

**Numerical methods for the option pricing in markets with jumps**

ROBERTO NATALINI

Istituto per le Applicazioni del Calcolo “M. Picone”



Consiglio Nazionale delle Ricerche

**Collaborators to this research:**

Maya Briani (IAC-CNR & LUISS)

Anna Lisa Amadori (IAC-CNR & Univ. Salerno)

Claudia La Chioma (IAC-CNR): **now moved to a bank**

Giovanni Russo (Univ. Catania)



## WHAT IS AN OPTION? (EUROPEAN CALL)

**CONTRACT:** The holder has the right to buy at a fixed data in the future (maturity date  $T$ ) a prescribed asset (underlying  $S$ ) at a prescribed price (strike price or exercise price  $K$ ).

The other party (writer) has the obligation to sell.

This  $T$ -contract is defined by the random amount (**payoff**)

$$\phi(S_T) = \max [S_T - K, 0],$$

---

## PROBLEM

► How to price the option?

The value of the option depends on:

- price of the underlying;
- time to expiry

- ▶  $(S_\tau)_{\tau \in [0, T]}$  is a stochastic process on a filtered probability space;
- ▶  $r$  is the riskless interest rate;

In an arbitrage-free market, the price of a European option with payoff  $\phi(S_T)$  on an underlying  $S_t$  may be computed as

$$C_t(\phi(S_T)) = e^{-r(T-t)} \mathbb{E}^{\mathcal{Q}}[\phi(S_T) | \mathcal{F}_t]$$

with respect to some measure  $\mathcal{Q}$  such that  $\tilde{S}_t = e^{-rt} S_t$  is a martingale.

option pricing model



specify the law of  $(S_\tau)_{\tau \in [0, T]}$  under  $\mathcal{Q}$   
the “risk neutral” dynamics of  $S$

## BLACK & SCHOLES MODEL (1973)

---

$$S_t = S_0 \exp \left[ \left( r - \frac{\sigma^2}{2} \right) t + \sigma W_t \right]$$

- ▶  $W_t$  is a Brownian motion;
- ▶  $\sigma$ : volatility coefficient, constant.

ASSUMPTIONS:      no arbitrage opportunities      continuous trading  
no transaction costs/dividends      completeness of the market

$$C_t \text{ call on } S_t, (T, K) : \quad C(S, T; \sigma, r) = \text{Max}(S - K, 0).$$

$$\text{For } 0 \leq t \leq T : \quad C(S, t; \sigma, r) = S \mathcal{N}(d_1) - K e^{-r(T-t)} \mathcal{N}(d_2),$$

$$\mathcal{N}(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}y^2} dy, \quad d_{1,2} = \frac{\log(S/K) \pm (r + \frac{1}{2}\sigma^2)(T-t)}{\sigma\sqrt{T-t}}.$$

## MERTON MODEL (1976)

---

$$S_t = S_0 \exp \left[ \mu t + \sigma W_t + \sum_{i=1}^{N_t} Y_i \right]$$

- ▶  $W_t$  is a Brownian motion;  $\sigma$  volatility of the Brownian
- ▶  $N_t$  is a Poisson process with intensity  $\lambda$  independent from  $W$ ;
  - $N_t$  counts the number of jumps up to and including time  $t$ ;
- ▶  $Y_i \sim N(m, \delta^2)$  are i.i.d. random variables independent from  $W, N$ .
  - $Y_i$  corresponds to a proportional jump in the asset price

- Such markets are **incomplete**: there are many possible choices for a “risk-neutral” measure  $\Rightarrow$  *option hedging is a risky affair: one has to specify a way to measure this risk and then try to minimize it*
- There are different ways to measure risk (then to choose  $Q$ ): superhedging, utility maximization, ...
- Merton model:

$$\mu^M = r - \frac{\sigma^2}{2} - \lambda \mathbb{E}[e^{Y_i} - 1] = r - \frac{\sigma^2}{2} - \lambda[\exp(m + \delta^2/2) - 1],$$

$Q^M$  obtained by shifting the drift of the Brownian motion but leaving the jump part unchanged.

- Such markets are **incomplete**: there are many possible choices for a “risk-neutral” measure  $\Rightarrow$  *option hedging is a risky affair: one has to specify a way to measure this risk and then try to minimize it*
- There are different ways to measure risk (then to choose  $Q$ ): superhedging, utility maximization, ...
- Merton model:

$$\mu^M = r - \frac{\sigma^2}{2} - \lambda \mathbb{E}[e^{Y_i} - 1] = r - \frac{\sigma^2}{2} - \lambda[\exp(m + \delta^2/2) - 1],$$

$Q^M$  obtained by shifting the drift of the Brownian motion but leaving the jump part unchanged.

**THIS IS NOT THE TOPIC OF THE TALK!**

The aim is to be able to calculate the price of an European option which guarantees the payoff  $\phi(S_T)$  at time  $T$ :

Two approaches:

1. ... as discounted expectations

$$C_t = \mathbb{E} \left[ e^{-r(T-t)} \phi(S_T) | \mathcal{F}_t \right]$$

2. ... as deterministic function  $C_t = V(S_t, t)$ ;  
by means of Ito's calculus,

$V(S, t)$  is solution of a Differential Problem



The aim is to be able to calculate the price of an European option which guarantees the payoff  $\phi(S_T)$  at time  $T$ :

Two approaches:

1. ... as discounted expectations

$$C_t = \mathbb{E} \left[ e^{-r(T-t)} \phi(S_T) | \mathcal{F}_t \right]$$

▶ The price estimator is:

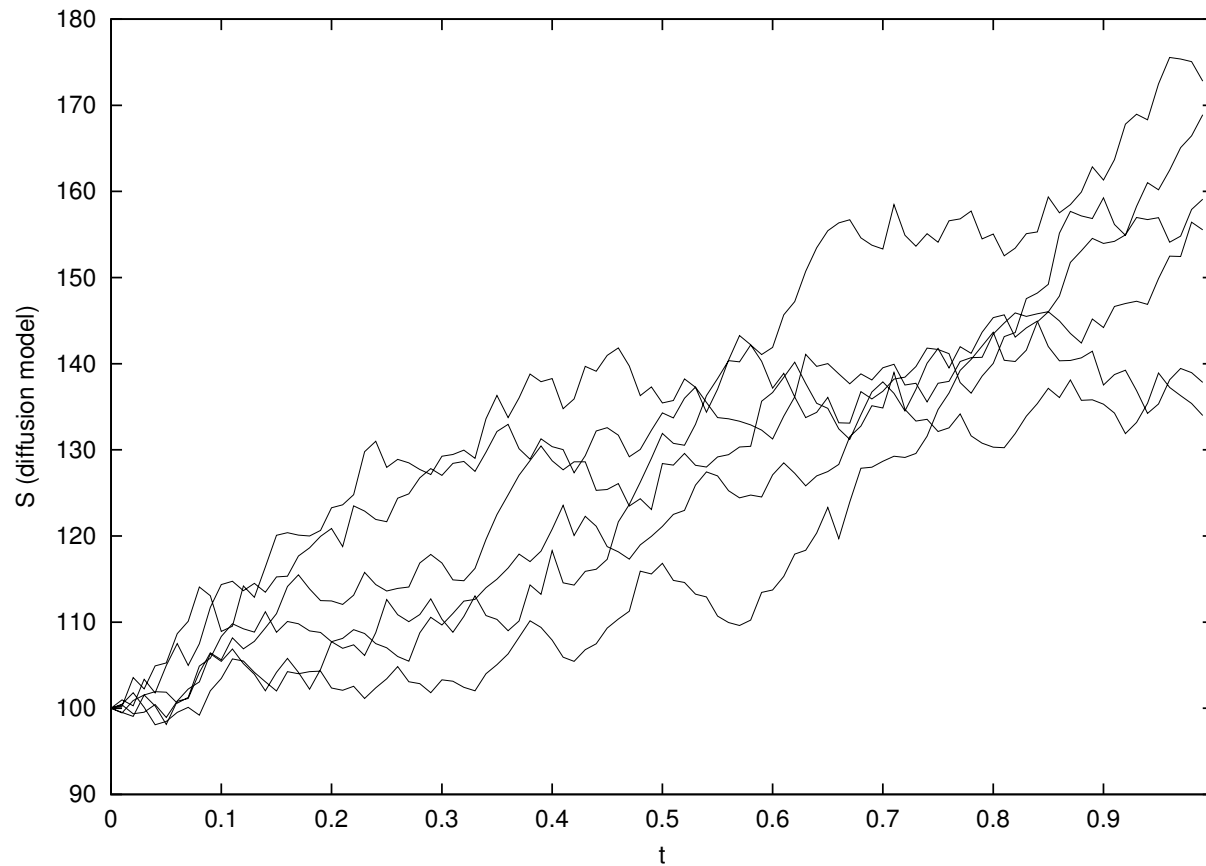
$$C_0 \approx \frac{e^{-rT}}{n} \sum_{i=1}^n \phi(S_T^{(i)})$$

2. ... as deterministic function  $C_t = V(S_t, t)$ ;

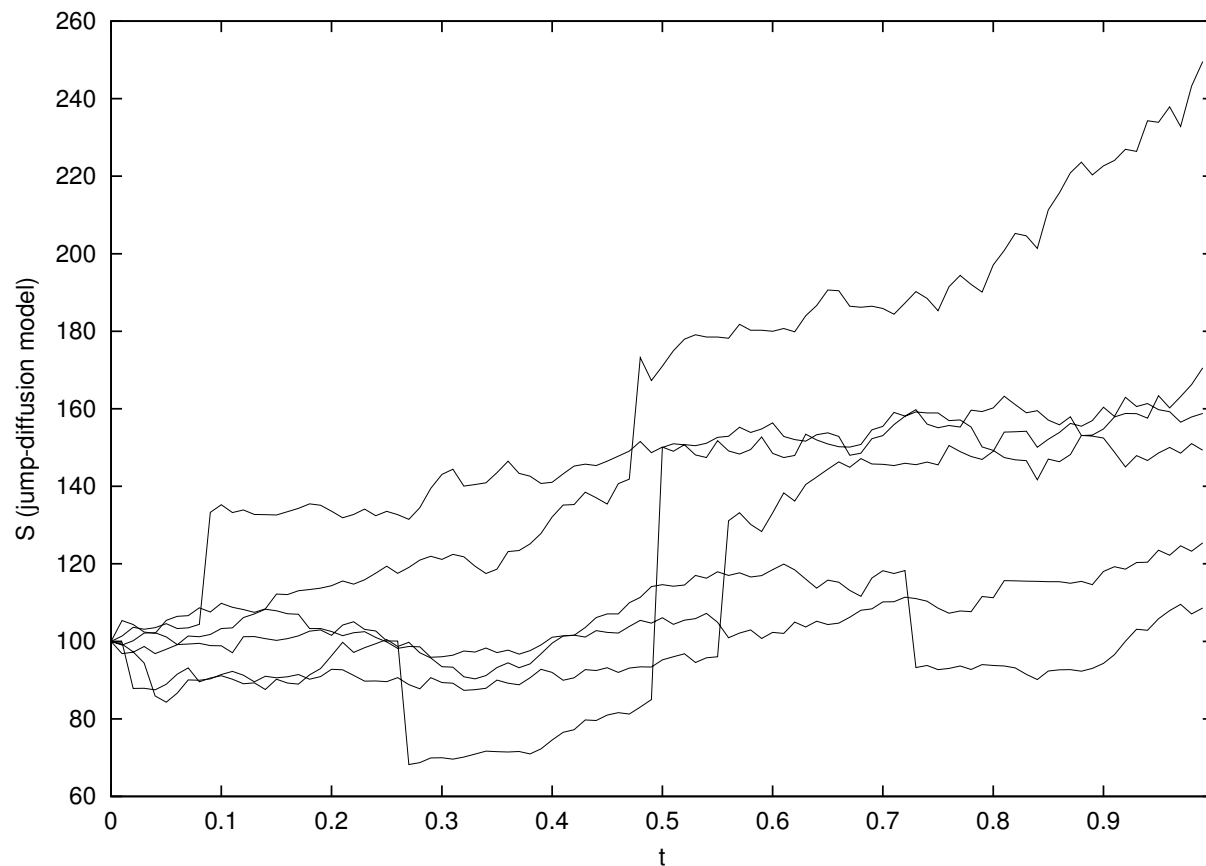
by means of Ito's calculus,

$V(S, t)$  is solution of a Differential Problem

▶ Finite Difference Schemes



**Figure 1:**  $\sigma = 0.15$ ,  $r = 0.5$  and  $\Delta t = 10^{-2}$



**Figure 2:**  $\lambda = 1$ ,  $\sigma = 0.15$ ,  $r = 0.5$ ,  $\gamma = 0$ ,  $\delta = 0.4$  and  $\Delta t = 10^{-2}$ .

