Using pseudo-parabolic equations for option pricing in jump diffusion models

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#### Outline

In mathematical finance a popular approach for pricing options under some Lévy model is to consider underlying that follows a Poisson jump diffusion process. As it is well known this results in a partial integro-differential equation (PIDE) that usually does not allow an analytical solution while numerical solution brings some problems. In this work we elaborate a new approach on how to transform the PIDE to some class of so-called pseudo-parabolic equations which are known in mathematics but are relatively new for mathematical finance. As an example we discuss several jump-diffusion models which Lévy measure allows such a transformation.

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#### Existing approaches - cont.

- Because of the integrals in the equations the methods have proven to be relatively expensive. Quadrature methods are expensive since the integrals must be evaluated at every point of the mesh. Though less so, Fourier methods are also computationally intensive since in order to avoid wrap around effects they require enlargement of the computational domain. They are also slow to converge when the parameters of the jump process are not smooth, and for efficiency require uniform meshes. And low order of accuracy.
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- Our approach: 1. represent a Lévy measure as the Green's function of some yet unknown differential operator *A*. If we manage to find an explicit form of this operator then the original PIDE reduces to a new type of equation - so-called pseudo-parabolic equation. Done for GTSP with finite activity when *α* is integer.

A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 4 / 13

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A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models

4 / 13 April 28, 2009

#### Alternative methods for real $\alpha$

- Suppose we consider GTSP/KoBoL/SSM class of models. We will transform the corresponding PIDE to a fractional PDE. Fractional PDEs for Lévy processes with finite variation were obtained by Boyarchenko and Levendorsky (2002) and later by Cartea (2007) using a characteristic function technique.
  - We derive it in all cases including processes with infinite variation using a different technique - shift operators.

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 and a pure jump PIDE (could be always obtained by using splitting) reads (positive jumps, but negative - by analogy)

$$\frac{\partial}{\partial \tau} C(x,\tau) = \mathcal{B}_1 C(x,\tau)$$
$$\mathcal{B}_1 \equiv \int_0^\infty \left[ \exp\left(y\frac{\partial}{\partial x}\right) - 1 - (e^y - 1)\frac{\partial}{\partial x} \right] \lambda_R \frac{e^{-\nu_R|y|}}{|y|^{1+\alpha_R}} dy \tag{2}$$

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#### Our results

#### **Positive jumps**

## Main theorem

#### Theorem (1)

The PIDE

$$\frac{\partial}{\partial \tau}C(x,\tau) = \int_0^\infty \left[C(x+y,\tau) - C(x,\tau) - \frac{\partial}{\partial x}C(x,\tau)(e^y - 1)\right] \lambda_R \frac{e^{-\nu_R|y|}}{|y|^{1+\alpha_R}} dy \tag{3}$$

is equivalent to the fractional PDE

$$\frac{\partial}{\partial \tau} C(x,\tau) = \lambda_R \Gamma(-\alpha_R) \left\{ \left( \nu_R - \frac{\partial}{\partial x} \right)^{\alpha_R} - \nu_R^{\alpha_R} + \left[ \nu_R^{\alpha_R} - (\nu_R - 1)^{\alpha_R} \right] \frac{\partial}{\partial x} \right\} C(x,\tau),$$
  

$$\mathbb{R}(\alpha_R) < 2, \ \mathbb{R}(\nu_R - \partial/\partial x) > 0, \ \mathbb{R}(\nu_R) > 1.$$
(4)

In special cases this equation changes to

$$\frac{\partial}{\partial \tau} C(\mathbf{x}, \tau) = \lambda_R \left\{ \log(\nu_R) - \log\left(\nu_R - \frac{\partial}{\partial x}\right) + \log\left(\frac{\nu_R - 1}{\nu_R}\right) \frac{\partial}{\partial x} \right\} C(\mathbf{x}, \tau)$$

$$\alpha_R = 0, \mathbb{R}(\nu_R - \partial/\partial \mathbf{x}) > 0, \mathbb{R}(\nu_R) > 1,$$
(5)

and

$$\begin{split} \frac{\partial}{\partial \tau} C(x,\tau) &= \lambda_R \Big\{ -\nu_R \log \nu_R + (\nu_R - \frac{\partial}{\partial x}) \log \left( \nu_R - \frac{\partial}{\partial x} \right) + \left[ \nu_R \log \nu_R - (\nu_R - 1) \log(\nu_R - 1) \right] \frac{\partial}{\partial x} \Big\} C(x,\tau) \\ \alpha_R &= 1, \mathbb{R}(\partial/\partial x) < 0, \mathbb{R}(\nu_R) > 1, \end{split}$$

