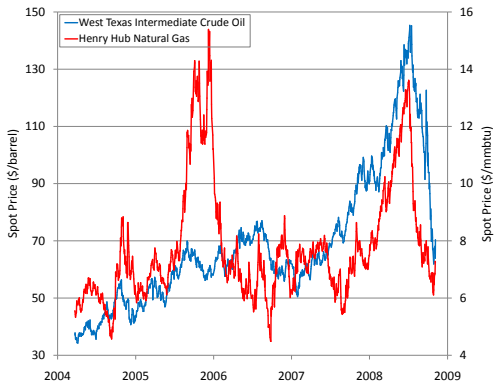
# Currency Derivatives



## Regime-switching exponential Lévy models

- Merton/Kou jump-diffusion with stochastic volatility
- Exponential Lévy models with stochastic volatility and skew

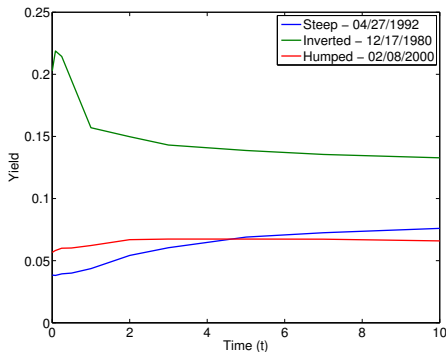
# Commodity Derivatives



## Mean-reverting jump-diffusion models

- Mean-reverting jump-diffusion with stochastic long-run level
- Decoupled mean-reverting diffusions and jumps
- Co-dependent jumps and correlated diffusions

# Interest-rate Derivatives



## Mean-reverting jump-diffusion short-rate models

- Time-inhomogeneous preference-free models:  
Vasicek++ model (a.k.a. extended Vasicek or Hull-White model)
- Two-factor time-inhomogeneous preference-free models:  
Hull-White 2-factor model and the equivalent G2++ model
- Jump extended Vasicek++ and G2++ models

# Numerical Methods for Option Pricing

- Monte Carlo methods
- Tree methods
- Finite difference methods
  - Alternating Direction Implicit-FFT - Andersen and Andreasen (2000)
  - Implicit-Explicit (IMEX) - Cont and Tankov (2004)
  - IMEX Runge-Kutta - Briani, Natalini, and Russo (2004)
  - Fixed Point Iteration - d'Halluin, Forsyth, and Vetzal (2005)
- Quadrature methods
  - Reiner (2001)
  - QUAD - Andricopoulos, Widdicks, Duck, and Newton (2003)
  - Q-FFT - O'Sullivan (2005)
- Transform-based methods
  - Carr and Madan (1999)
  - Raible (2000)
  - Lewis (2001)
  - Lord, Fang, Bervoets, and Oosterlee (2008)

- 1 Introduction
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# Independent Increment Pricing Framework

## The Model

$$\mathbf{S}(t) = \mathbf{S}(0) e^{\mathbf{X}(t)}$$

where  $\mathbf{X}(t)$  is a Lévy process with characteristic triplet  $(\gamma, \Sigma, \nu)$

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- The discount-adjusted and log-transformed price process  $v(t, \mathbf{X}(t)) \triangleq e^{r(T-t)} V(t, \mathbf{S}(0) e^{\mathbf{X}(t)})$  satisfies a PIDE

$$\begin{cases} (\partial_t + \mathcal{L}) v(t, \mathbf{x}) &= \mathbf{0}, \\ v(T, \mathbf{x}) &= \varphi(\mathbf{S}(0) e^{\mathbf{x}}), \end{cases}$$

where  $\mathcal{L}$  is the infinitesimal generator of the Lévy process:

$$\mathcal{L}g(\mathbf{x}) = (\gamma' \partial_{\mathbf{x}} + \frac{1}{2} \partial_{\mathbf{x}}' \Sigma \partial_{\mathbf{x}}) g(\mathbf{x}) + \int (g(\mathbf{x} + \mathbf{y}) - g(\mathbf{x}) - \mathbb{1}_{\{|\mathbf{y}| < 1\}} \mathbf{y}' \partial_{\mathbf{x}} g(\mathbf{x})) \nu(d\mathbf{y})$$

# Fourier Transform

- A function in the space domain  $g(\mathbf{x})$  can be transformed to a function in the frequency domain  $\hat{g}(\boldsymbol{\omega})$ , where  $\boldsymbol{\omega}$  is given in radians per second, and vice-versa using the continuous Fourier transform

$$\mathcal{F}[g](\boldsymbol{\omega}) \triangleq \int_{-\infty}^{\infty} g(\mathbf{x}) e^{-i\boldsymbol{\omega}'\mathbf{x}} d\mathbf{x}$$

$$\mathcal{F}^{-1}[\hat{g}](\mathbf{x}) \triangleq \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{g}(\boldsymbol{\omega}) e^{i\boldsymbol{\omega}'\mathbf{x}} d\boldsymbol{\omega}$$

