

# Parametrix approximations in option pricing

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– *Workshop Modelling and numerical techniques in quantitative finance* –

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- ▶ **Ait-Sahalia (2002)**: Hermite polynomials approximation

# Option pricing and fundamental solutions

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$\Gamma =$  **transition density of  $S$  / fundamental solution**

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- ▶  $L_{(T, \xi)}$  is a **heat operator (Black&Scholes)**
- ▶  $\Pi(t, S; T, \xi)$  is a **Gaussian function in  $(t, S)$**

# The backward parametrization

$\Pi(z; \zeta)$  is a Gaussian function **as a function of  $z$**  but

$$\text{Option price} = \int_{\mathbb{R}} \phi(\xi) \Pi(\underbrace{t, S}_z ; \underbrace{T, \xi}_{\zeta}) d\xi$$

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## Our idea:

define a parametrix  $P$  using the **backward (adjoint)** PDE

$$\Gamma(z; \zeta) = P(z; \zeta) + \text{“correction term”}$$

$P(z; \zeta)$  is a **Gaussian function in  $\zeta$**

## The backward parametrix, II

**Second idea**: look for  $\Gamma$  in the form

$$\Gamma(z; \zeta) = P(z; \zeta) + \int_t^T \int_{\mathbb{R}} P(z; \cdot) W(\cdot; \zeta)$$

here  $z = (t, S)$  and  $\zeta = (T, \xi)$



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► being  $L\Gamma = 0$ , the unknown function  $W$  satisfies

$$0 = LP(z; \zeta) - W(z; \zeta) + \int_t^T \int_{\mathbb{R}} LP(z; \cdot) W(\cdot; \zeta)$$

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- ▶ recursive formula

$$W(z; \zeta) = LP(z; \zeta) + \int_t^T \int_{\mathbb{R}} LP(z; \cdot) LP(\cdot; \zeta) + \dots$$

# Global error estimates

$$|\Gamma(t, x) - P_n(t, x)| \leq \delta \frac{t^n}{n!} \Gamma_{\text{heat}}(t, x)$$

$P_n$  = parametrix expansion of order  $n$

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- ▶ rate of convergence independent on dimension
- ▶ asymptotically exact as  $t \rightarrow 0^+$  and  $|x| \rightarrow \infty$

# Option price expansion

$C$  option price with payoff  $\phi$ :

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- ▶ First term: **Black&Scholes price with  $\sigma = \sigma(t, S)$**

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- ▶  $n$ -th term:  
B&S price with  $\sigma = \sigma(t, S)$  and transaction cost  $LC_{n-1}$

$$C_n(t, S) = \int_t^T \int_{\mathbb{R}} LC_{n-1}(\zeta) P(t, S; \zeta) d\zeta$$

## Example 1: explicit 2-parametrix for local volatility

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Call option with strike  $K$ :

$$C(t, S) \simeq C_{\text{BS}}(t, S) + \frac{K(T-t)}{2} (a(T, K) - a(t, S)) P(t, S; T, K)$$

## Numerical test: CEV model

$$\frac{dS_t}{S_t} = rdt + \sigma_0 S_t^{-\alpha} dW_t, \quad \alpha \in ]0, 1[$$

- ▶ Black - Scholes for  $\alpha = 0$

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- ▶ **Cox** (1975), **Schroder** (1989): Gamma-function series
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- ▶ **Hagan** and **Woodward** (1999)
- ▶ Matched Asympt. Expansion (MAE): **Howison** (2005)
- ▶ **Svoboda-Greenwood** (2009)



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Parameters:

- ▶  $\alpha = \frac{1}{4}, \frac{1}{2}, \frac{3}{4}$
- ▶ maturity from 1 week to 1 year
- ▶  $\sigma_0 = 30\%$ ,  $r = 5\%$