# CALL OPTION PRICES $V_t(S)$ AT THE EXERCISE PRICE $S = K = 100$

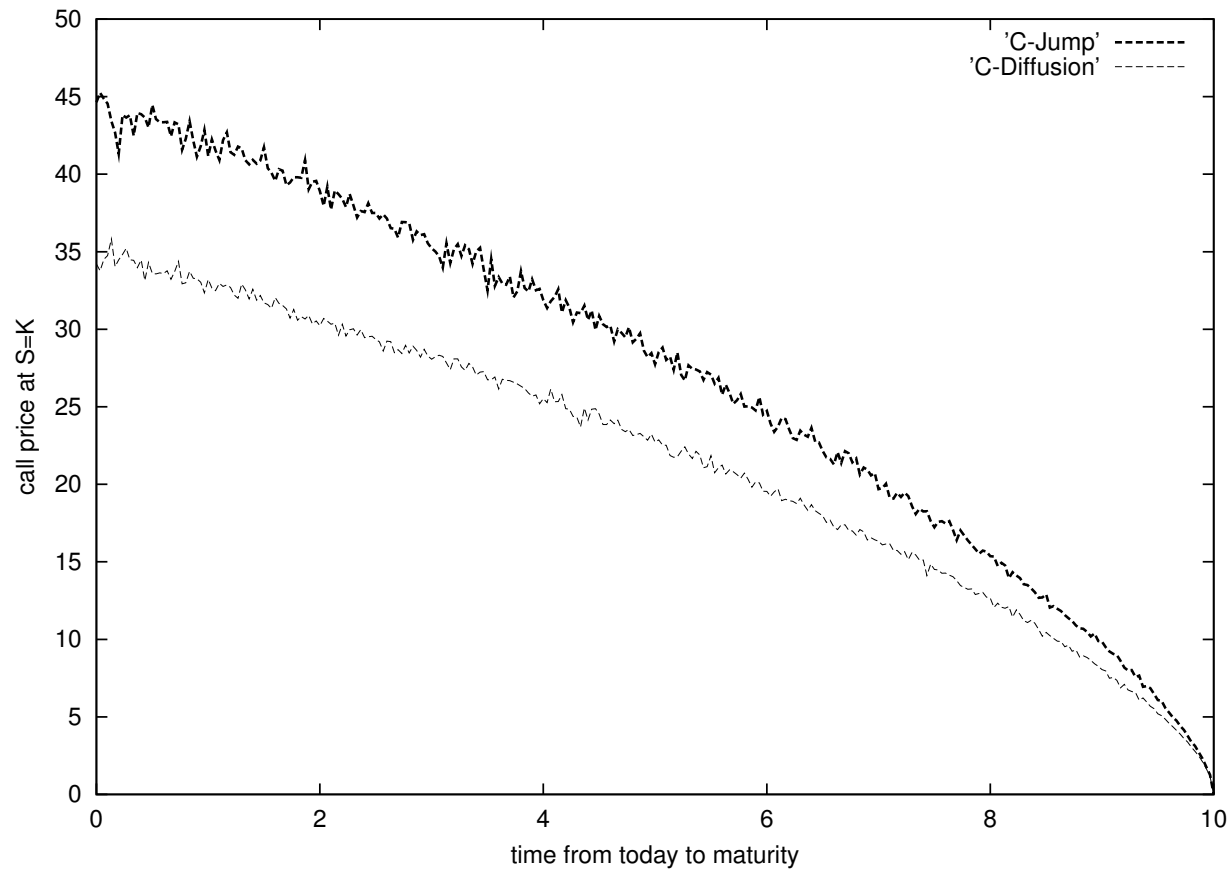


Figure 3:  $t$  from today to maturity  $T = 10$

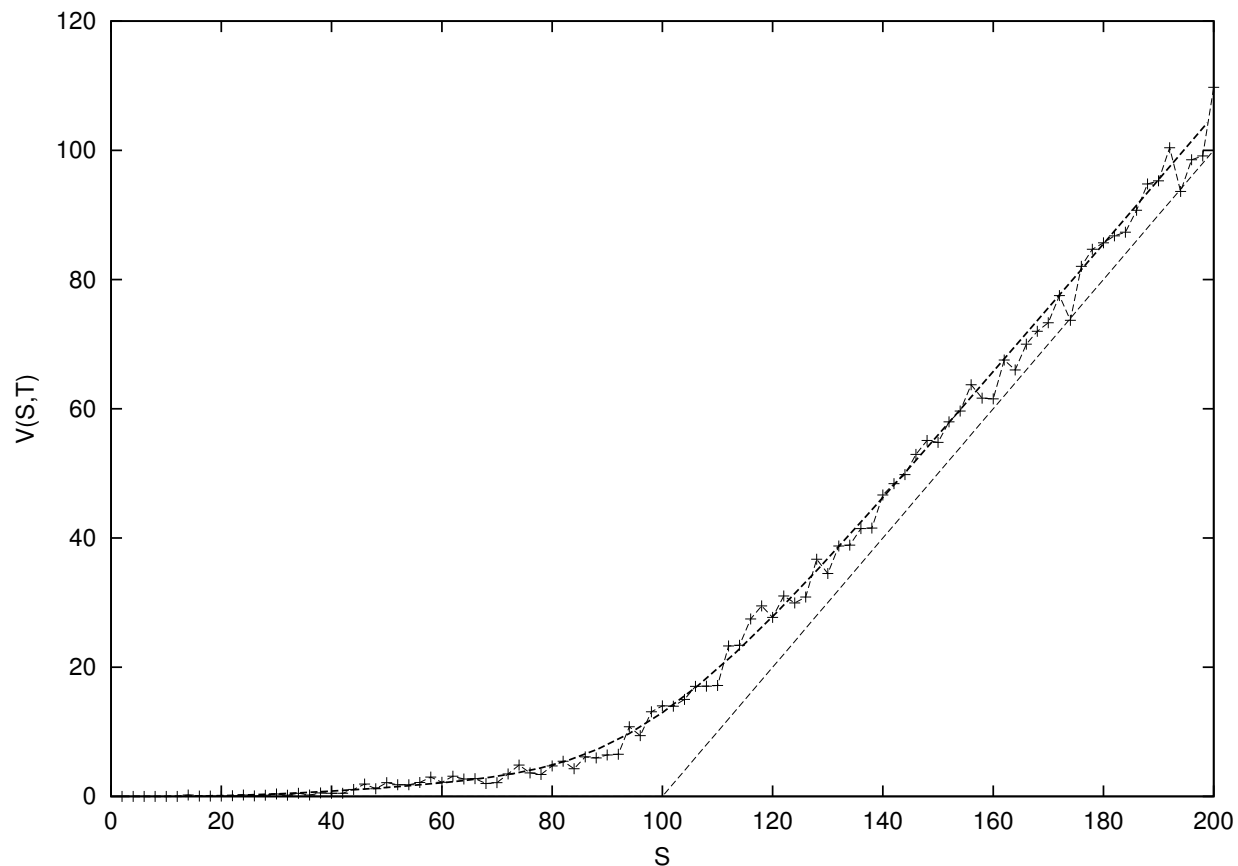


Figure 4: Call option price computed by the Midpoint-122 approximation (—) and by the Monte Carlo algorithm ( $\times$ )

- In **Diffusion models**  $C_t = V(S_t, t)$  (value of the European option) given by a **PARABOLIC DIFFERENTIAL EQUATION**:

$$\frac{\partial V}{\partial t}(S, t) + rS \frac{\partial V}{\partial S}(S, t) + \frac{\sigma^2 S^2}{2} \frac{\partial^2 V}{\partial S^2}(S, t) - rV(S, t) = 0$$

- In **Jump-Diffusion models**  $C_t = V(S_t, t)$  is given by a **PARTIAL INTEGRO-DIFFERENTIAL EQUATION**

$$\frac{\partial V}{\partial t}(S, t) + (r - \lambda \mathbb{E}[\eta - 1])S \frac{\partial V}{\partial S}(S, t) + \frac{\sigma^2 S^2}{2} \frac{\partial^2 V}{\partial S^2}(S, t) - rV(S, t) + \lambda \underbrace{\left( \int_0^{+\infty} V(S\eta) \tilde{\Gamma}(\eta) d\eta - V \right)}_{\mathcal{J}V} = 0$$

with boundary conditions depending on the type of option considered.

JUMPS  $\implies$  INTEGRAL TERM IN THE EQUATION

▶  $\mathcal{J}V$  is a **nonlocal** term: it depends on the whole solution  $V(\cdot, t)$  and not only on its behavior at the point  $S$ .

$\implies$  new **theoretical** and **numerical** issues

PLAN

- ▶ (quite short) analytical backgrounds
- ▶ A general numerical convergence result
- ▶ Numerical algorithms (numerical boundary conditions...)
- ▶ High order (in time) Implicit-Explicit methods
- ▶ Well-balanced schemes for nonlocal Pbs.

▶ **Uniformly parabolic problems:** Existence and uniqueness of **classical solutions** for  $\sigma > 0$  (Garroni-Menaldi 1995)

▶ **Problems:**

- **degeneracy of coefficients** (*incomplete market*)
- **non linearity** (*large investor economy: the interest rate  $r$  is influenced by the agents*)
- in general, solutions are **not regular**

▶ The class of *viscosity solutions* is sufficiently large to allow for existence of solutions in the cases where smooth solutions do not exist, it is sufficiently small to obtain uniqueness, using the *comparison principle*



$$\begin{cases} \partial_t u - \mathcal{L}_{\mathcal{I}}(x, t, \mathcal{I}, \mathcal{D}, \mathcal{D}^2)u + H(x, t, \mathcal{D}u, \mathcal{I}u) = 0 \\ u(x, 0) = u_0(x) \end{cases}$$

- $u_0 \in \mathcal{C}(\mathbb{R}^d)$ ;
  - $\mathcal{L}_{\mathcal{I}}$  **linear degenerate elliptic operator:**  

$$\mathcal{L}_{\mathcal{I}}(x, t, \mathcal{I}, \mathcal{D}, \mathcal{D}^2)u = \frac{1}{2} \text{tr}[\sigma \sigma^T(x, t) \mathcal{D}^2 u] + (\mu - \gamma k)^T(x, t) \mathcal{D}u - \mathcal{I}u;$$
  - $H$  **nonlinear first order operator;**
  - $\mathcal{I}u = \int_{\mathbb{R}^d} M(u(x+z, t), u(x, t)) \mu_{x,t}(dz)$ ;  
 $\mu_{x,t}$  **positive bounded measure;  $M$  continuous function, s.t.:**  

$$M(u, v) \leq M(w, v) \text{ if } u \leq w; M(u, u) = 0;$$

$$M(u, v) - M(w, z) \leq c((u - w)_+ + |v - z|).$$
- ▶ Amadori (2000): [Existence and uniqueness for Bounded Lévy processes](#)
- ▶ Amadori, La Chioma, Karlsen (2004): [General Lévy processes](#)

$$F(x, t, u, \mathcal{I}u, \mathcal{D}u, \mathcal{D}^2u) = -\mathcal{L}_{\mathcal{I}}(x, t, \mathcal{I}, \mathcal{D}, \mathcal{D}^2)u + H(x, t, \mathcal{D}u, \mathcal{I}u)$$

## *Problem*

$$\partial_t u + F(x, t, u, \mathcal{I}u, \mathcal{D}u, \mathcal{D}^2u) = 0. \quad (1)$$

**Numerical grid in  $\mathbb{R}^d \times [0, T]$**

- $\mathbf{h} \in \mathbb{R}^d$ ,  $\mathbf{k} \in \mathbb{R}$ : **space, time grid steps.**
- $(\mathbf{x}_j, \mathbf{t}_n) = (jh, nk)$ ,  $j \in \mathbb{Z}^d$ ,  $n \in \mathbb{N}$ ,  $\mathbf{v}_j^n = v(\mathbf{x}_j, \mathbf{t}_n)$ ,  $\tilde{\mathbf{v}}^n = (\mathbf{v}_j^n)_j$ .
- $\mathcal{I}_h \tilde{\mathbf{v}}$  **integral approximation.**