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- Continuous Fourier transform is a linear operator that maps spatial derivatives  $\partial_x$  into multiplications in the frequency domain

$$\mathcal{F}[\partial_x^n g](\boldsymbol{\omega}) = i\boldsymbol{\omega} \mathcal{F}[\partial_x^{n-1} g](\boldsymbol{\omega}) = \dots = (i\boldsymbol{\omega})^n \mathcal{F}[g](\boldsymbol{\omega})$$

# Pricing Framework in Fourier Space

- Applying the Fourier transform to the infinitesimal generator  $\mathcal{L}$  of  $\mathbf{X}(t)$  allows the characteristic exponent  $\Psi(\boldsymbol{\omega})$  to be factored out:

$$\mathcal{F}[\mathcal{L}v](t, \boldsymbol{\omega}) = \Psi(\boldsymbol{\omega})\mathcal{F}[v](t, \boldsymbol{\omega}),$$

where

$$\Psi(\boldsymbol{\omega}) = i\boldsymbol{\gamma}'\boldsymbol{\omega} - \frac{1}{2}\boldsymbol{\omega}'\boldsymbol{\Sigma}\boldsymbol{\omega} + \int \left( e^{i\boldsymbol{\omega}'\mathbf{y}} - 1 - i\mathbb{1}_{\{|\mathbf{y}|<1\}}\boldsymbol{\omega}'\mathbf{y} \right) \nu(d\mathbf{y})$$

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- The PIDE is therefore transformed into a  $d$ -parameter family of ODEs parameterized by  $\boldsymbol{\omega}$ :

$$\begin{cases} \partial_t \mathcal{F}[v](t, \boldsymbol{\omega}) + \Psi(\boldsymbol{\omega})\mathcal{F}[v](t, \boldsymbol{\omega}) & = 0, \\ \mathcal{F}[v](T, \boldsymbol{\omega}) & = \mathcal{F}[\varphi](\boldsymbol{\omega}) \end{cases}$$

# Characteristic Exponents for Lévy Processes

Model	Characteristic Exponent $\Psi(\omega)$
Black-Scholes-Merton	$i\gamma\omega - \frac{\sigma^2\omega^2}{2}$
Merton jump-diffusion	$i\gamma\omega - \frac{\sigma^2\omega^2}{2} + \lambda(e^{i\tilde{\mu}\omega - \tilde{\sigma}^2\omega^2/2} - 1)$
Kou jump-diffusion	$i\gamma\omega - \frac{\sigma^2\omega^2}{2} + \lambda\left(\frac{\eta_p}{1-i\omega\eta_+} + \frac{1-\eta_p}{1+i\omega\eta_-} - 1\right)$
Variance Gamma	$-\frac{1}{\mu} \ln\left(1 - i\gamma\mu\omega + \frac{\sigma^2\mu\omega^2}{2}\right)$
Normal Inverse Gaussian	$-\frac{1}{\mu} (\sqrt{1 - 2i\gamma\mu\omega + \sigma^2\mu\omega^2} - 1)$
Carr-Geman-Madan-Yor	$C\Gamma(-Y) [(M-i\omega)^Y - M^Y + (G+i\omega)^Y - G^Y]$

# PIDE Solution in Fourier Space

- Given the value of  $\mathcal{F}[v](t, \omega)$  at time  $t_2 \leq T$ , the system is easily solved to find the value at time  $t_1 < t_2$ :

$$\mathcal{F}[v](t_1, \omega) = \mathcal{F}[v](t_2, \omega) \cdot e^{\Psi(\omega)(t_2 - t_1)}$$

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- In discrete space, a step backwards is computed via

## FST Method

$$\mathbf{v}_{m-1} = \text{FFT}^{-1} \left[ \text{FFT}[\mathbf{v}_m] \cdot e^{\Psi(\cdot)\Delta t_m} \right]$$

# Fourier Space Time-stepping

- European options

$$\mathbf{v}_0 = \text{FFT}^{-1} \left[ \text{FFT} [\mathbf{v}_1] \cdot e^{\Psi(\cdot)(T-t)} \right]$$

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$$\mathbf{v}_{m-1}^* = \text{FFT}^{-1} \left[ \text{FFT} [\mathbf{v}_m] \cdot e^{\Psi(\cdot)\Delta t_m} \right],$$

$$\mathbf{v}_{m-1} = \max \{ \mathbf{v}_{m-1}^*, \mathbf{v}_M \},$$

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- Barrier options

$$\mathbf{v}_{m-1} = \text{FFT}^{-1} \left[ \text{FFT} [\mathbf{v}_m] \cdot e^{\Psi(\cdot)\Delta t_m} \right] \cdot \mathbb{1}_{\{x < \mathbf{B}\}} + R \cdot \mathbb{1}_{\{x \geq \mathbf{B}\}}$$

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- Exotic options such as swing, shout, etc.