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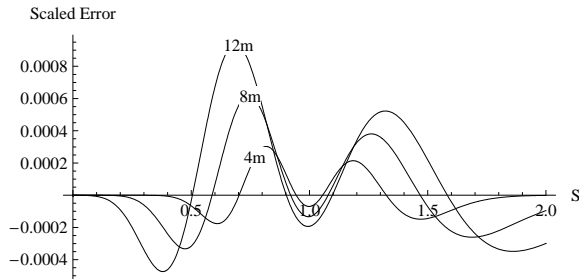
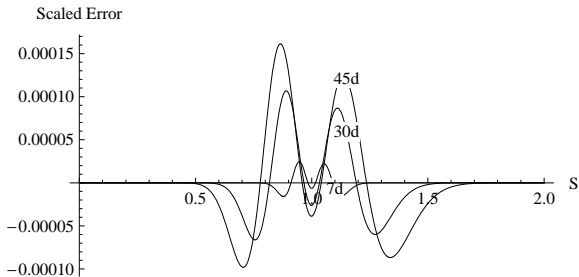
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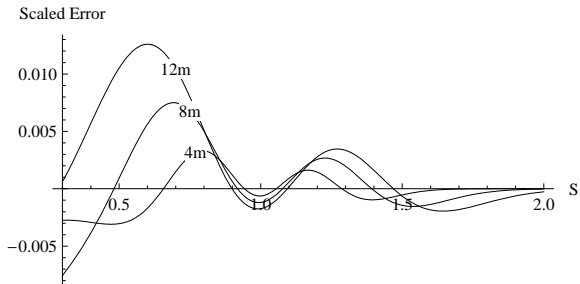
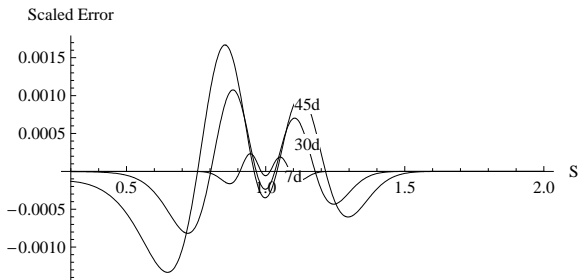
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- ▶  $\sigma_0 = 30\%$ ,  $r = 5\%$
- ▶ **Cox** and **Shaw** approx. are equivalent
- ▶ **Hagan-Woodward** very accurate and *closed-form* solution  $\implies$  used as reference prices

# CEV: parametrix vs Hagan-Woodward, $\alpha = \frac{1}{4}$

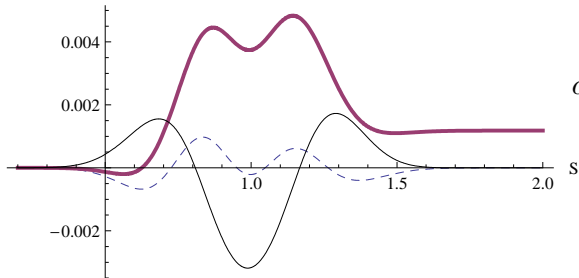


# CEV: parametrix vs Hagan-Woodward, $\alpha = \frac{3}{4}$



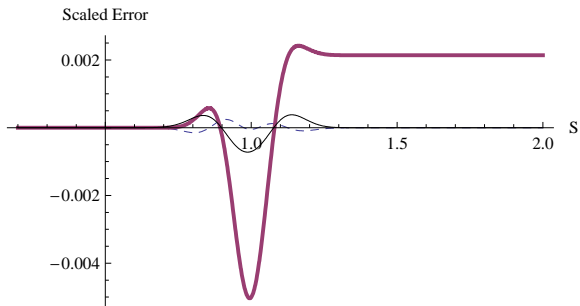
# CEV: param. (dashed), MAE (thick), Svoboda-G. (black)