## *Scheme*

$$Q(h, k, j, n, v_j^n, \mathcal{I}_h \tilde{v}, \tilde{v}) = 0. \quad (2)$$

$$Q(h, k, j, n, v_j^n, \mathcal{I}_h \tilde{v}, \tilde{v}) = 0$$

[H1] **Stability**

[H2] **Consistency**

► [H3] **MONOTONICITY OF THE APPROXIMATION OF THE INTEGRAL**

[H4] **Monotonicity ► also for the integral part**

[H5] **Comparison Principle for the problem**

The scheme is said to be **STABLE** if for a bounded initial condition, the solution  $v_j^n$  is uniformly bounded at all point of the grid, independetely from  $k, h$ :

$$\exists C > 0, \forall k, h > 0, j \in \{0, \dots, N\}, n \in \{0, \dots, M\} : |v_j^n| \leq C.$$

► *Stability ensures that the numerical solution does not blow up when  $(h, k) \rightarrow 0$*

---

The scheme is said to be (locally) **CONSISTENT** with the continuous equation if the discretized operator  $Q$  converges to its continuous version:  $\forall \psi \in \mathcal{C}^\infty(\mathbb{R} \times [0, T]), \forall (x, t) \in \mathbb{R} \times [0, T],$

$$Q(h, k, j, n\psi_j^n, \mathcal{I}_h \tilde{\psi}, \tilde{\psi}) \rightarrow F(x, t, u, \mathcal{I}u, \mathcal{D}u, \mathcal{D}^2u) \text{ as } (h, k) \rightarrow 0$$

**MONOTONICITY OF THE INTEGRAL APPROXIMATION:** **if**  $v_j^n = w_j^n$   
**and**  $\tilde{v} \leq \tilde{w}$ , **then**

$$\mathcal{I}_h \tilde{v} \leq \mathcal{I}_h \tilde{w}$$

---

**MONOTONICITY PROPERTY:** **if**  $v_j^n = w_j^n$  **and**  $\tilde{v} \leq \tilde{w}$ , **then**

$$Q(h, k, j, n, v_j^n, \mathcal{I}_h \tilde{v}, \tilde{v}) \geq Q(h, k, j, n, w_j^n, \mathcal{I}_h \tilde{w}, \tilde{w})$$

---

**MONOTONICITY OF THE INTEGRAL APPROXIMATION:** if  $v_j^n = w_j^n$   
and  $\tilde{v} \leq \tilde{w}$ , then

$$\mathcal{I}_h \tilde{v} \leq \mathcal{I}_h \tilde{w}$$

---

**MONOTONICITY PROPERTY:** if  $v_j^n = w_j^n$  and  $\tilde{v} \leq \tilde{w}$ , then

$$Q(h, k, j, n, v_j^n, \mathcal{I}_h \tilde{v}, \tilde{v}) \geq Q(h, k, j, n, w_j^n, \mathcal{I}_h \tilde{w}, \tilde{w})$$

---

**⇒ DISCRETE COMPARISON PRINCIPLE:** if  $w^0, v^0$  are two initial conditions, then

$$w^0 \geq v^0 \Rightarrow \forall n \geq 1, w^n \geq v^n$$

► It has an important *financial interpretation*: it is equivalent to say that the option values computed using the scheme verify **arbitrage inequalities**: *inequalities between payoff lead to inequalities between values of options*

**Theorem** Let assumption (H1)–(H5) hold true. Then,  
as  $(h, k) \rightarrow 0$ , the solution  $\tilde{u}$  of (2) converges  
locally uniformly to the unique continuous viscosity solution

- ▶ Convergence - purely second order problems: Barles, Souganidis (1991)
- ▶ M. Briani, C. La Chioma, R. Natalini (preprint 2001, Num. Matematik 2004)
- ▶ R. Cont, E. Voltchkova (preprint 2003, linear case)
- ▶ C. La Chioma (2004, PhD Thesis, general Lévy)

$$\left\{ \begin{array}{l} \frac{\partial u}{\partial t}(x, t) + a \frac{\partial u}{\partial x}(x, t) = b \frac{\partial^2 u}{\partial x^2}(x, t) - cu(x, t) + \mathcal{I}u(x, t) \quad (x, t) \in \mathbb{R} \times [0, T] \\ u(x, 0) = u_0(x) \end{array} \right.$$

$$\mathcal{I}u = \int_{-\infty}^{+\infty} [u(x+z, t) - u(x, t)] \Gamma_{\delta}(z) dz$$

with

$$\Gamma_{\delta} \geq 0 \quad \text{and} \quad \int_{\mathbb{R}} \Gamma_{\delta}(x) dx = 1.$$

## OPTION PRICING - MERTON MODEL

$$u_0(x) = (e^x - E)_+, \quad (\text{call option}); \quad \Gamma_{\delta}(x) = \lambda \frac{1}{\delta \sqrt{2\pi}} \exp\left(-\frac{x^2}{2\delta^2}\right)$$



▶ Truncation of the problem domain  $\longrightarrow [a, b] \in \mathbb{R}$

▶ Truncation of the integral domain  $\int_{-\infty}^{+\infty} \longrightarrow \int_{-z_\epsilon}^{+z_\epsilon}$

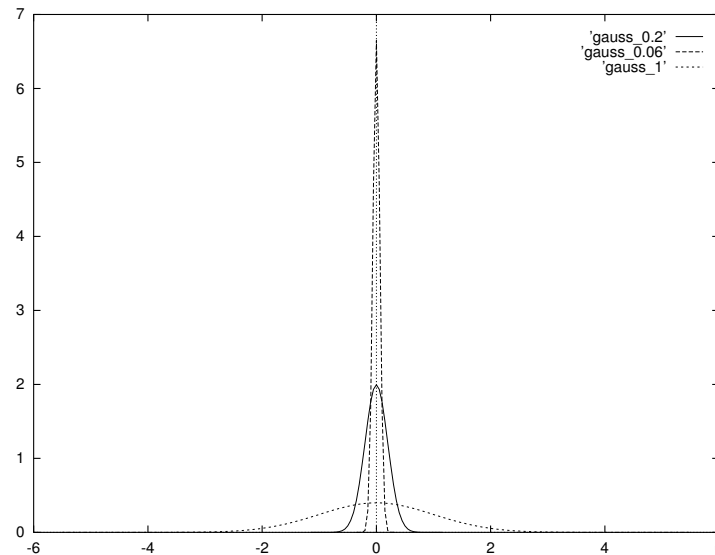
▶ **Difficulty: nonlocal nature of the integral term**

$\Rightarrow$  relation between integral and problem domain (**boundary conditions**).

$\Rightarrow$  **implicit** time differencing scheme (unconditionally stable)  
**not** practicaly feasible.

# TRUNCATION OF THE INTEGRAL DOMAIN (GAUSSIAN DISTRIBUTION)

$$\mathcal{I}u = \int_{-\infty}^{+\infty} [u(x+z, t) - u(x, t)] \Gamma_{\delta}(z) dz, \quad \Gamma_{\delta}(z) = \frac{1}{\delta\sqrt{2\pi}} \exp\left(-\frac{z^2}{2\delta^2}\right)$$



$$\Gamma_{\delta}(z) \geq \varepsilon \iff z \in [-z_{\varepsilon}, z_{\varepsilon}], \quad z_{\varepsilon} = \sqrt{-2\delta^2 \log(\varepsilon\delta\sqrt{2\pi})}$$

if  $u(\cdot; t) \in Lip(\mathbb{R})$

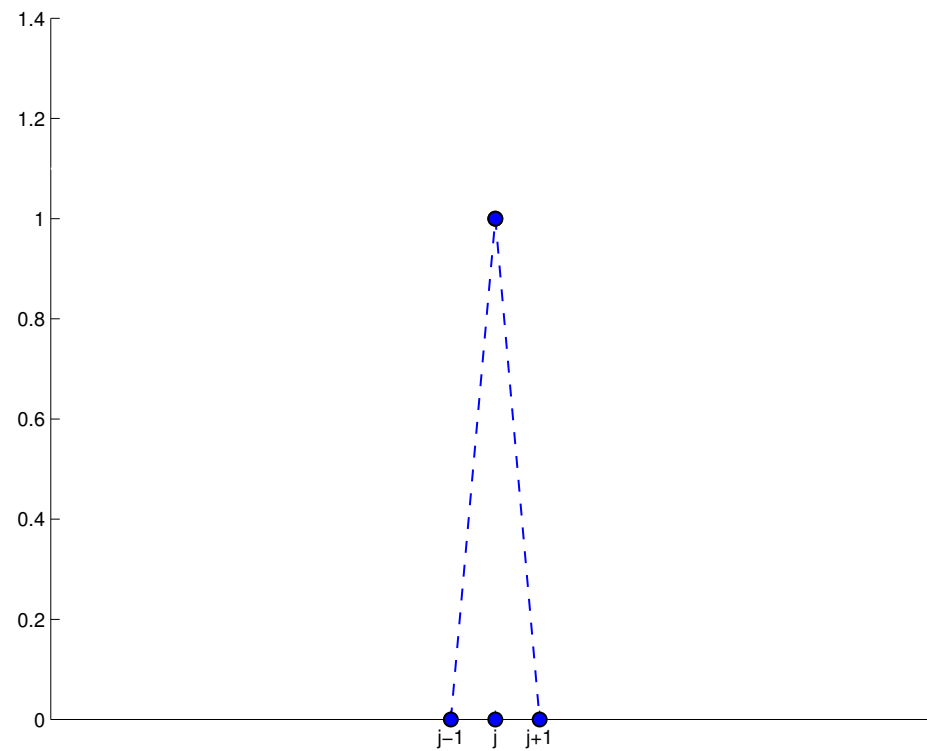
$$\left| \int_{-\infty}^{+\infty} [u(x+z, t) - u(x, t)] \Gamma_{\delta}(z) dz - \int_{-z_{\varepsilon}}^{+z_{\varepsilon}} [u(x+z, t) - u(x, t)] \Gamma_{\delta}(z) dz \right| \leq c\delta^2\varepsilon$$



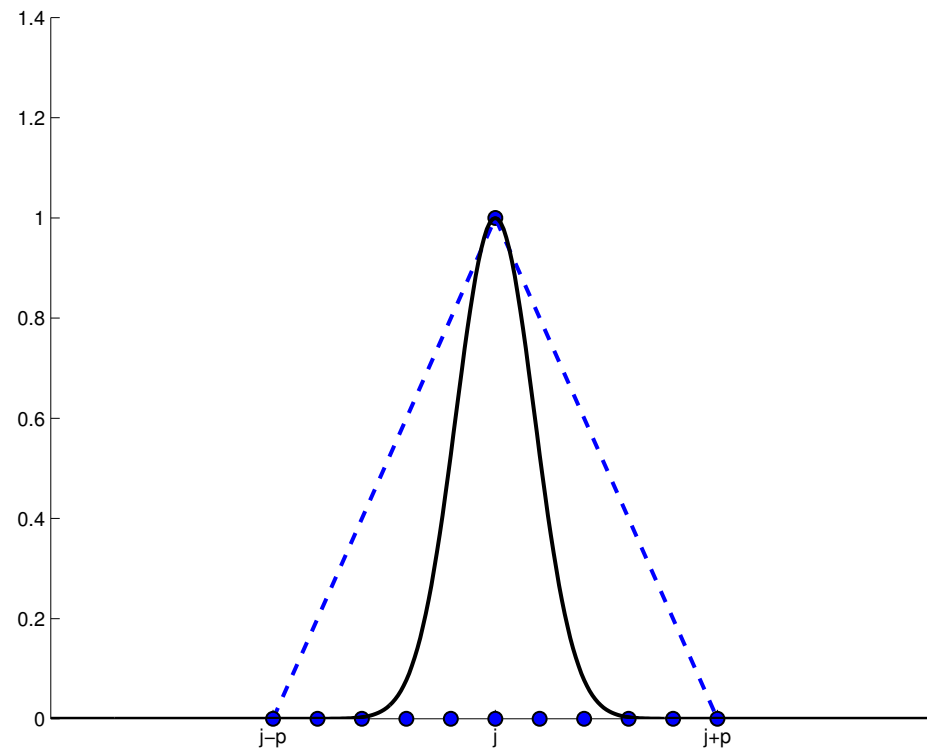
**For**  $j = 0, \dots, M,$

$$\begin{aligned} \frac{dv_j}{dt}(t) &= \left( \frac{b}{h^2} + \frac{a}{2h} \right) v_{j-1}(t) + \left( \frac{2b}{h^2} + c \right) v_j(t) + \left( \frac{b}{h^2} - \frac{a}{2h} \right) v_{j+1}(t) \\ &\quad + \lambda \left[ h \sum_{l=-p}^p \beta_l v_{j+l}(t) (\Gamma_\delta)_l + h \sum_{i < -p, i > p} \alpha_i v_{j+i}(t) (\Gamma_\delta)_i - v_j(t) \right], \end{aligned}$$