# American Put Option Results with Penalty Method

N	M	Value	Change	$\log_2(\text{Ratio})$	Time (s)
2048	128	9.22478538			0.027
4096	256	9.22523484	0.0004495		0.109
8192	512	9.22538196	0.0001471	1.6114	0.451
16384	1024	9.22542478	0.0000428	1.7808	1.869
32768	2048	9.22543516	0.0000104	2.0444	8.195

- *Option*: American put option  $S = 90.0$ ,  $K = 98.0$ ,  $T = 0.25$
- *Model*: CGMY model  $C = 0.42$ ,  $G = 4.37$ ,  $M = 191.2$ ,  $Y = 1.0102$ ,  $r = 0.06$
- *Convergence*: 2 in space and 2 in time
- *Reference price*: 9.2254803 from Forsyth, Wan and Wang 2007

# Computation of Greeks

- Delta –  $\partial_{S_k} v(t, \mathbf{x})$

- Differentiation in real space computed via scaling in Fourier space

$$\begin{aligned}\partial_{S_k} v(t, \mathbf{x}) &= \partial_{x_k} v(t, \mathbf{x}) / (\mathbf{S}_k(0) e^{x_k}) \\ &= \mathcal{F}^{-1} [i\omega_k \cdot \mathcal{F}[v](t, \omega)](\mathbf{x}) / (\mathbf{S}_k(0) e^{x_k})\end{aligned}$$

- The discrete method for computing Deltas is then given by

$$\Delta_{k,m-1} = \text{FFT}^{-1} [i\omega_k \cdot \hat{\mathbf{v}}_{m-1}] / (\mathbf{S}_k(0) e^{x_k}) ,$$

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- Gamma –  $\partial_{S_k^2} v(t, \mathbf{x})$

- Differentiation in real space computed via scaling in Fourier space

$$\begin{aligned}\partial_{S_k^2} v(t, \mathbf{x}) &= \left(-\partial_{x_k} + \partial_{x_k^2}\right) v(t, \mathbf{x}) / (\mathbf{S}_k(0) e^{x_k})^2 \\ &= \mathcal{F}^{-1} \left[(-i\omega_k - \omega_k^2) \cdot \mathcal{F}[v](t, \omega)\right](\mathbf{x}) / (\mathbf{S}_k(0) e^{x_k})^2\end{aligned}$$

- The discrete method for computing Gammas is then given by

$$\Gamma_{k,m-1} = \text{FFT}^{-1} \left[-(i\omega_k + \omega_k^2) \cdot \hat{\mathbf{v}}_{m-1}\right] / (\mathbf{S}_k(0) e^{x_k})^2$$

# Computation of Greeks (ctd.)

- Vega –  $\partial_{\sigma_k} v(t, \mathbf{x})$

- Vega satisfies a PIDE

$$\partial_{\sigma_k} \{(\partial_t + \mathcal{L}) v(t, \mathbf{x})\} = (\partial_t + \mathcal{L}) \partial_{\sigma_k} v(t, \mathbf{x}) + \mathcal{H}_{\sigma_k} v(t, \mathbf{x}) = \mathbf{0},$$

where  $\mathcal{H}_{\sigma_k} = (\partial_{\sigma_k} \gamma)' \partial_{\mathbf{x}} + \partial_{\mathbf{x}}' (\partial_{\sigma_k} \boldsymbol{\Sigma}) \partial_{\mathbf{x}}$ .

- The PIDE can be solved explicitly in Fourier space

$$\partial_{\sigma_k} v(t, \mathbf{x}) = (T - t) \mathcal{F}^{-1} [\mathcal{F} [\mathcal{H}_{\sigma_k}](\boldsymbol{\omega}) \cdot \mathcal{F} [v](t, \boldsymbol{\omega})](\mathbf{x}).$$

- The greekFST method for computing Vegas is then given by

$$\nabla_{k,m-1} = \Delta t_m \text{FFT}^{-1} [\mathcal{F} [\mathcal{H}_{\sigma_k}](\boldsymbol{\omega}) \cdot \hat{\mathbf{v}}_{m-1}]$$

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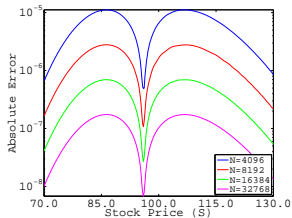
- Rho –  $\partial_r v(t, \mathbf{x})$

- Rho satisfies a similar PIDE and the discrete method for computing Rho is

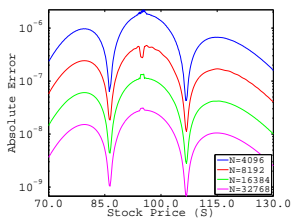
$$\mathbf{P}_{m-1} = \Delta t_m \text{FFT}^{-1} [\mathcal{F} [\mathcal{H}_r](\boldsymbol{\omega}) \cdot \hat{\mathbf{v}}_{m-1}]$$