Scaled Error

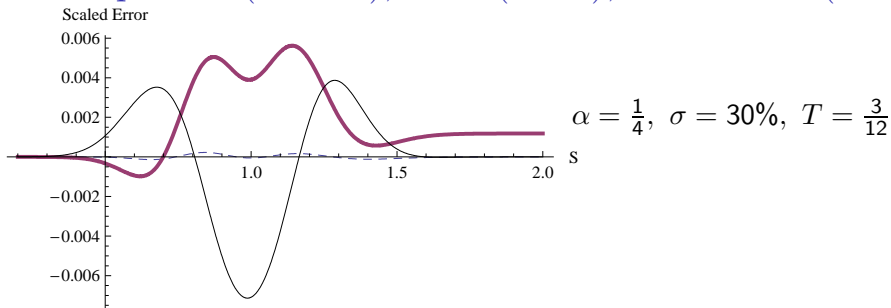


$$\alpha = \frac{1}{2}, \sigma = 30\%, T = \frac{3}{12}$$

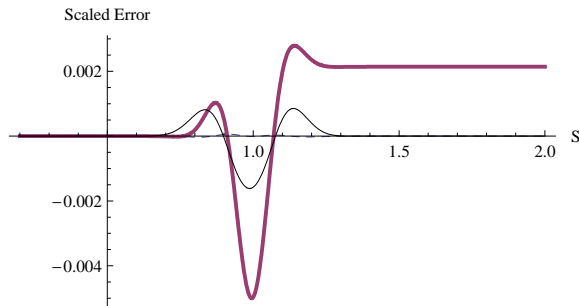
$$\alpha = \frac{1}{2}, \sigma = 15\%, T = \frac{3}{12}$$



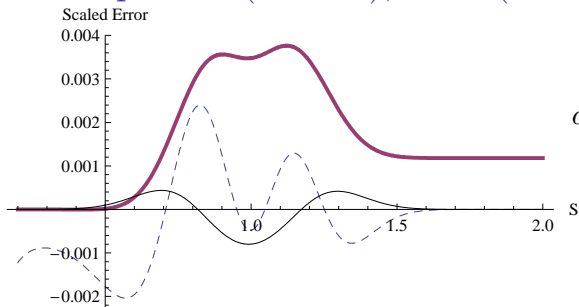
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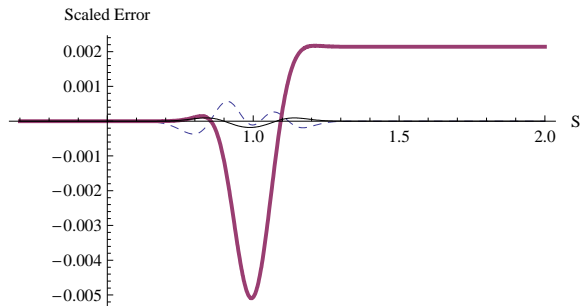


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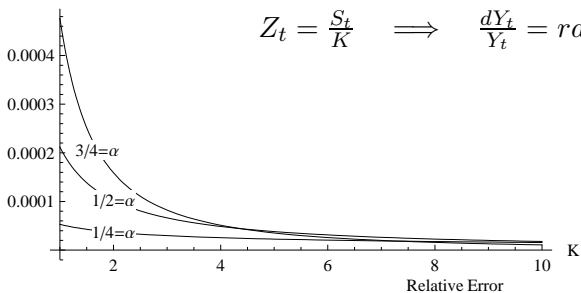
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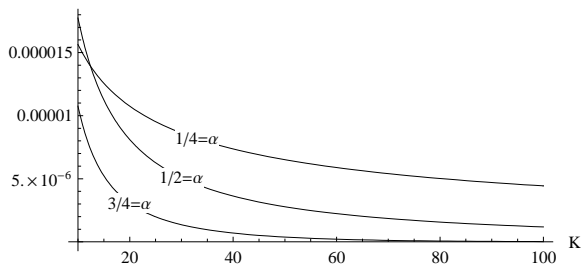