$$\frac{dv_j}{dt} = kw_{-1}v_{j-1} + (1 - kw_0)v_j + kw_1v_{j+1}$$

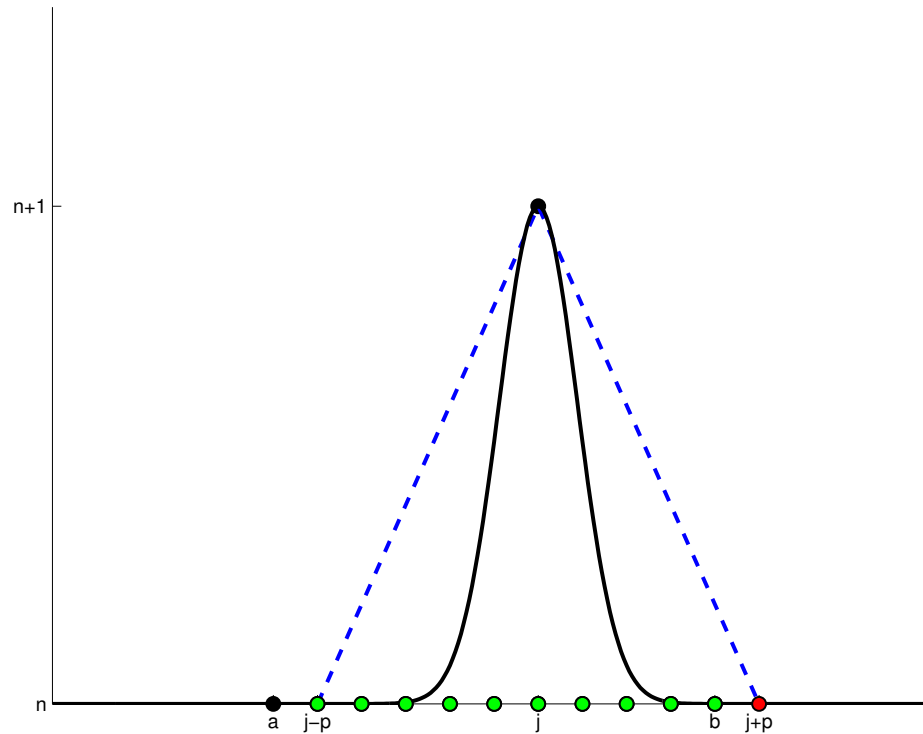


$$+ \lambda kh \sum_{i=-p}^p \alpha_i v_{j+i} (\Gamma_\delta)_i$$



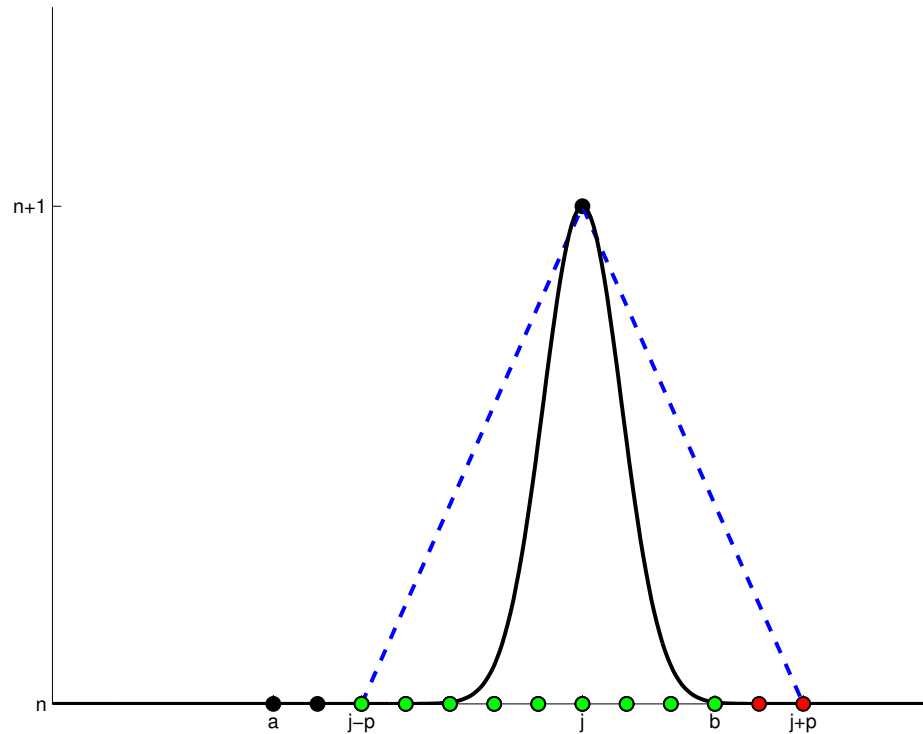
# NUMERICAL DOMAIN

$$+\lambda kh \sum_{i=-p}^p \alpha_i v_{j+i}(\Gamma_\delta)_i + \lambda kh \sum_{i>p} \alpha_i \overset{?}{v_{j+i}} (\Gamma_\delta)_i,$$



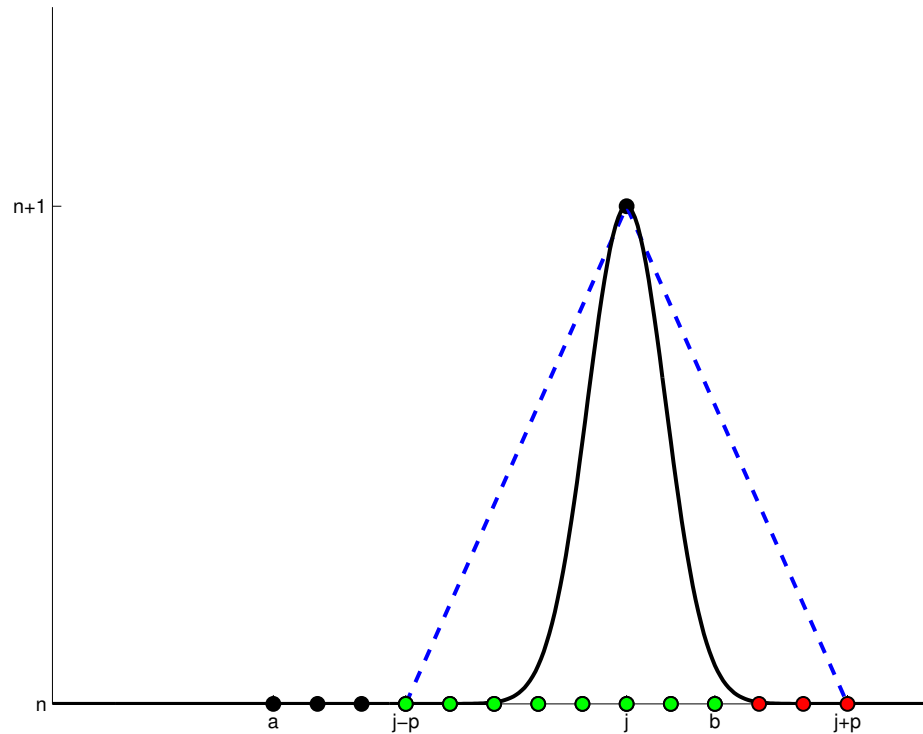
# NUMERICAL DOMAIN

$$+\lambda kh \sum_{i=-p}^p \alpha_i v_{j+i}(\Gamma_\delta)_i + \lambda kh \sum_{i>p} \alpha_i \overset{?}{v_{j+i}} (\Gamma_\delta)_i,$$



# NUMERICAL DOMAIN

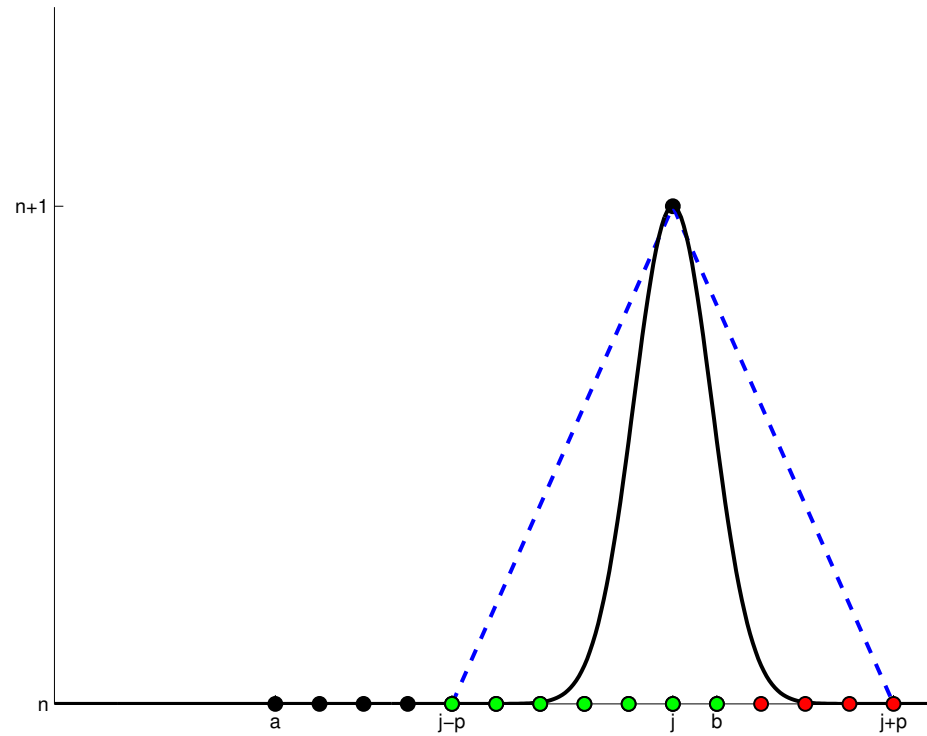
$$+\lambda kh \sum_{i=-p}^p \alpha_i v_{j+i}(\Gamma_\delta)_i + \lambda kh \sum_{i>p} \alpha_i \overset{?}{v_{j+i}} (\Gamma_\delta)_i,$$





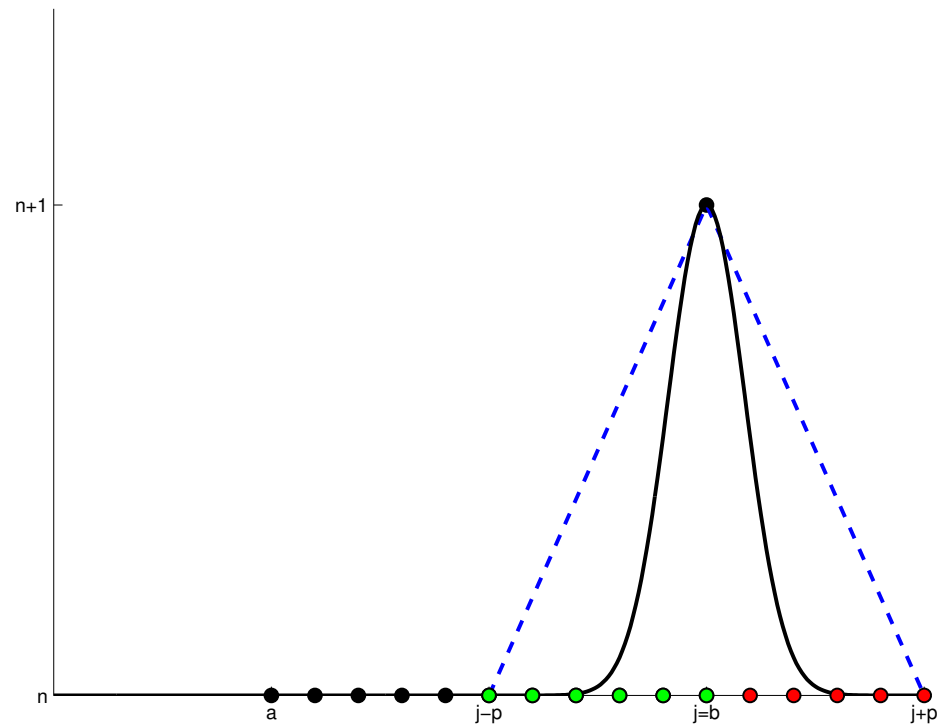
# NUMERICAL DOMAIN

$$+\lambda kh \sum_{i=-p}^p \alpha_i v_{j+i}(\Gamma_\delta)_i + \lambda kh \sum_{i>p} \alpha_i \overset{?}{v_{j+i}} (\Gamma_\delta)_i,$$