# Greeks Errors

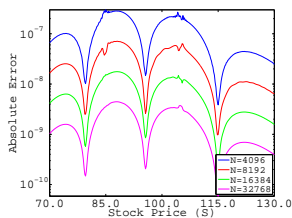
Price



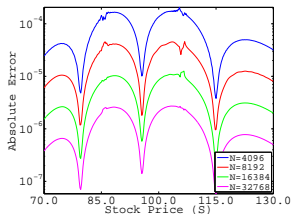
Delta



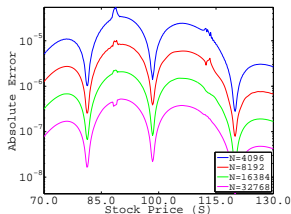
Gamma



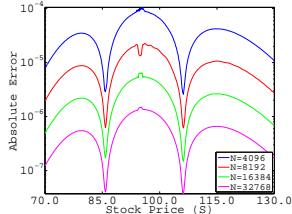
Vega



Theta



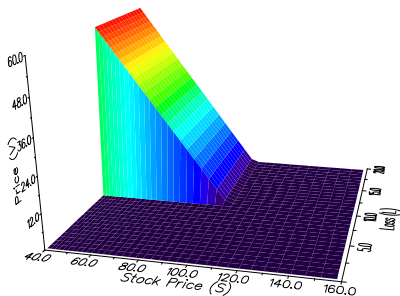
Rho



- Digital call option  $K = 100, T = 0.5$
- Merton jump-diffusion model  $\sigma = 0.15, r = 0.05, q = 0.02, \lambda = 0.1,$   
 $\tilde{\mu} = -1.08, \tilde{\sigma} = 0.4$

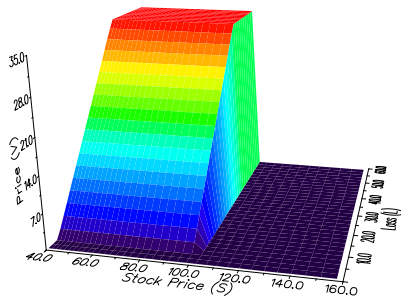
# Catastrophe Options

- Catastrophe options pay the holder a function of total losses and the company's equity value



Catastrophe equity put

$$\varphi(S, L) = \mathbb{1}_{\{L > L^*\}} (K - S)_+$$



Double-trigger stop-loss

$$\varphi(S, L) = \mathbb{1}_{\{S < K\}} [(L - L_a)_+ - (L - L_d)_+]$$

# Catastrophe Options (ctd.)

- Joint model for losses and share price

$$S(t) = S(0) e^{\gamma t - \chi L(t) + \sigma W(t)},$$

$$L(t) = \sum_{n=1}^{N(t)} I_n,$$

where  $\chi$  represents the percentage drop in the share price per unit of loss

- The 2-dimensional Lévy density is  $\nu(dy_1 \times dy_2) = f_L(y_2) \delta(y_1 + \chi y_2) dy_1 dy_2$  resulting in the characteristic function

$$\Psi(\omega_1, \omega_2) = i\gamma\omega_1 - \frac{1}{2}\sigma^2\omega_1^2 + \lambda \left[ \left[ 1 - \frac{i\nu_l}{m_l}(-\chi\omega_1 + \omega_2) \right]^{-\frac{m_l^2}{\nu_l}} - 1 \right]$$

# FST Method Summary

- Various jump-diffusion and exponential Lévy models are handled generically
- Option values are obtained for a range of spot prices
- Can readily price path-dependent and multi-asset options
- Two FFTs per time-step are required
- No time-stepping for European options or between monitoring dates of discretely monitored options
- Second order convergence in space and second order convergence in time for American options with penalty method
- Extendable to computation of the Greeks

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# Regime Switching Pricing Framework

- Let  $\mathbb{K} \triangleq \{1, \dots, K\}$  denote the possible states of the world
- Let  $Z(t) \in \mathbb{K}$  denote the prevailing state of the world at time  $t$ , driven by a continuous time Markov chain with generator  $\mathbf{A}$ . The transition probability from state  $k$  at time  $t_1$  to state  $l$  at time  $t_2$  is  $P_{kl}^{t_1 t_2} = \exp\{(t_2 - t_1)\mathbf{A}\}_{kl}$
- Given that  $Z(t) = k$ , the joint stock price process  $\mathbf{S}(t)$  is assumed to be following a  $d$ -dimensional exponential Lévy model

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- Given that  $Z(t) = k$ , the joint stock price process  $\mathbf{S}(t)$  is assumed to be following a  $d$ -dimensional exponential Lévy model
- The discounted-adjusted and log-transformed option prices at time  $t$  conditional on the state  $Z(t) = k$ , denoted by  $v^{(k)}(t, \mathbf{x})$ , satisfy the following system of PIDEs:

$$\begin{cases} (\partial_t + A_{kk} + \mathcal{L}^{(k)}) v^{(k)}(t, \mathbf{x}) + \sum_{l \neq k} A_{kl} v^{(l)}(t, \mathbf{x}) & = \mathbf{0}, \\ v^{(k)}(t, \mathbf{x}) & = \varphi(\mathbf{S}(0)e^{\mathbf{x}}) \end{cases}$$

