

Marcello Minenna - Paolo Verzella  
Structured Products Italia 2006 - Milan

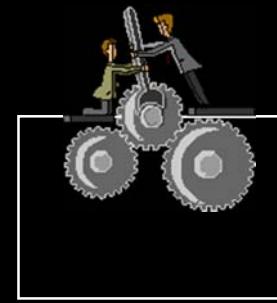


- Unbundling structured products
  - Structured Bonds Italian market
- Review of Option Pricing in Affine Jump Diffusion Models
- Implementation of Fourier Trasform approach
- Implementation of Fast Fourier Trasform approach
  - Rules of Thumb

- **Unbundling structured products**
  - Structured Bonds Italian market
- Review of Option Pricing in Affine Jump Diffusion Models
- Implementation of Fourier Trasform approach
- Implementation of Fast Fourier Trasform approach
  - Rules of Thumb



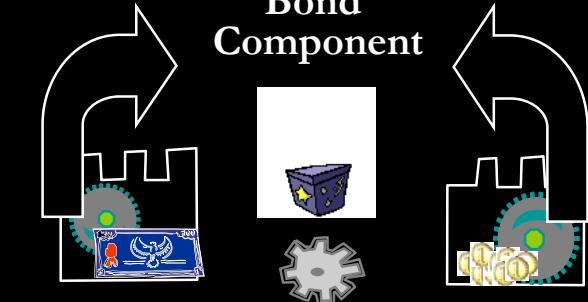
4



5



## Bond Component



6



## Present Value



7



## Derivative Component



8



## Derivative Component



REDUCE

## Bond Component



## Derivative Component



REDUCE

## Bond Component



INCREASE

## Structured Products



10



### Syllabus of the presentation

- Unbundling structured products
  - Structured Bonds Italian market
- Review of Option Pricing in Affine Jump Diffusion Models
- Implementation of Fourier Transform approach
- Implementation of Fast Fourier Transform approach
  - Rules of Thumb