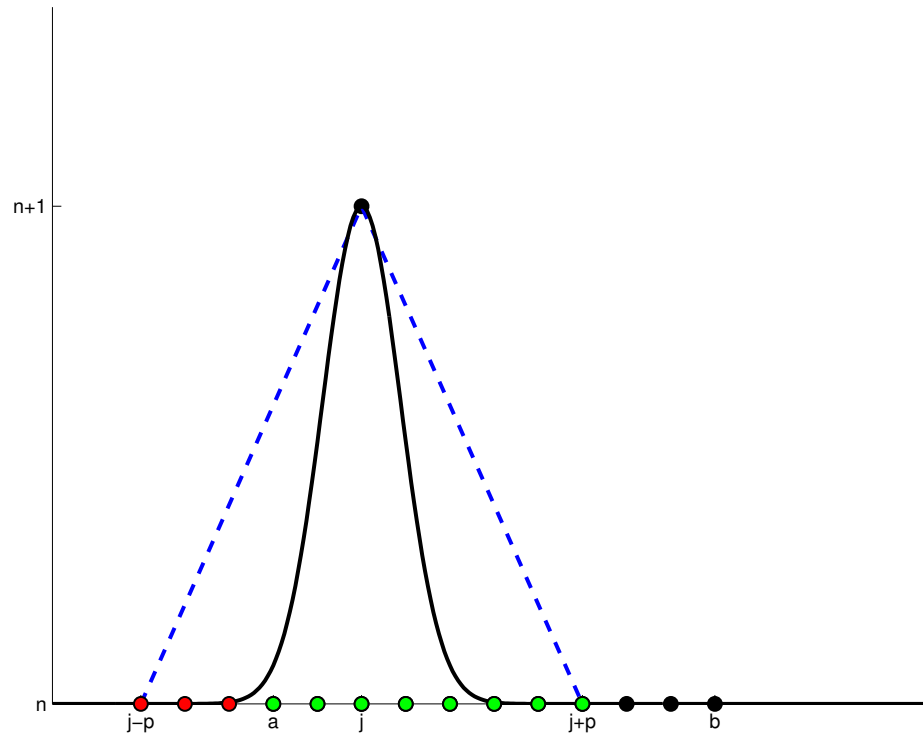
# NUMERICAL DOMAIN

$$+ \lambda kh \sum_{i=-p}^p \alpha_i v_{j+i} (\Gamma_\delta)_i + \lambda kh \sum_{i>p} \alpha_i \overset{?}{v_{j+i}} (\Gamma_\delta)_i,$$



# NUMERICAL DOMAIN

$$+ \lambda k h \sum_{i=-p}^p \alpha_i v_{j+i} (\Gamma_\delta)_i + \lambda k h \sum_{i < -p} \alpha_i \overset{?}{v_{j+i}} (\Gamma_\delta)_i,$$



► We are going to approximate our one space dimension operator

$$u_t + au_x - bu_{xx} + cu = \frac{1}{\delta\sqrt{2\pi}} \int_{-\infty}^{\infty} [u(x+z, t) - u(x, t)] \exp\left(-\frac{z^2}{2\delta^2}\right) dz$$

by using the Taylor expansion of the integral term

$$v_t + av_x - bv_{xx} + cv = \frac{\lambda\delta^2}{2} v_{xx}$$

where  $\lambda = \int_{-\infty}^{\infty} z^2 \exp\left(-\frac{z^2}{2\delta^2}\right) dz$

► If  $\delta \ll 1$ , and  $u(x, 0) = v(x, 0) = u_0(x) \in L^1(\mathbb{R}) \cap L^\infty(\mathbb{R})$ ,

$$\|u - v\|_{L^\infty(0, T; L^1(\mathbb{R}))} \leq O(T\delta^3).$$

$$u_t(x, t) + \mathcal{H}(Du(x, t), \mathcal{I}u(x, t)) = \mathcal{G}(u(x, t), D^2u(x, t))$$

**Euler forward:** 
$$\frac{u_j^{n+1} - u_j^n}{k} + \mathcal{H}(u_j^n, \mathcal{I}_h u_j^n) = \mathcal{G}(u_j^n)$$

► The scheme is second order in time **but** stable under the parabolic CFL condition:

$$k \leq ch^2 \rightsquigarrow h = 1/100 \Rightarrow k = 1/10000 \quad !!!$$

**Euler backward:** 
$$\frac{u_j^{n+1} - u_j^n}{k} + \mathcal{H}(u_j^{n+1}, \mathcal{I}_h u_j^{n+1}) = \mathcal{G}(u_j^{n+1})$$

► The scheme is unconditionally stable **but** not practically feasible:

$$\mathcal{I}_h \tilde{u}^{n+1} \rightsquigarrow \text{dense system} \quad !!!$$

---

## IMEX (IMPLICIT-EXPLICIT) TECHNIQUE

---

- ▶ **IMEX** technique has been introduced for time dependent partial-differential equations that involve terms of different types.
- ▶ **IMEX** schemes have been widely used for the time integration of spatially discretized PDEs of diffusion-convection type.
- ▶ Instances of these methods have been successfully applied to the Navier-Stokes equations ...
- ▶ For recent developments: L. PARESCHI AND G. RUSSO

$$u_t(x, t) + \mathcal{H}(Du(x, t), \mathcal{I}u(x, t)) = \mathcal{G}(u(x, t), D^2u(x, t))$$

**Implicit in  $\mathcal{G}(u(x, t), D^2u(x, t))$**   $\longrightarrow$

**To avoid  
parabolic CFL condition**

**To avoid  
dense system**

$\longleftarrow$  **Explicit in  $\mathcal{H}(Du(x, t), \mathcal{I}u(x, t))$**

**Main goal  $\longrightarrow$  second (or higher) order accuracy in time**

► **The Midpoint(1, 2, 2) scheme**

$$\begin{cases} u^{(2)} = u^n - \frac{k}{2}\mathcal{H}(Du^n, \mathcal{I}u^n) + \frac{k}{2}\mathcal{G}(u^{(2)}, D^2u^{(2)}) \\ u^{n+1} = u^n - k\mathcal{H}(Du^{(2)}, \mathcal{I}u^{(2)}) + k\mathcal{G}(u^{(2)}, D^2u^{(2)}) \end{cases}$$

► **A two-stage, third-order DIRK scheme**

$$\begin{cases} u^{(2)} = u^n - k\gamma\mathcal{H}(Du^n, \mathcal{I}u^n) + k\gamma\mathcal{G}(u^{(2)}, D^2u^{(2)}) \\ u^{(3)} = u^n - k(\gamma - 1)\mathcal{H}(Du^n, \mathcal{I}u^n) - 2k(\gamma - 1)\mathcal{H}(Du^{(2)}, \mathcal{I}u^{(2)}) \\ \quad + k(1 - 2\gamma)\mathcal{G}(u^{(2)}, D^2u^{(2)}) + k\gamma\mathcal{G}(u^{(3)}, D^2u^{(3)}) \\ u^{n+1} = u^n - \frac{k}{2}\left(\mathcal{H}(Du^{(2)}, \mathcal{I}u^{(2)}) + \mathcal{H}(Du^{(3)}, \mathcal{I}u^{(3)})\right) \\ \quad + \frac{k}{2}\left(\mathcal{G}(u^{(2)}, D^2u^{(2)}) + \mathcal{G}(u^{(3)}, D^2u^{(3)})\right) \end{cases}$$



$$u_t(x, t) + \mathcal{H}(Du(x, t), \mathcal{I}u(x, t)) = \mathcal{G}(u(x, t), D^2u(x, t)). \quad (3)$$

► **Time approximation, for  $i = 1, \dots, \nu$**

$$\left\{ \begin{array}{l} u^{(i)} = u^n - k \sum_{j=1}^{\nu} \tilde{a}_{ij} \mathcal{H}(Du^{(j)}, \mathcal{I}u^{(j)}) + k \sum_{j=1}^{\nu} a_{ij} \mathcal{G}(u^{(j)}, D^2u^{(j)}) \\ u^{n+1} = u^n - k \sum_{i=1}^{\nu} \tilde{\omega}_i \mathcal{H}(Du^{(i)}, \mathcal{I}u^{(i)}) + k \sum_{i=1}^{\nu} \omega_i \mathcal{G}(u^{(i)}, D^2u^{(i)}) \end{array} \right. \quad (4)$$

The matrices  $\tilde{A} = (\tilde{a}_{ij})$ ,  $\tilde{a}_{ij} = 0$  for  $j \geq i$  and  $A = (a_{ij})$  are  $\nu \times \nu$  matrices such that the resulting scheme is explicit in  $H$  and implicit in  $G$ .

► **The Midpoint(1, 2, 2) scheme**

$$\tilde{A} = (\tilde{a}_{ij}) = \begin{pmatrix} 0 & 0 \\ 1/2 & 0 \end{pmatrix} \quad A = (a_{ij}) = \begin{pmatrix} 0 & 0 \\ 0 & 1/2 \end{pmatrix}$$

$$(\tilde{\omega}_1, \tilde{\omega}_2) = (0, 1) \quad (\omega_1, \omega_2) = (0, 1).$$

► **A two-stage, third-order scheme ARS-233  
(Ascher-Ruuth-Spiteri) ( $\gamma = (3 + \sqrt{6})/3$ )**

$$\tilde{A} = (\tilde{a}_{ij}) = \begin{pmatrix} 0 & 0 & 0 \\ \gamma & 0 & 0 \\ \gamma - 1 & 2 - 2\gamma & 0 \end{pmatrix} \quad A = (a_{ij}) = \begin{pmatrix} 0 & 0 & 0 \\ 0 & \gamma & 0 \\ 0 & 1 - 2\gamma & \gamma \end{pmatrix}$$

$$(\tilde{\omega}_1, \tilde{\omega}_2, \tilde{\omega}_3) = (0, 1/2, 1/2) \quad (\omega_1, \omega_2, \omega_3) = (0, 1/2, 1/2).$$

► **VON NEUMANN STABILITY ANALYSIS:** represents  $v^n$  as a discrete Fourier transform and requires the corresponding Fourier coefficients to be stable ( $\longrightarrow$  essentially equivalent to stability in  $L^2$  norm)

$\Rightarrow$  advancing the solution of the scheme by one time step is equivalent to multiplying the Fourier transform of the solution by the **AMPLIFICATION FACTOR**  $g(h\xi)$ .

$$\hat{u}^n(\xi) = g(h\xi)\hat{u}^0(\xi).$$

**Theorem (von Neumann)** A one-step finite difference scheme is stable if and only if there is a constant  $K$  and some time and space steps  $k_0$  and  $h_0$  s.t.:

$$|g(\theta, k, h)| \leq 1 + Kk$$

If  $g(\theta, k, h)$  is independent of  $h$  and  $k$ , one can choose  $K = 0$

**Proposition** Given a stable one-step finite difference scheme for the differential operator, the scheme is stable for every bounded additional term (e.g.: Integral operator), for small values of the mesh size.