# CEV: strike $K$ large, volatility $\sigma$ small

Relative Error

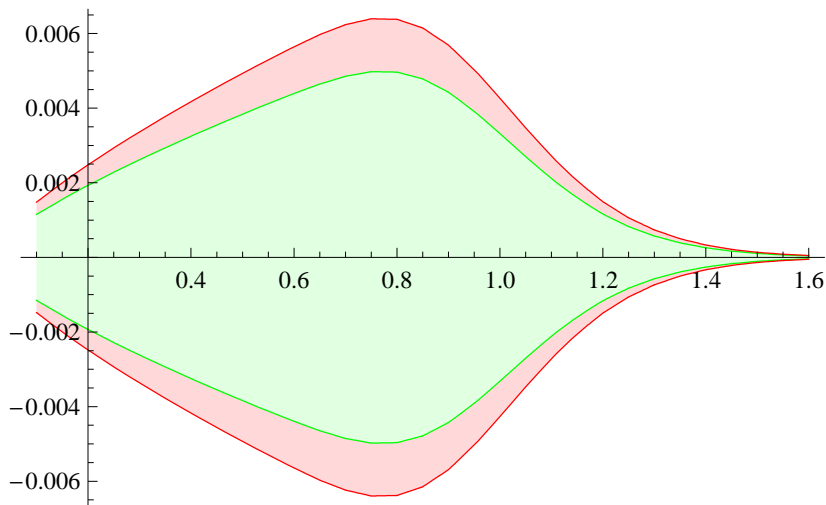


$$Z_t = \frac{S_t}{K} \implies \frac{dY_t}{Y_t} = rdt + \frac{\sigma}{K^\alpha} Y_t^{-\alpha} dW_t$$



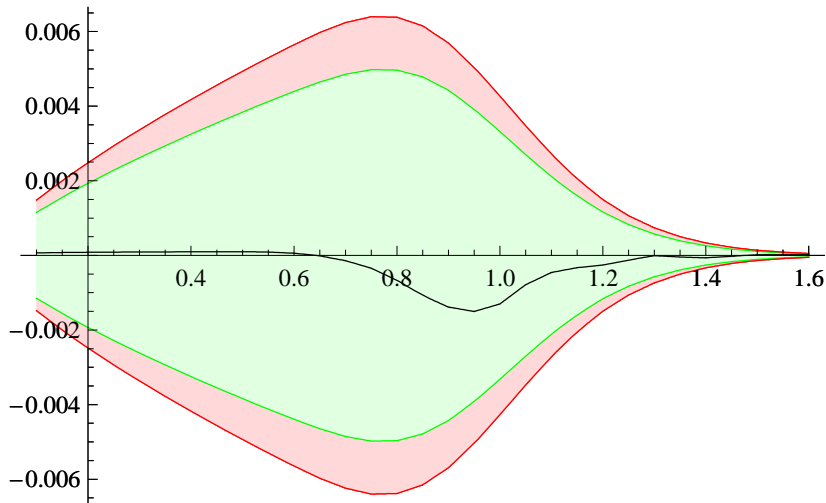
# Monte Carlo

100-Euler, 500.000-MC,  $\alpha = \frac{1}{2}$ ,  $T = \frac{3}{12}$



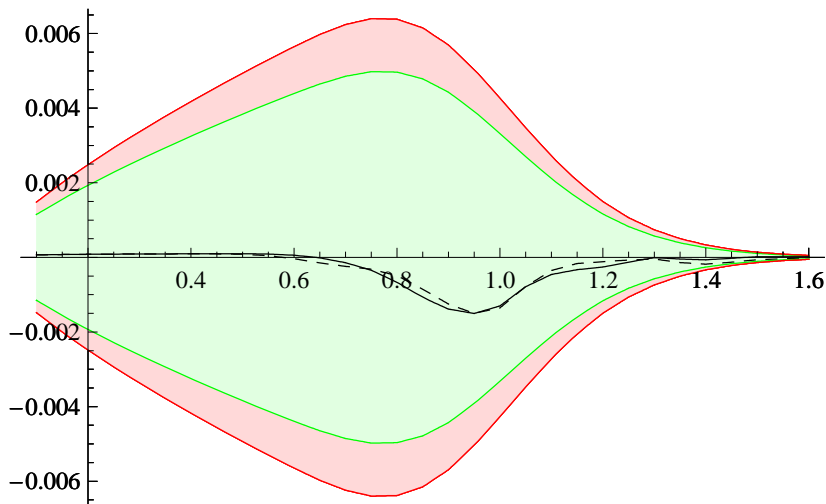
# Monte Carlo, Hagan (black)