A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 6 / 13

### Main theorem - cont.

#### Theorem (2)

The PIDE

$$\frac{\partial}{\partial \tau}C(x,\tau) = \int_{-\infty}^{0} \left[ C(x+y,\tau) - C(x,\tau) - \frac{\partial}{\partial x}C(x,\tau)(e^{y}-1) \right] \lambda_{L} \frac{e^{-\nu_{L}|y|}}{|y|^{1+\alpha_{L}}} dy$$
(6)

is equivalent to the fractional PDE

$$\frac{\partial}{\partial \tau} C(x,\tau) = \lambda_L \Gamma(-\alpha_L) \left\{ \left( \nu_L + \frac{\partial}{\partial x} \right)^{\alpha_L} - \nu_L^{\alpha_L} + \left[ \nu_L^{\alpha_L} - (\nu_L + 1)^{\alpha_L} \right] \frac{\partial}{\partial x} \right\} C(x,\tau),$$
  

$$\mathbb{R}(\alpha_L) < 2, \ \mathbb{R}(\nu_L + \partial/\partial x) > 0, \ \mathbb{R}(\nu_L) > 0.$$
(7)

In special cases this equation changes to

$$\frac{\partial}{\partial \tau} C(x,\tau) = \lambda_L \left\{ \log \left( \nu_L + \frac{\partial}{\partial x} \right) - \log(\nu_L) - \log \left( \frac{\nu_L + 1}{\nu_L} \right) \frac{\partial}{\partial x} \right\}$$

$$\alpha_L = 0, \ \mathbb{R}(\nu_L + \partial/\partial x) > 0, \ \mathbb{R}(\nu_L) > 0,$$
(8)

and

$$\begin{split} & \frac{\partial}{\partial \tau} C(\mathbf{x},\tau) = \lambda_L \Big\{ -\nu_L \log \nu_L + \left[ \nu_L \log \nu_L - (\nu_L + 1) \log(\nu_L + 1) \right] \frac{\partial}{\partial x} + (\nu_L + \frac{\partial}{\partial x}) \log \left( \nu_L + \frac{\partial}{\partial x} \right) \Big\} C(\mathbf{x},\tau) \\ & \alpha_R = 1, \ \mathbb{R}(\partial/\partial x) < 0, \ \mathbb{R}(\nu_L) > 0, \end{split}$$

A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 7 / 13

Remember that, for instance, the first equation could be represented in the form

$$\frac{\partial}{\partial t}C(x,t) = -\mathcal{B}C(x,t) \tag{9}$$

This equation can be formally solved analytically to give

$$C(x,t) = e^{\mathcal{B}(T-t)}C(x,T), \qquad (10)$$

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Suppose that the whole time space is uniformly divided into N steps, so the time step θ = T/N is known. Assuming that the solution at time step k, 0 ≤ k < N is known and we go backward in time, we could rewrite the above as</p>

$$C^{k+1}(x) = e^{\mathcal{B}\theta} C^k(x), \qquad (11)$$

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To get representation of the rhs of the Eq. (11) with given order of approximation in θ, we can substitute the whole exponential operator with its Padé approximation of the corresponding order m.
 A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 8 / 13

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A.Itkin, P.Carr

Using pseudo-parabolic equations for option pricing in jump diffusion models

April 28, 2009 8 / 13

#### Padé approximations

**1** First, consider the case m = 1. A symmetric Padé approximation of the order (1, 1) for the exponential operator is

$$e^{\mathcal{B}\theta} = \frac{1 + \mathcal{B}\theta/2}{1 - \mathcal{B}\theta/2} \tag{12}$$

so we obtain a Crank-Nicolson scheme

$$\left(1 - \frac{1}{2}\mathcal{B}\theta\right)C^{k+1}(x) = \left(1 + \frac{1}{2}\mathcal{B}\theta\right)C^{k}(x).$$
(13)