# Regime Switching Pricing Framework in Fourier Space

- The transformed PIDE can be written in a compact matrix form

$$\begin{cases} (\partial_t + \Psi(\omega)) \mathcal{F}[\vec{v}](t, \omega) & = 0, \\ \mathcal{F}[\vec{v}](T, \omega) & = \mathcal{F}[\varphi](\omega) \vec{1}, \end{cases}$$

where  $\vec{v}$  is the collection of  $v^{(k)}$ 's stacked into a column vector and the elements of the “matrix characteristic function”  $\Psi$  are

$$\langle \Psi(\omega) \rangle_{kl} \triangleq \begin{cases} A_{kk} + \Psi^{(k)}(\omega), & k = l, \\ A_{kl}, & k \neq l \end{cases}$$

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## rsFST Method

$$\vec{v}_{m-1} = \text{FFT}^{-1} \left[ e^{\Psi(\Delta t_m)} \cdot \text{FFT}[\vec{v}_m] \right]$$

# Hidden vs. Visible Regimes for Bermudan Options

- The conditional holding price  $\vec{\mathbf{v}}_{m-1}^*$  at time step  $m - 1$  is

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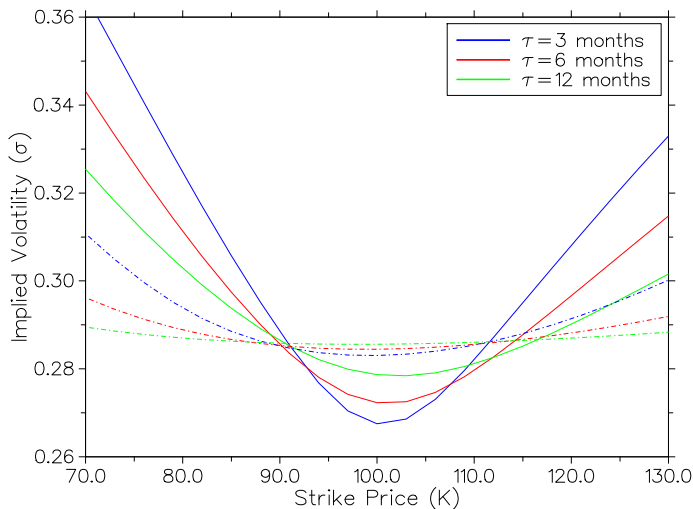
- If the states are hidden, the exercise boundary is the same for every state.

$$\mathbf{v}_{m-1} = \max \left\{ \mathbf{v}_{m-1}^*, \mathbf{v}_M \right\}$$

where  $\mathbf{v}_{m-1}^*$  is the (single) unconditional holding price

$$\mathbf{v}_{m-1}^* = \exp\{(m-1)\Delta t \mathbf{A}\} \vec{\rho} \cdot \vec{\mathbf{v}}_{m-1}^*$$

# Implied Volatility Surface



Merton-jump diffusion model with constant and regime-switching volatility

# rsFST Method Summary

- Can incorporate regime-switching on any parameter of the model
- Option values are obtained for a range of spot prices
- No time-stepping for European options or between monitoring dates of discretely monitored options
- Computational complexity scales linearly with the number of states

- 1 Introduction
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- 4 Mean-Reverting FST method**
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# Mean-Reverting Pricing Framework

## The Model

$$d\mathbf{Y}(t) = -\kappa\mathbf{Y}(t-)dt + d\mathbf{J}(t),$$

$$\mathbf{X}(t) = \boldsymbol{\theta} + \boldsymbol{\Lambda}\mathbf{Y}(t),$$

$$\mathbf{S}(t) = \mathbf{S}(0)e^{\mathbf{X}(t)},$$

where  $\mathbf{J}(t)$  is a Lévy process with characteristic triplet  $(\gamma, \boldsymbol{\Sigma}, \nu)$

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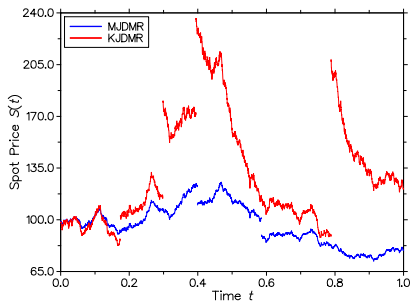
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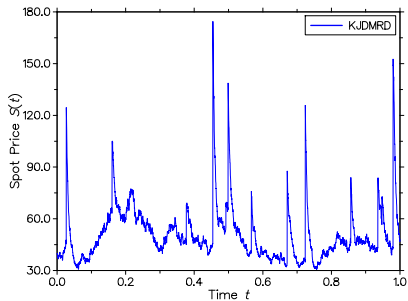
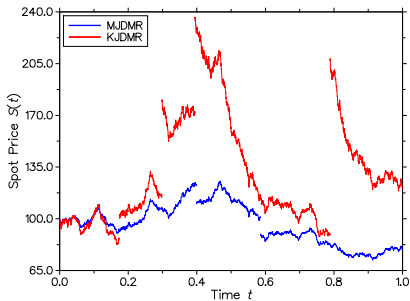
Generalizes a number of well-known models