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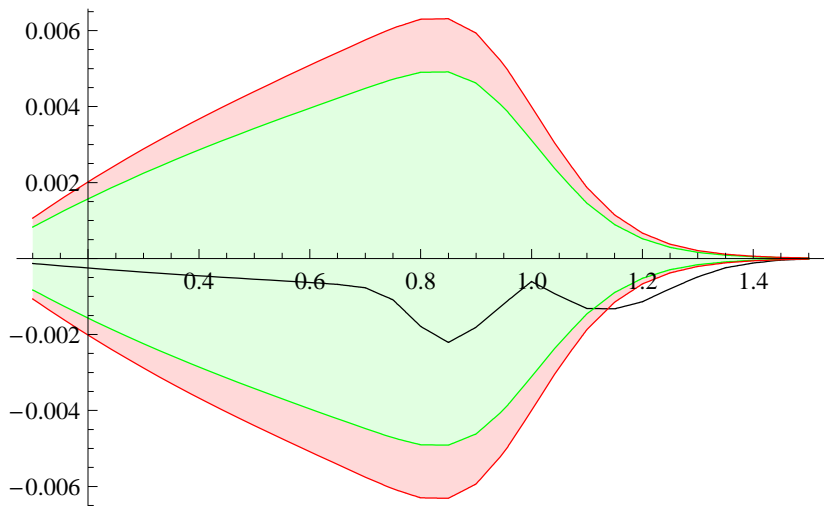
# Monte Carlo, Hagan (black) and parametrix (dashed)

100-Euler, 500.000-MC,  $\alpha = \frac{1}{2}$ ,  $T = \frac{3}{12}$



Quadratic LV:  $\sigma(S) = \sigma_0 \min\{2, \sqrt{1 + (S - 1)^2}\}$

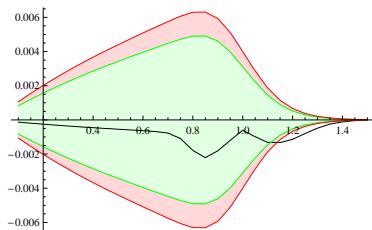
100-Euler, 500.000-MC,  $\sigma_0 = 20\%$ ,  $T = \frac{3}{12}$



## Quadratic LV: MC vs parametrix

	MC	Parametrix
$S = 1$	0.046159	0.046149
$S = 1.1$	0.119926	0.119883
$S = 1.2$	0.213524	0.213482
$S = 1.3$	0.312541	0.312524
$S = 1.4$	0.412432	0.412429
$S = 1.5$	0.512423	0.512423
$S = 1.6$	0.612422	0.612422

- ▶  $\sigma(S) = \sigma_0 \min\{2, \sqrt{1 + (S - 1)^2}\}$
- ▶ strike  $K = 1$ , maturity  $T = \frac{3}{12}$
- ▶ 100-Euler, 500.000-MC



# Path dependent volatility (2-dim)

**Hobson-Rogers** (1998), **Foschi-P.**(2007)

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$Z_t = \log\text{-price}$

$$dZ_t = \mu(D_t)dt + \nu(D_t)dW_t$$

$D_t = \text{deviation from the normal trend}$

$$D_t = Z_t - \int_0^{+\infty} e^{-s} Z_{t-s} ds$$



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**Pricing PDE (degenerate parabolic):**

$$a(x, y, t)\partial_{xx} + x\partial_y + \partial_t, \quad (x, y, t) \in \mathbb{R}^3$$

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**Hobson-Rogers** (1998), **Foschi-P.**(2007)