## Derivative Component



INCREASE

## Bond Component



11



13



## Structured Bonds in Italy Primary Market trend 1995 – sept '05

## Derivative Component

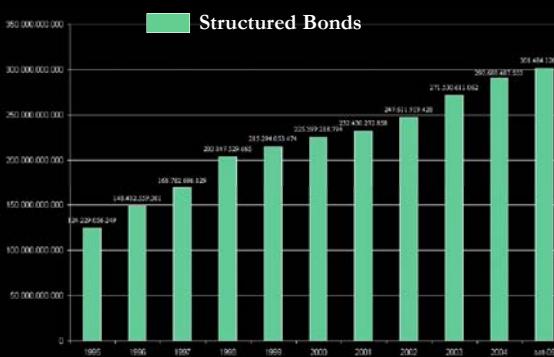


INCREASE

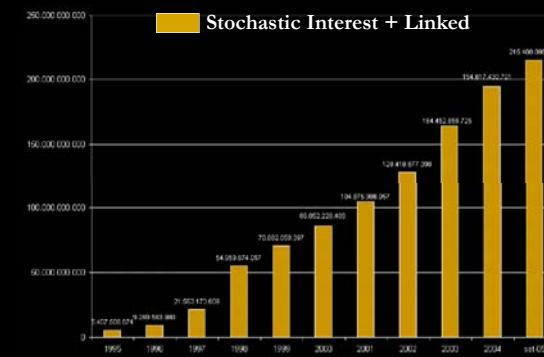
## Bond Component



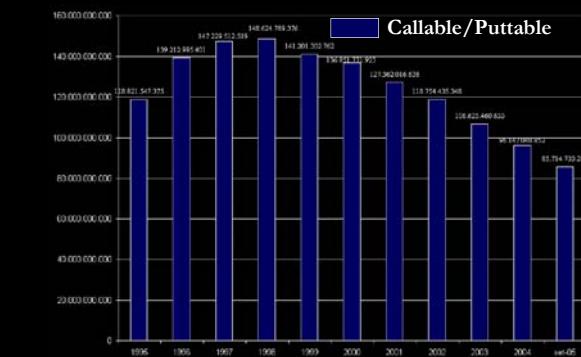
12



16

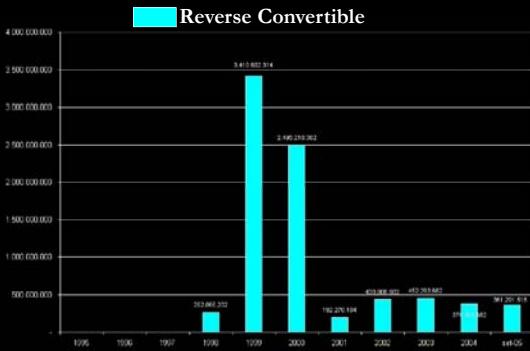


17

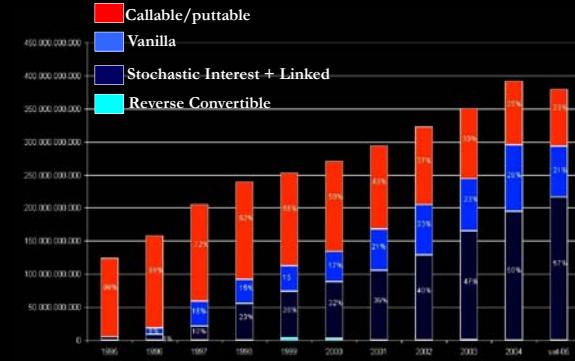


18

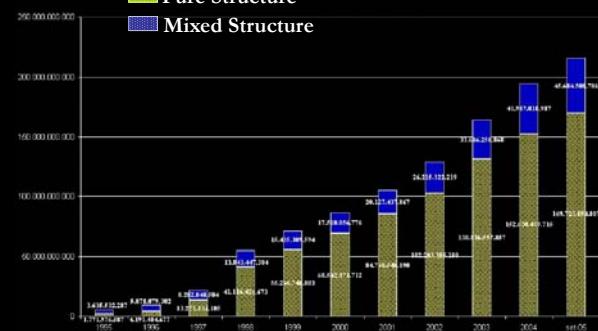




19



20



21



### Syllabus of the presentation

- Unbundling structured products
  - Structured Bonds Italian market
- Review of Option Pricing in Affine Jump Diffusion Models
- Implementation of Fourier Trasform approach
- Implementation of Fast Fourier Trasform approach
  - Rules of Thumb

Minenna, A guide to quantitative Finance, RiskBooks 2006



22



European Call Price  $C_t$   
Spot Price  $S_t$

$P_1(\Theta), P_2(\Theta) = \Pr(\ln S_T \geq \ln K)$   
under different martingale measures



AJD models compute  $C_t$

$$C_t = h[P_1(\Theta), P_2(\Theta)]$$

European Call Price  $C_t$   
Spot Price  $S_t$

$P_1(\Theta), P_2(\Theta) = \Pr(\ln S_T \geq \ln K)$   
under different martingale measures



AJD models compute  $C_t$

$$C_t = h[P_1(\Theta), P_2(\Theta)]$$

$$C_t = u\{g^{-1}[P_2(\Theta, \textcolor{yellow}{A})]\}$$



23



European Call Price  $C_t$   
Spot Price  $S_t$

$f_j(\ln S_t, \xi | \ln S_t) = \int_{-\infty}^{+\infty} e^{i\xi \ln S_t} q_j(\ln S_t | \ln S_t) d \ln S_t$   
under different martingale measures



AJD models compute  $C_t$

$$C_t = h[P_1(\textcolor{yellow}{f}_1), P_2(\textcolor{yellow}{f}_2)]$$

$$C_t = u(\textcolor{yellow}{f}_2)$$



25



26



27



European Call Price  $C_t$

$$f_2(\ln S_T, \xi | \ln S_t) = \int_{-\infty}^{+\infty} e^{i\xi \ln S_T} q(\ln S_T | \ln S_t) d \ln S_T$$