The case *m* = 2 could be achieved either by using symmetric (2,2) or diagonal (1,2) Padé approximations of the operator exponent. The (1,2) Padé approximation reads

$$e^{\mathcal{B}\theta} = \frac{1 + \mathcal{B}\theta/3}{1 - 2\mathcal{B}\theta/3 + \mathcal{B}^2\theta^2/6},\tag{14}$$

and the corresponding finite difference scheme for the solution of the Eq. (11) is

$$I - 2\mathcal{B}\theta/3 + \mathcal{B}^2\theta^2/6] C^{k+1}(x) = [I + \mathcal{B}\theta/3] C^k(x)$$
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which is of the third order in  $\theta$ .

#### A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 9 / 13

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A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 9 / 13

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A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 9 / 13

### Approximation in space - finite variation

Exclude the last compensator term in the PIDE - move it to a differential part. Use a one-sided forward approximation of the first derivative which is a part of the operator  $\left(\nu_R - \frac{\partial}{\partial x}\right)^{\alpha_R}$ . Define  $h = (x_{max} - x_{min})/N$  to be the grid step in the x-direction, N is the total number of steps,  $x_{min}$  and  $x_{max}$  are the left and right boundaries of the grid. Also define  $c_i^k = C^k(x_i)$ . To make our method to be of the second order in x we use the following numerical approximation

$$\frac{\partial C^{k}(x)}{\partial x} = \frac{-C_{i+2}^{k} + 4C_{i+1}^{k} - 3C_{i}^{k}}{2h} + O(h^{2})$$
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A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 10 / 13

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To resolve this we propose another scheme which is inspired by Meerschaert and Tadjeran (2004). In this paper the authors proved that using a standard Grunwald-Letnikov approximation of the fractional derivative gives rise to a unconditionally unstable scheme. Instead they proposed to use a shifted Grunwald-Letnikov approximation which resolved this instability and produced a stable scheme. So by analogy we will use a shifted finite difference approximation of the first derivative in the fractional operator and unshifted version in the non-fractional operator. Then the unconditional stability could be proved again.

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  - To summarize these results let us represent them in Table 1,2 where F means the forward approximation, B means the backward approximation and S means the shifted approximation.

	Operator	
$\alpha < 0$	F	В
$0 < \alpha < 1$	F	

OperatorB<sup>-</sup>B<sup>+</sup>1 partSFSB2 partFB

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1 part	SF	SB	
2 part	F	В	

Table: Finite difference approximation of the operators at  $1 < \alpha_R < 2$  and  $1 < \alpha_L < 2$ .

- **()** Special cases  $\alpha_r = 0, 1, \alpha_l = 0, 1$  are not considered in Cartea (2007). In Boyarchenko & Levendorsky (2002) a corresponding characteristic function of the KoBoL process is obtained in these cases as well, but they did not consider the fractional PDE and its numerical solution.
- In Cartea (2007) a Crank-Nicolson type numerical scheme was proposed to solve the obtained FPDE which is of the first order in space. Here we derived high-order schemes in both time and space.

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- We show that despite it is a common practice to integrate out all Lévy compensators in the integral term when one considers jumps with finite activity and finite variation, this breaks the stability of the scheme, at least for the fractional PDE. Therefore, in order to construct the unconditionally stable scheme one must keep the other terms under the integrals. To resolve this in Cartea (2007) the authors were compelled to change their definition of the fractional derivative.

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- We also proposed the idea of solving FPDE with real α by using interpolation between option prices computed for the closest integer values of α. For the latter an efficient scheme is proposed that results in LU factorization of the band matrix O((1 α)<sup>2</sup>N/2).

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A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 12 / 13

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**Our contribution** 



# Thank you!

A.Itkin, P.Carr Using pseudo-parabolic equations for option pricing in jump diffusion models April 28, 2009 13 / 13