$$g(\theta; h, k) = 1 + k(H(\theta; h)\tilde{\omega}^T + G(\theta; h)\omega^T)(I_\nu - k\tilde{\mathcal{A}}_\tau - k\mathcal{A}_\tau)^{-1}e.$$

$$e = (1, \dots, 1)^T, \quad U = (\hat{u}^{(1)}, \dots, \hat{u}^{(\nu)})^T, \quad U^n = (\hat{u}^n, \dots, \hat{u}^n)^T$$

$$\mathcal{A}_\tau = AG(\theta; h) \quad \tilde{\mathcal{A}}_\tau = \tilde{A}H(\theta; h)$$

► For the Merton's model, with centered finite-difference approximation,

$$H(\theta; h) = -iF + J = -i\frac{a}{h}\sin\theta + \lambda\left(\int_{-\infty}^{+\infty} e^{iz\theta/h}\Gamma_\delta(z)dz - 1\right),$$

$$G(\theta; h) = \frac{2b}{h^2}(\cos\theta - 1).$$

(5)

**[Implicit-Explicit Midpoint(1,2,2)]** The scheme is second order in time.

The Midpoint scheme is stable for

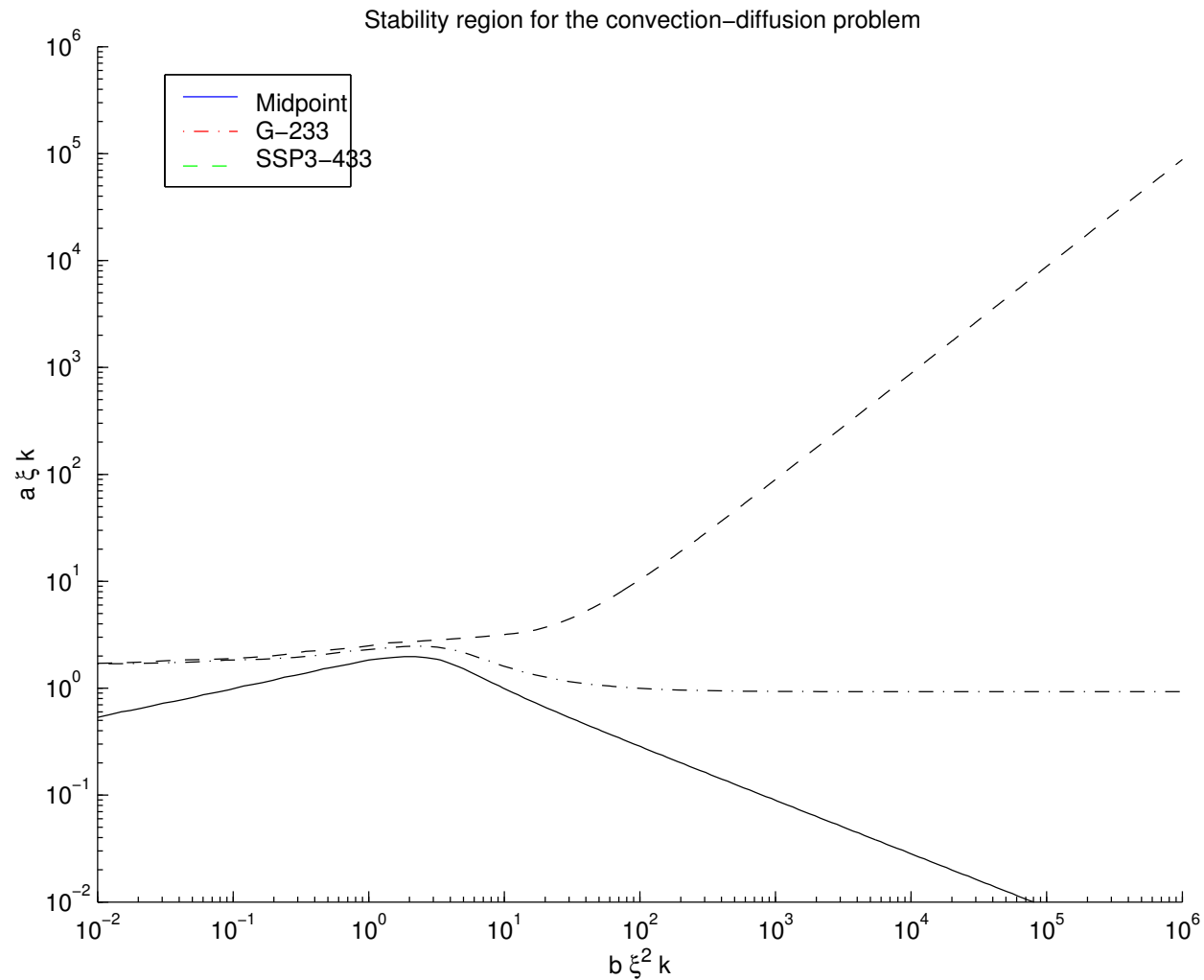
$$k \leq \frac{3}{2(a^2b)^{1/3}} h^{4/3}. \quad (6)$$

**[ARS(2,3,3)] (Ascher-Ruuth-Spiteri)** Fix  $\gamma = (3 + \sqrt{6})/3$ . The scheme is third order in time.

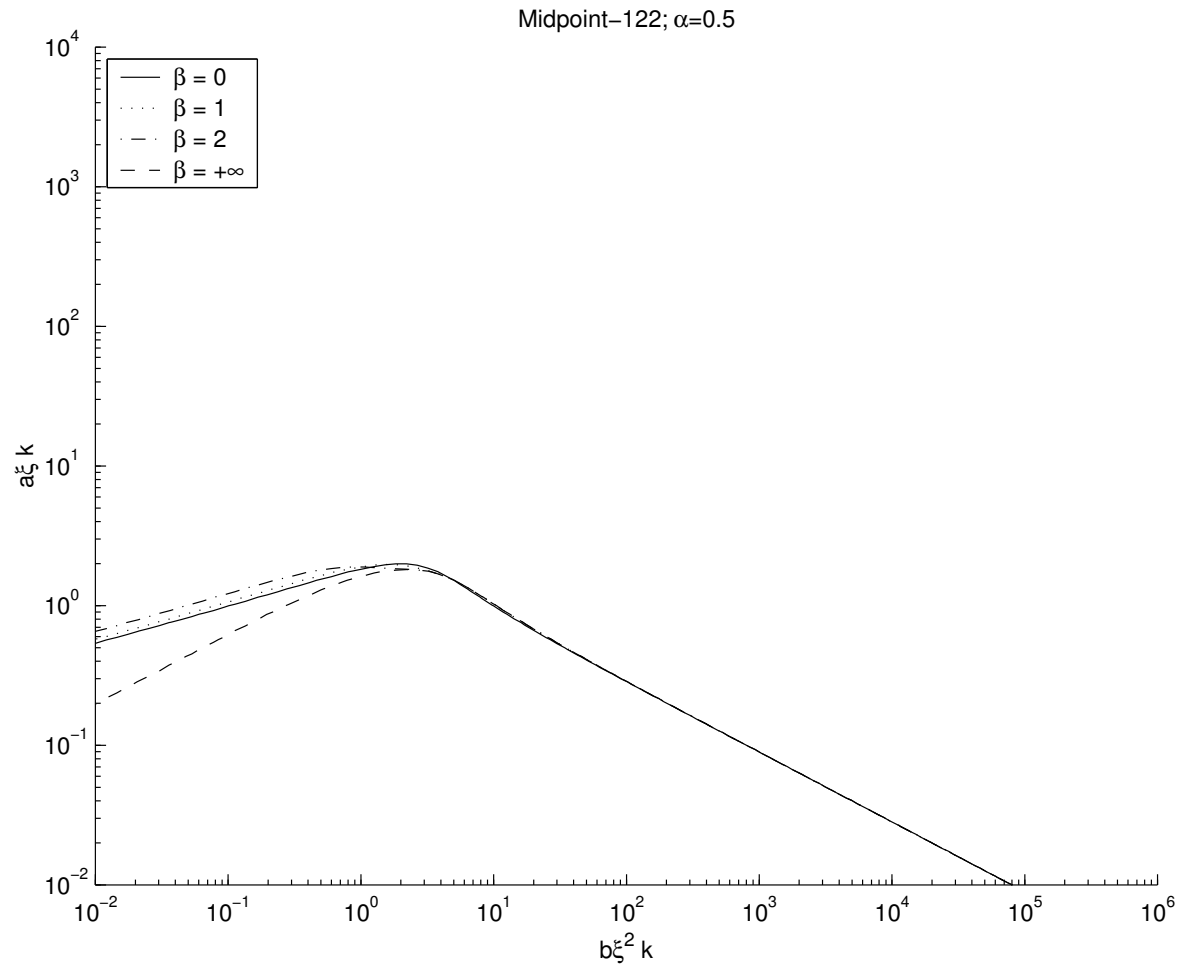
The scheme is stable for

$$k \leq \frac{\sqrt{4\gamma - 1}}{2\gamma} \frac{h}{a}. \quad (7)$$

# STABILITY - IMEX AMPLIFICATION FACTOR

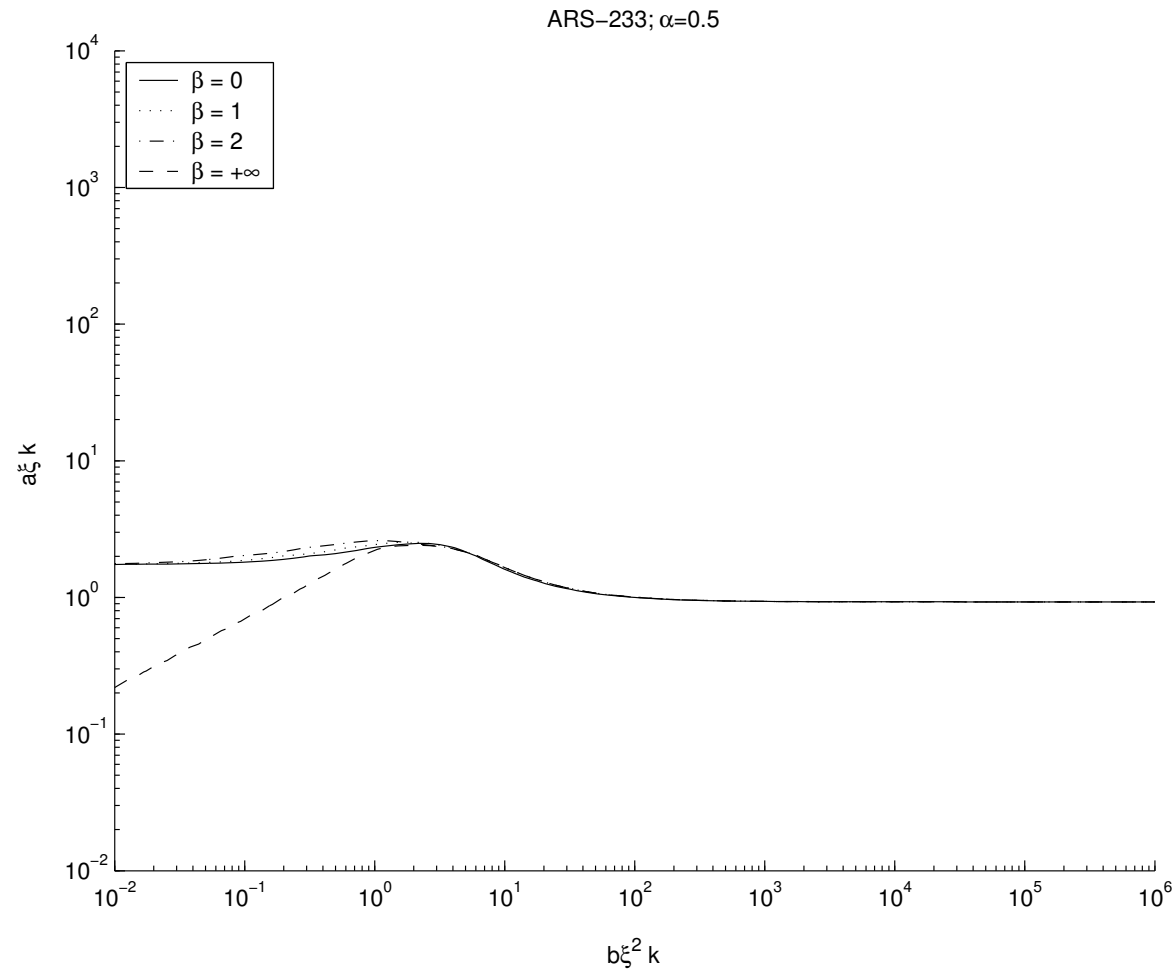


Advection-diffusion problem: border of the stability regions in the  $b\xi^2 k$ - $a\xi k$  plane.