- Gibson and Schwartz (1990)
- Clewlow and Strickland (2000)
- Barlow, Gusev and Lai (2004)
- Hikspoors and Jaimungal (2007)

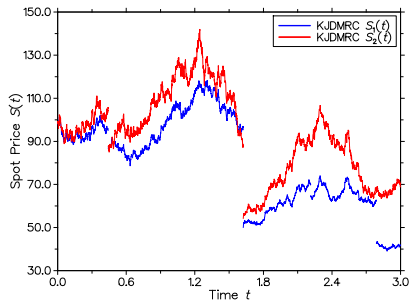
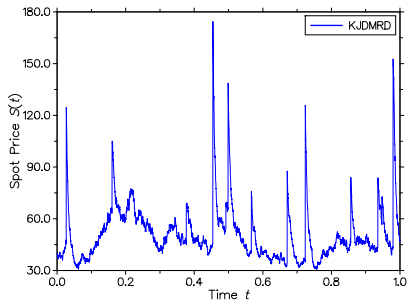
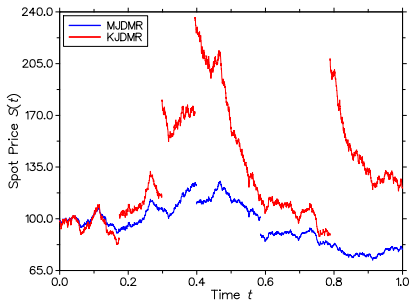
# Examples of obtainable spot-price dynamics



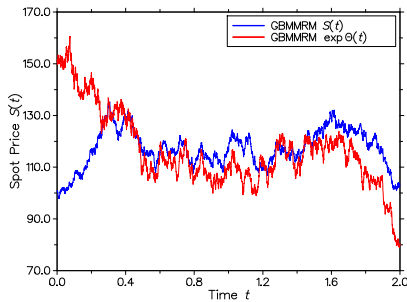
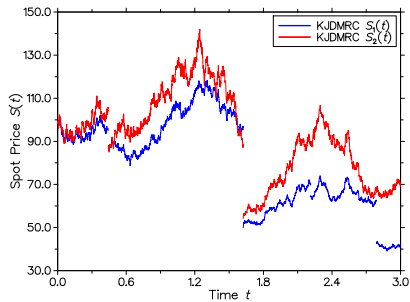
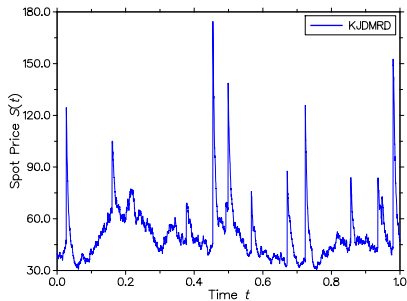
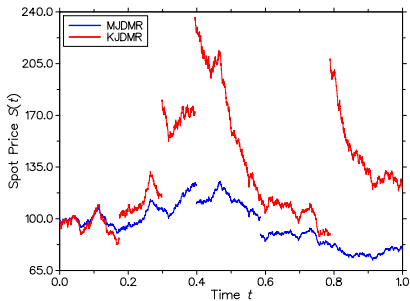
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# PIDE Transformation into Fourier Space

- The option price satisfies a PIDE

$$\begin{cases} (\partial_t + \mathcal{L})v(t, \mathbf{y}) &= 0, \\ v(T, \mathbf{y}) &= \varphi(\mathbf{S}(0)e^{\boldsymbol{\theta} + \boldsymbol{\Lambda} \mathbf{y}}), \end{cases}$$

where  $\mathcal{L}$  is the infinitesimal generator of the  $\mathbf{Y}(t)$  process:

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- The PIDE is transformed into a  $d$ -parameter family of ODEs parameterized by  $\boldsymbol{\omega}$ :

$$\begin{cases} (\partial_t + \psi(e^{-\boldsymbol{\kappa}'(T-t)}\boldsymbol{\omega}) + \text{Tr } \boldsymbol{\kappa}) \mathcal{F}[\tilde{v}](t, \boldsymbol{\omega}) & = 0, \\ \mathcal{F}[\tilde{v}](T, \boldsymbol{\omega}) & = \mathcal{F}[\tilde{\varphi}](\boldsymbol{\omega}), \end{cases}$$

where

$$\mathcal{F}[\tilde{v}](t, \boldsymbol{\omega}) = \mathcal{F}[v](t, e^{-\boldsymbol{\kappa}'(T-t)}\boldsymbol{\omega})$$

# Fourier Space Solution

- The ODE can be solved in Fourier space

$$\mathcal{F}[v](t_1, \omega) = \mathcal{F}[v](t_2, e^{\kappa'(t_2-t_1)} \omega) \cdot e^{\Psi(t_2-t_1, \omega) + (t_2-t_1)\text{Tr } \kappa}$$

where

$$\Psi(t_2 - t_1, \omega) = \int_0^{t_2-t_1} \psi(e^{\kappa' u} \omega) du, \quad \text{and}$$

$$\psi(\omega) = -\frac{1}{2} \omega' \Sigma \omega + \int (e^{i\omega' y} - 1) \nu(dy)$$