$Z_t = \log\text{-price}$

$$dZ_t = \mu(D_t)dt + \nu(D_t)dW_t$$

$D_t = \text{deviation from the normal trend}$

$$D_t = Z_t - \int_0^{+\infty} e^{-s} Z_{t-s} ds$$

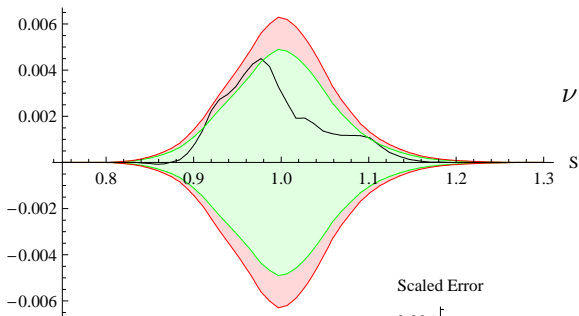
**Pricing PDE (degenerate parabolic):**

$$a(x, y, t)\partial_{xx} + x\partial_y + \partial_t, \quad (x, y, t) \in \mathbb{R}^3$$

- ▶ Complete model - hypoelliptic PDE

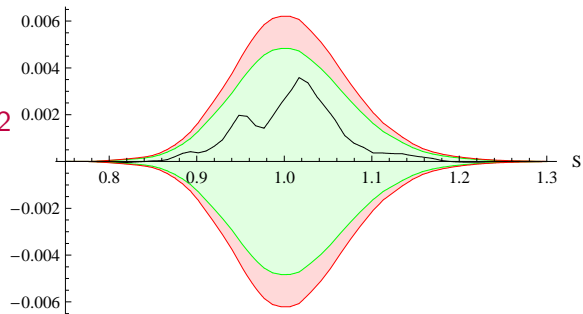
# Path dependent volatility

Scaled Error



$$\nu = 20\% \sqrt{2x^2 + 1}, \quad D_0 = 0$$

Scaled Error



$$\nu = 20\% \sqrt{2x^2 + 1}, \quad D_0 = -0.2$$

# Path dependent volatility

	$D_0 = 0$			$D_0 = -0.2$	
	MC	Parametrix		MC	Parametrix
$S = 0.8$	0.00065	0.00067		0.00061	0.00069
$S = 0.9$	0.00820	0.00819		0.00904	0.00926
$S = 1$	0.04003	0.03991		0.04302	0.04299
$S = 1.1$	0.09982	0.09980		0.10256	0.10250
$S = 1.2$	0.16806	0.16810		0.16929	0.16927
$S = 1.3$	0.23093	0.23094		0.23129	0.23126

# Conclusions

## PROs

- ▶ **simple explicit formulas** for plain vanilla options  $\leftrightarrow$  fast calibration

## CONs

- ▶ difficult to compute higher order terms

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# Conclusions

## PROs

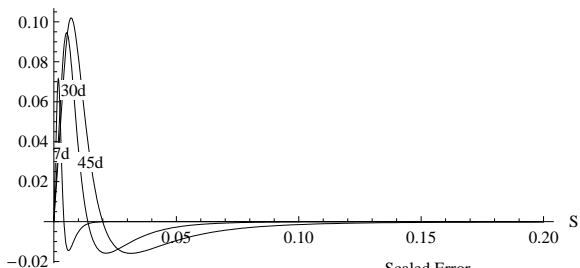
- ▶ **simple explicit formulas** for plain vanilla options  $\leftrightarrow$  fast calibration
- ▶ **rigorous justification** of the method: asymptotically exact as  $t \rightarrow 0$  and  $|x| \rightarrow \infty$
- ▶ expansion using as starting point the Black&Scholes formula

## CONs

- ▶ difficult to compute higher order terms
- ▶ large theoretical error bounds
- ▶ works with uniformly parabolic PDEs  $\leftrightarrow$  Gaussian type densities

CEV:  $\alpha = \frac{3}{4}$ ,  $S \sim \frac{1}{100}$ , strike  $K = 1$

Scaled Error



Scaled Error