**Integro-differential problem: stability regions of Midpoint scheme in the  $b\xi^2 k$ - $a\xi k$  plane.  $\alpha = \lambda b/a^2 = 0.5$  and  $\beta = \delta a/b$**





**Integro-differential problem: stability regions of ARS-233 scheme in the  $b\xi^2k$ - $a\xi k$  plane.  $\alpha = \lambda b/a^2 = 0.5$  and  $\beta = \delta a/b$**

**Proposition 0.1 (Order conditions)** *Let  $H$  and  $G$  be two given consistent space approximation of accuracy order  $O(h^\beta)$ , i.e.*

$$H(h\xi; h) + G(h\xi; h) = -q_1(\xi) - q_2(\xi) + O(h^\beta) = -q(\xi) + O(h^\beta)$$

*The scheme is accurate of order  $O(h^\beta + k^\alpha)$  if the two matrices  $\tilde{A}$  and  $A$  and the coefficients vectors  $\tilde{\omega} = (\tilde{\omega}_1, \dots, \tilde{\omega}_\nu)^T$  and  $\omega = (\omega_1, \dots, \omega_\nu)^T$  verify for  $j = 1, \dots, \alpha$*

$$\left( q_1(\xi)\tilde{\omega}^T + q_2(\xi)\omega^T \right) \left( \frac{d^{j-1}\Lambda(k)}{dk^{j-1}} \right)_{k=0} = (-1)^{j+1} \frac{\left( q_1(\xi) + q_2(\xi) \right)^j}{j},$$

*where  $\Lambda(k) = (I_\nu + k\tilde{A}q_1(\xi) + kAq_2(\xi))^{-1}e$ .*

- *first order* ( $\alpha = 1$ )

$$(\omega_1 + \omega_2) q_2 + (\tilde{\omega}_1 + \tilde{\omega}_2) q_1 = q_1 + q_2$$

- *second order* ( $\alpha = 2$ )

$$\begin{aligned} & (\tilde{a}_{21}\tilde{\omega}_2) q_1^2 + ((a_{22} + a_{21})\omega_2 + \omega_1 a_{11}) q_2^2 \\ & + (\tilde{a}_{21}\omega_2 + (a_{22} + a_{21})\tilde{\omega}_2 + \tilde{\omega}_1 a_{11}) q_1 q_2 = \frac{q_1^2}{2} + \frac{q_2^2}{2} + q_1 q_2. \end{aligned}$$

## COMPUTATIONAL COSTS

IMEX-DIRK scheme		
time	space	integral
$M \times$	$N$	$\times \mathcal{P}^2$ or $(3\mathcal{P} \log_2(\mathcal{P}) + \mathcal{P})$
		+
		...
		+
	$N$	$\times \mathcal{P}^2$ or $(3\mathcal{P} \log_2(\mathcal{P}) + \mathcal{P})$
	$= O(MN(3\mathcal{P} \log_2(\mathcal{P}) + \mathcal{P}))$	

The explicit approximation		
time	space	integral
$M \times$	$N$	$\times \mathcal{P}^2$ or $(3\mathcal{P} \log_2(\mathcal{P}) + \mathcal{P})$
	$= O(MN(3\mathcal{P} \log_2(\mathcal{P}) + \mathcal{P}))$	

## CPU TIMES

---

N	explicit scheme	Midpoint-122	ARS-233	SSP
256	0.02s	0.04s	0.04s	0.07s
512	0.3s	0.35s	0.32s	0.24s
1024	4.4s	3.21s	2.33s	1.73s
2048	1m56,6s	29.58s	17.62s	13.48s

Table 1: CPU times on 1.6 GHz Pentium IV PC when  $T = 1$ .

<i>T</i> = 1 Midpoint-122					
h	k	$l^1$	$\gamma_1$	$l^\infty$	$\gamma_\infty$
0.125000	0.066986	1.277415		1.126198	
0.062500	0.027205	0.456599	1.484227	0.551598	1.029771
0.031250	0.011049	0.140942	1.695827	0.150568	1.873204
0.015625	0.004487	0.034098	2.047359	0.026901	2.484656
0.007812	0.001822	0.019122	0.834463	0.017256	0.640585

Table 2: Errors and convergence orders of the Midpoint-122 scheme The process parameters are  $E = 100$ ,  $r = 0$ ,  $\sigma = 0.2$ ,  $\lambda = 0.1$ ,  $\delta = 0.5$ .

$T = 5$ Midpoint-122					
h	k	$l^1$	$\gamma_1$	$l^\infty$	$\gamma_\infty$
0.125000	0.066986	5.097443		0.606756	
0.062500	0.027205	1.095825	2.217757	0.188375	1.687508
0.031250	0.011049	0.296186	1.887442	0.069759	1.433159
0.015625	0.004487	0.093908	1.657181	0.019274	1.855694
0.007812	0.001822	0.031876	1.558776	0.002183	3.142337

Table 3: Errors and convergence orders of the Midpoint-122 scheme The process parameters are  $E = 100$ ,  $r = 0$ ,  $\sigma = 0.2$ ,  $\lambda = 0.1$ ,  $\delta = 0.5$ .

$T = 1$ ARS-233					
h	k	$l^1$	$\gamma_1$	$l^\infty$	$\gamma_\infty$
0.125000	0.125000	1.357048		1.408900	
0.062500	0.062500	0.334417	2.020753	0.318280	2.146199
0.031250	0.031250	0.090647	1.883314	0.072331	2.137617
0.015625	0.015625	0.028081	1.690643	0.017782	2.024171
0.007812	0.007812	0.010154	1.467582	0.004429	2.005380

Table 4: Errors and convergence orders of the ARS-233 scheme The process parameters are  $E = 100$ ,  $r = 0$ ,  $\sigma = 0.2$ ,  $\lambda = 0.1$ ,  $\delta = 0.5$ .



$T = 5$ ARS-233					
h	k	$l^1$	$\gamma_1$	$l^\infty$	$\gamma_\infty$
0.125000	0.125000	5.129721		0.584784	
0.062500	0.062500	1.056785	2.279198	0.138810	2.074792
0.031250	0.031250	0.245218	2.107543	0.034354	2.014543
0.015625	0.015625	0.076278	1.684720	0.008568	2.003464
0.007812	0.007812	0.031807	1.261932	0.002141	2.000850

Table 5: Errors and convergence orders of the ARS-233 scheme The process parameters are  $E = 100$ ,  $r = 0$ ,  $\sigma = 0.2$ ,  $\lambda = 0.1$ ,  $\delta = 0.5$ .

N	M	Explicit		M	ARS-233	
		PUT	CALL		PUT	CALL
256	46	8.167357	13.579657	17	8.230058	12.861993
512	185	8.268306	13.378464	35	8.289777	13.449324
1024	743	8.319940	13.286915	70	8.326102	13.287427
$V$		8.341444	13.218501		8.341444	13.218501

**Table 6:**  $V$  is the analytical price of the European put and call option of Merton model. The process parameters are  $T = 1$ ,  $E = 100$ ,  $r = 0.05$ ,  $\sigma = 0.2$ ,  $\lambda = 0.1$ ,  $\delta = 0.8$ ,  $x = \ln(100)$ .

---

## Well-balanced schemes for integro-differential problems

Work in progress (M. Briani, R. N.)



- ▶ Problems to correctly resolve the source term
- ▶ Main idea: find a discrete solver which preserves the steady state solutions
- ▶ LeRoux, Greenberg, Gosse, Shi Jin, LeVeque, Perthame, Simeoni, Bouchut ...

$$\textit{Example:} \begin{cases} u_t + u_x + u = 0 \\ u_0(x) = e^{-x} \end{cases} \Rightarrow u(x, t) = u_0(x)$$

Plain upwind

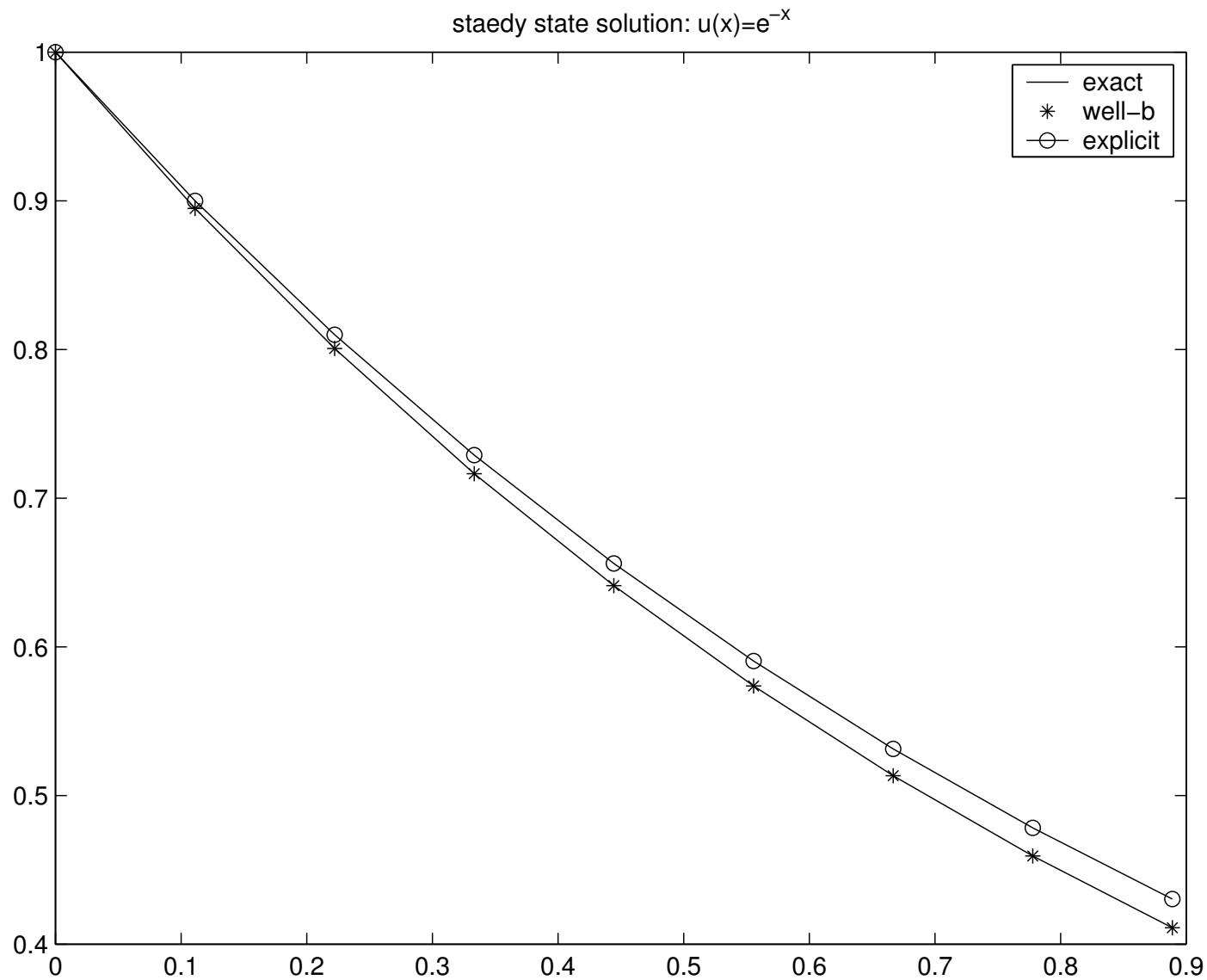
$$\frac{du_j}{dt} + \frac{u_j - u_{j-1}}{h} + u_j = 0$$

Well-balanced

$$\frac{du_j}{dt} + \frac{u_j - e^{-h}u_{j-1}}{h} = 0$$

(Botchorishvili-Perthame-Vasseur)

# “WELL-BALANCED SCHEMES”



- ▶ A simple Integro Differential Equation: the Hammer model

$$u_t + \frac{1}{2}(u^2)_x = \int K(x-y)(u(y) - u(x))dy \quad K(x) = \frac{1}{2}e^{-|x|}$$