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- In discrete space, a step backwards is computed via

## mrFST Method

$$\mathbf{v}_{m-1} = \text{FFT}^{-1} \left[ \text{FFT} [\check{\mathbf{v}}_m] \cdot e^{\Psi(\Delta t_m, \cdot)} \right], \quad \text{where } \check{v}(\mathbf{y}) \triangleq v(\mathbf{y} e^{-\kappa' \Delta t})$$

# Discrete Barrier Option Results

N	M	Value	Change	$\log_2$ Ratio	Time (sec.)
2048	126	0.58232029			0.033
4096	126	0.58303190	0.0007116		0.067
8192	126	0.58244527	0.0005866	0.2786	0.156
16384	126	0.58232817	0.0001171	2.3248	0.318
32768	126	0.58231749	0.0000107	3.4541	0.674

- *Option*: Up-and-out barrier call  $S = 100, K = 105, T = 0.5,$   
 $B = 115, R = 0.5$  with daily monitoring
- *Model*: Kou jump-diffusion with mean reversion  
 $\sigma = 0.3, \lambda = 4.0, \eta_p = 0.95, \eta_+ = 0.3, \eta_- = 0.1, \theta = 92.0, \kappa = 5.0, r = 0.06$
- *Monte Carlo*: 0.58289924 – 95% CI width of 0.0028937 @ 56 sec.

# mrFST Method Summary

- Readily extends from the FST method using interpolation of option prices grid
- Retains all of the computational efficiency of the FST method
- For high mean-reversion rates, several time-steps may be required

- 1 Introduction
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Pricing Framework under Risk-adjusted Measure  $\mathbb{Q}$ 

## The Model

$$r(t) = X_1(t) + \dots + X_d(t) + \delta(t)$$

where  $\mathbf{X}(t) = [X_1(t), \dots, X_d(t)]'$  satisfy

$$d\mathbf{X}(t) = -\kappa\mathbf{X}(t)dt + d\mathbf{J}(t), \quad \mathbf{X}(0) = \mathbf{0}$$

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The option price satisfies a PIDE

$$\begin{cases} (\partial_t + \mathcal{L}) V_Q(t, \mathbf{x}) = 0, \\ V_Q(T, \mathbf{x}) = \varphi(\mathbf{x}), \end{cases}$$

where  $\mathcal{L}$  is

$$\mathcal{L}f(\mathbf{x}) = (-\kappa\mathbf{x}'\partial_{\mathbf{x}} + \frac{1}{2}\partial_{\mathbf{x}}'\Sigma\partial_{\mathbf{x}})f(\mathbf{x}) + \int [f(\mathbf{x} + \mathbf{z}) - f(\mathbf{x})] \nu(d\mathbf{z}) - (\Sigma\mathbf{x}_i + \delta(t))$$

Pricing Framework under T-forward Measure  $\mathbb{T}$ 

## The Model

$$r(t) = X_1(t) + \dots + X_d(t) + \delta(t)$$

where  $\mathbf{X}(t) = [X_1(t), \dots, X_d(t)]'$  satisfy

$$d\mathbf{X}(t) = -(\kappa\mathbf{X}(t) + \Sigma\mathbf{B}(t))dt + d\mathbf{J}(t), \quad \mathbf{X}(0) = \mathbf{0}$$

and  $\mathbf{B}(t) = [B_1(t), \dots, B_d(t)]'$ ,  $B_i(t) = (1 - e^{-\kappa_i(T-t)})/\kappa_i$

Pricing Framework under T-forward Measure  $\mathbb{T}$ 

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$$\mathcal{L}f(\mathbf{x}) = (-\kappa\mathbf{x}'\partial_{\mathbf{x}} - \partial_{\mathbf{x}}'\Sigma\mathbf{B}(t) + \frac{1}{2}\partial_{\mathbf{x}}'\Sigma\partial_{\mathbf{x}})f(\mathbf{x}) + \int [f(\mathbf{x} + \mathbf{z}) - f(\mathbf{x})] e^{\mathbf{B}(t)\mathbf{z}} \nu(d\mathbf{z})$$

# Pricing Framework under T-forward Measure $\mathbb{T}$

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Now we can apply the results from mrFST method!