↑

$$\begin{cases} u_t + \frac{1}{2}(u^2)_x = -q_x \\ -q_{xx} + q = -u_x \end{cases}$$

simplified model describing the evolution of a radiating gas

- ▶ Smooth solutions: S. Kawashima et al.
- ▶ Entropy solutions: P. Marcati et al.
- ▶ Asymptotic behavior: D.Serre

$$\begin{cases} u_t + au_x = K * u - u \\ u(x, 0) = u_0(x) \end{cases} \quad K(x) = \frac{1}{2}e^{-|x|}; \quad a > 0$$

### Steady States

$$\left\{ \begin{array}{l} w(x) = e^{\lambda x} \\ \lambda = 0 \quad \lambda = \Lambda = \frac{-1 + \sqrt{1 + 4a^2}}{2a} \end{array} \right.$$

For a given integral approximation:

$$K * u = \sum_{l=-m}^m \beta_l K_l u_{i+l} + O(h^d), \quad d \geq 1,$$

General Scheme:

$$h \frac{du_i}{dt} + \sum_{l=-m}^m \alpha_l(h) u_{i+l} = 0$$

Conditions on  $\{\alpha_l(h)\}_l$ :

▶ Consistency

▶ Well-balanced: 
$$\sum_{l=-m}^m \alpha_l(h) e^{\Lambda_j lh} = 0 \quad \forall j$$

$$\left\{ \begin{array}{l} \alpha_l(h) + h\beta_l K_l = 0 \quad \forall l \neq -1, 0, 1, \\ \alpha_{-1}(h) + \alpha_0(h) + \alpha_1(h) - h(1 - (\beta_{-1}K_{-1} + \beta_0K_0 + \beta_1K_1)) \\ -\alpha_{-1}(h) + \alpha_1(h) - a = c_1h \\ \sum_{l=-m}^m \alpha_l(h) e^{\Lambda_l h} = 0 \end{array} \right.$$

▶ Monotonicity:  $\alpha_0 > a$



Take, as for the upwind approximation  $\alpha_0 = h + a - h\beta_0 K_0$ , then

$$\alpha_{-1} = \frac{(1 - e^{\Lambda h})}{e^{\Lambda h} - e^{-\Lambda h}} a - \frac{h \left( -\Lambda^2 + e^{\Lambda h} \beta_{-1} K_{-1} (\Lambda^2 - 1) + \beta_{-1} K_{-1} e^{-\Lambda h} (1 - \Lambda^2) \right)}{(e^{\Lambda h} - e^{-\Lambda h}) (\Lambda^2 - 1)}$$

$$\alpha_1 = \frac{(-1 + e^{-\Lambda h})}{-e^{-\Lambda h} + e^{\Lambda h}} a - \frac{h \left( \Lambda^2 + e^{-\Lambda h} \beta_1 K_1 - e^{-\Lambda h} \beta_1 K_1 \Lambda^2 - \beta_1 K_1 e^{\Lambda h} + \beta_1 K_1 e^{\Lambda h} \Lambda^2 \right)}{(e^{\Lambda h} - e^{-\Lambda h}) (\Lambda^2 - 1)}$$

$$\begin{cases} u_t + au_x + u - K * u = 0 \\ u(x, 0) = u_0(x) \end{cases} \quad K(x) = \frac{1}{2}e^{-|x|}; \quad a > 0$$


---

$$\begin{cases} u_0(x) = e^{\Lambda x} + e^{\lambda x} & |\lambda| < 1 \\ u(x, t) = e^{\Lambda x} + e^{\delta t + \lambda x} & \delta = \frac{\lambda(a\lambda^2 + \lambda - a)}{1 - \lambda^2} \end{cases}$$

$$\delta < 0, \quad \lim_{t \rightarrow +\infty} u(x, t) = w(x) = e^{\Lambda x}$$

# TEST 1

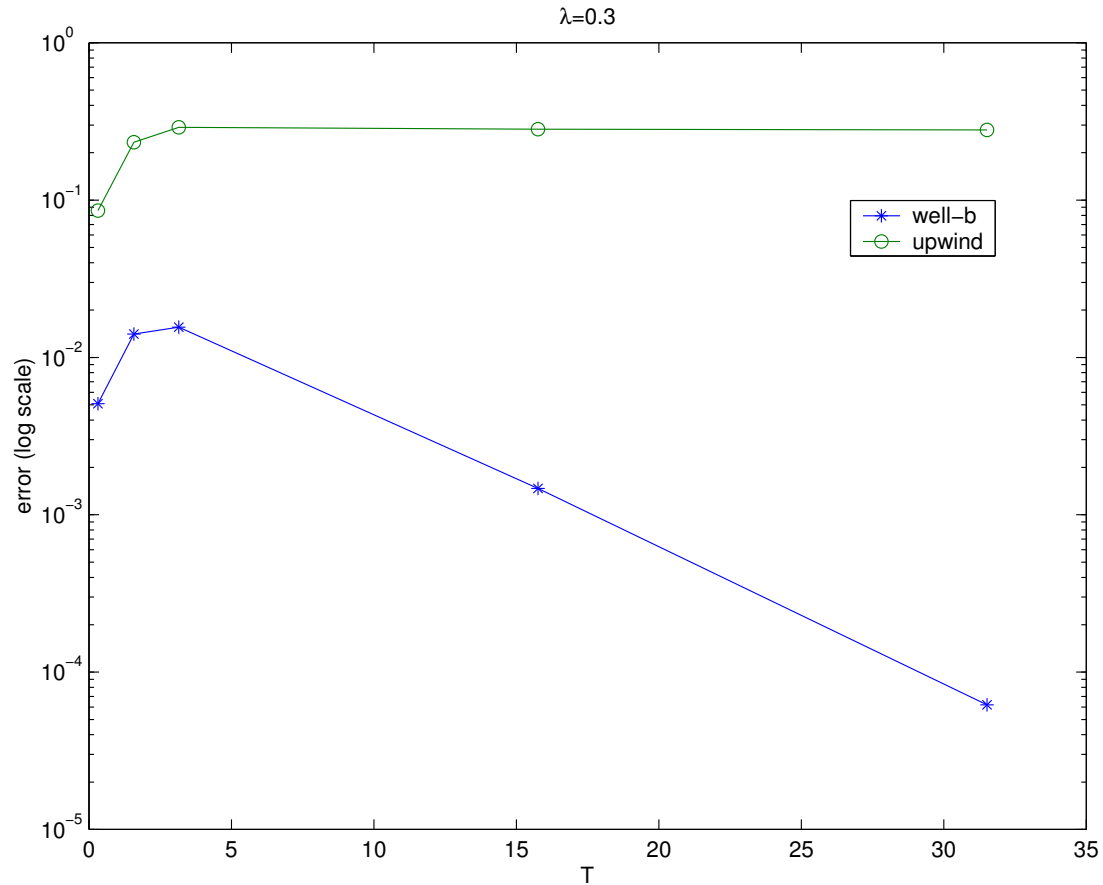


Figure 5:  $l^\infty$  error in log. scale for increasing  $T$

$$\begin{cases} u_t + au_x + u - K * u = 0 \\ u(x, 0) = u_0(x) \end{cases} \quad K(x) = \frac{1}{2}e^{-|x|}; \quad a > 0$$

---


$$\begin{cases} u_0(x) = e^{\Lambda x} + \sin(x)\Xi_{[0,\pi]} \\ u(x, t) = e^{\Lambda x} + O(t^{-1/4}) \Leftarrow (D.Serre) \end{cases}$$

## TEST 2

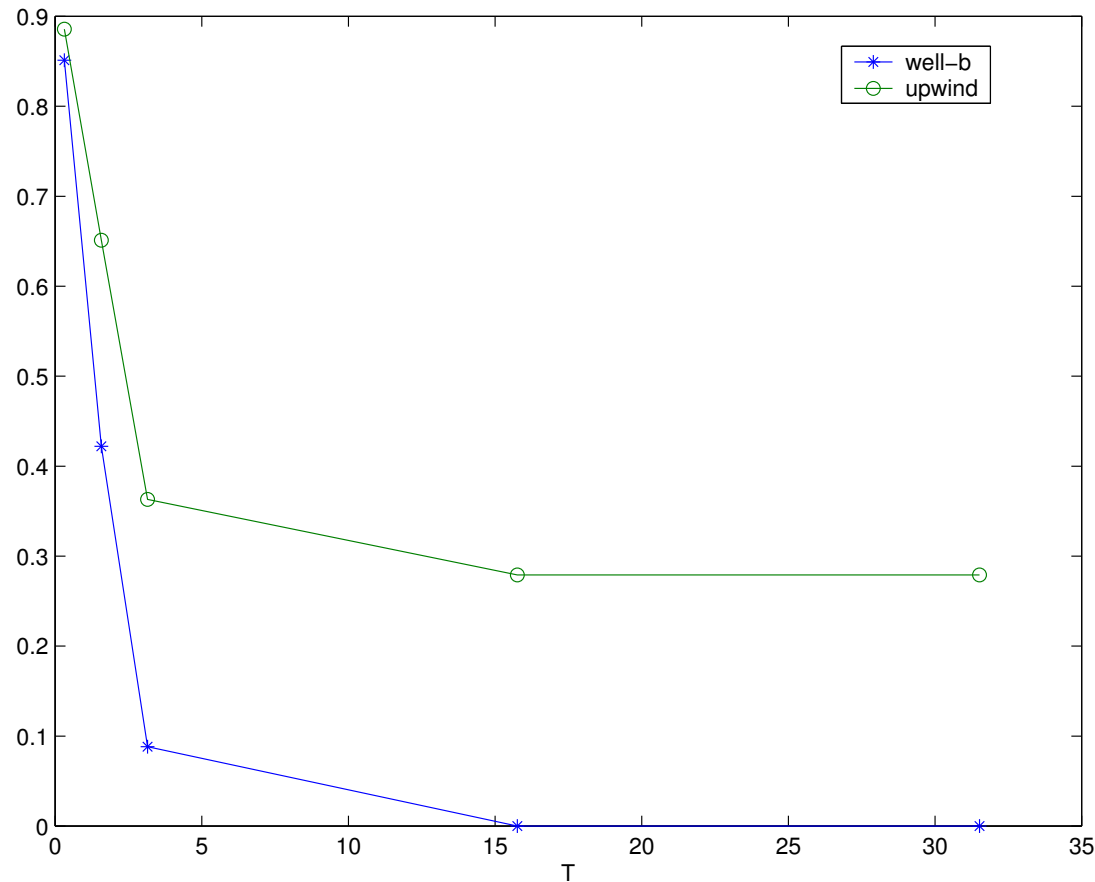


Figure 6:  $l^\infty$  distance between the numerical solution and the asymptotic solution  $w(x)$  for increasing  $T$ .

## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation





## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation

FURTHER DEVELOPMENTS



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation

### FURTHER DEVELOPMENTS

- ▷ Well-balanced for Merton case



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation

### FURTHER DEVELOPMENTS

- ▷ Well-balanced for Merton case
- ▷ More general Lévy models



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation

### FURTHER DEVELOPMENTS

- ▷ Well-balanced for Merton case
- ▷ More general Lévy models
- ▷ American options



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation

### FURTHER DEVELOPMENTS

- ▷ Well-balanced for Merton case
- ▷ More general Lévy models
- ▷ American options
- ▷ Interest rate models with jumps



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation

### FURTHER DEVELOPMENTS

- ▷ Well-balanced for Merton case
- ▷ American options
- ▷ Calibration
- ▷ More general Lévy models
- ▷ Interest rate models with jumps
- ▷ ...



## CONCLUSIONS

---

- Rigorous convergence result for monotone schemes for general Lévy processes
- Fast and accurate finite difference schemes for option pricing in the jump-diffusion model
- A new approach to the space approximation

### FURTHER DEVELOPMENTS

- ▷ Well-balanced for Merton case
- ▷ American options
- ▷ Calibration
- ▷ More general Lévy models
- ▷ Interest rate models with jumps
- ▷ ...

### REFERENCES

- M. Briani, C. La Chioma, R. Natalini; *Numerische Mathematik* - 2004.
- M. Briani, G. Russo, R. Natalini; *Implicit-Explicit numerical scheme for integro-differential parabolic problems etc...* preprint 2004.
- M. Briani, R. Natalini; *Well-balanced schemes for integro-differential equations* (work in progress).