# Fourier Space Solution

- The solution to the PIDE in Fourier space is

$$V_T(t_1, \mathbf{x}) = \mathcal{F}^{-1} \left[ \mathcal{F} [V_T](t_2, e^{\kappa'(t_2-t_1)} \boldsymbol{\omega}) \cdot e^{\Psi(t_2-t_1, \boldsymbol{\omega}) + (t_2-t_1) \text{Tr} \boldsymbol{\kappa}} \right] (\mathbf{x})$$

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$$\psi(t, \boldsymbol{\omega}) = -i\boldsymbol{\omega} \boldsymbol{\Sigma} \mathbf{B}(t) - \frac{1}{2} \boldsymbol{\omega}' \boldsymbol{\Sigma} \boldsymbol{\omega} + \int \left( e^{i\boldsymbol{\omega}' \mathbf{z}} - 1 \right) e^{\mathbf{B}(t) \mathbf{z}} \nu(d\mathbf{z}),$$

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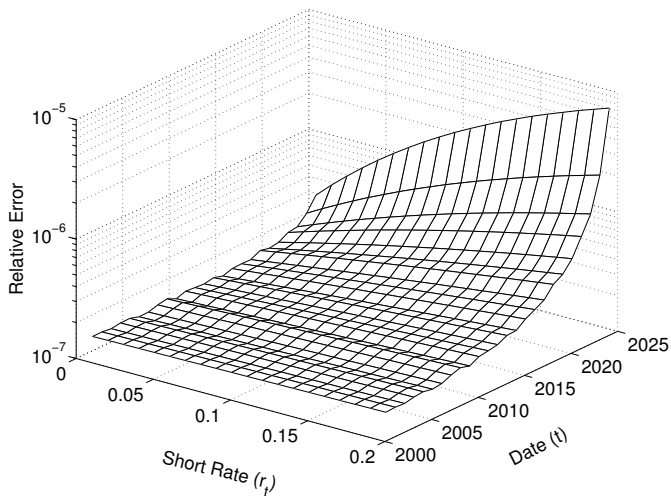
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- In discrete space, a step backwards is computed via

## irFST Method

$$\mathbf{V}_{T,m-1} = \text{FFT}^{-1} \left[ \text{FFT} \left[ \check{\mathbf{V}}_{T,m} \right] \cdot e^{\Psi(t_{m-1}, t_m, \cdot)} \right], \text{ where } \check{v}_T(\mathbf{y}) \triangleq v_T(\mathbf{y} e^{-\kappa' \Delta t})$$

# Numerical Results



Relative error for pricing a 5-year swap under the Hull-White model

# irFST Method Summary

- Currently, handles the following models
  - 1-factor: Vasicek, Hull-White, and Vasicek-J++
  - 2-factor: Hull-White 2F, G2++, and G2-J++
- Readily extends from the FST method using interpolation of option prices grid
- Retains all of the computational efficiency of the FST method
- For high mean-reversion rates, several time-steps may be required

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# FST Framework Summary

## The Approach

- Consider the PIDE for the option price
- Transform the PIDE into ODE in Fourier space and solve the ODE analytically
- Utilize FFT to efficiently switch between real and Fourier spaces

# FST Framework Summary

## The Approach

- Consider the PIDE for the option price
  - Transform the PIDE into ODE in Fourier space and solve the ODE analytically
  - Utilize FFT to efficiently switch between real and Fourier spaces
- 
- Various jump-diffusion and exponential Lévy models are handled generically
  - Option values are obtained for a range of spot prices
  - Can readily price path-dependent and multi-asset options
  - Two FFTs per time-step are required
  - No time-stepping for European options or between monitoring dates of discretely monitored options
  - Second order convergence in space and second order convergence in time for American options with penalty method



## Option:

American

Option Type  Call  PutSpot Price (S) Strike Price (K) Time to Maturity (T) 

Payoff:

 $\max(K - S_T, 0)$ 

## Stock Price Process:

Kou Jump-Diffusion Mean-Reverting

Risk Free Rate (r) Dividend Yield (q) Volatility ( $\sigma$ ) Jump Arrival Rate ( $\lambda$ ) Jump Direction Prob (p) Jump Mean Up ( $\eta^+$ ) Jump Mean Down ( $\eta^-$ ) Reversion Level ( $\theta$ ) Reversion Speed ( $\kappa$ ) 

Model:

$$S_t = \exp(X_t)$$

$$dX_t = \kappa(\log(\theta) - X_t)dt + \sigma dW_t + JdN_t$$

$$\log(J) \sim \text{Double Exponential}(p, \eta^+, \eta^-)$$

$$dN_t \sim \text{Poisson}(\lambda)$$

## FST Parameters:

Space Points (N): Time Points (M): Plot Format: Plot Colour:  Grayscale

Price

MATLAB code is also available

# Thank You!



Jackson, K. R., S. Jaimungal, and V. Surkov (2007).

Option pricing with regime switching Lévy processes using Fourier space time-stepping.

*In Proceeding of the Fourth IASTED International Conference on Financial Engineering and Applications*, pp. 92–97.



Jackson, K. R., S. Jaimungal, and V. Surkov (2008).

Fourier space time-stepping for option pricing with Lévy models.

*Journal of Computational Finance* 12(2), 1–28.



Jaimungal, S. and V. Surkov (2008).

A general Lévy-based framework for energy price modeling and derivative valuation via FFT.

Working paper, available at <http://ssrn.com/abstract=1302887>.

More at <http://ssrn.com/author=879101>