Risk-neutral measure

$$C_t = S_t \left[ \frac{1}{2} + \frac{1}{\pi} \int_0^{+\infty} \Re \left( \frac{e^{i\xi \ln S_T}}{i\xi} f_1 \right) d\xi \right] - K e^{-r(T-t)} \left[ \frac{1}{2} + \frac{1}{\pi} \int_0^{+\infty} \Re \left( \frac{e^{i\xi \ln S_T}}{i\xi} f_2 \right) d\xi \right]$$

PDE

AJD models compute  $C_t$

PDE

$$C_t = h[P_1(\Delta_{1,2} f_2), P_2(f_2)]$$

FT

$$C_t = u(f_2)$$

$$\Delta_{1,2} = \frac{\partial^2}{\partial t^2}$$



28



AJD models compute  $C_t$

PDE

$$C_t = h[P_1(f_1), P_2(f_2)]$$

FT

$$C_t = u(f_2)$$

Numerical evaluation

Newton-Cotes

Hybrid

Gauss



31



#### Review of Fourier Methods in Option Pricing – theory

$$C_t \approx \left( \frac{1}{2} + \frac{1}{\pi} \right) \left[ S_t \sum_{j=0}^N \frac{1}{2} a_j^N \cdot \Re \left( \frac{e^{-i\frac{j}{N} \ln K}}{i(j + \varepsilon \delta_j)} f_1 \left( j \frac{a}{N} \right) \right) - K e^{-r(T-t)} \sum_{j=0}^N \frac{1}{2} a_j^N \cdot \Re \left( \frac{e^{-i\frac{j}{N} \ln K}}{i(j + \varepsilon \delta_j)} f_2 \left( j \frac{a}{N} \right) \right) \right]$$

PDE

Newton-Cotes schemes compute  $C_t$   
Trapezoid rule



34



AJD models compute  $C_t$

AJD models compute  $C_t$

FT

$$C_t = \frac{e^{-\alpha \ln K}}{\pi} \int_0^{+\infty} \Re \left( e^{i\xi \ln K} \frac{e^{-rt} f_2(\xi - (\alpha+1)i)}{\alpha^2 + \alpha - \xi^2 + i(2\alpha+1)\xi} \right) d\xi$$



29



AJD models compute  $C_t$

PDE

$$C_t = h[P_1(f_1), P_2(f_2)]$$

FT

$$C_t = u(f_2)$$

Numerical evaluation

Newton-Cotes

Hybrid

Gauss



Newton-Cotes schemes compute  $C_t$   
Trapezoid rule



32



#### Review of Fourier Methods in Option Pricing – theory

$$C_t \approx \left( \frac{1}{2} + \frac{1}{\pi} \right) \left[ S_t \sum_{j=0}^N \frac{1}{2} a_j^N \cdot \Re \left( \frac{e^{-i\frac{j}{N} \ln K}}{i(j + \varepsilon \delta_j)} f_1 \left( j \frac{a}{N} \right) \right) - K e^{-r(T-t)} \sum_{j=0}^N \frac{1}{2} a_j^N \cdot \Re \left( \frac{e^{-i\frac{j}{N} \ln K}}{i(j + \varepsilon \delta_j)} f_2 \left( j \frac{a}{N} \right) \right) \right]$$

FT

Newton-Cotes schemes compute  $C_t$   
Trapezoid rule



33



#### Review of Fourier Methods in Option Pricing – theory

$$C_t \approx \frac{e^{-\alpha \ln K}}{2\pi} \frac{a}{N} \Re \left[ \sum_{j=0}^N e^{-ijh \ln K} f_2 \left( j \frac{a}{N} \right) \right]$$



35



Newton-Cotes schemes compute  $C_t$   
Simpson rule



36



$$C_t \approx \left( \frac{1}{2} + \frac{1}{\pi} \right) \left[ S \sum_{j=0}^N [3 + (-1)^{j+1} - \delta_j] \cdot \Re \left[ \frac{e^{-ij\frac{a}{N}\ln K}}{i(j+\varepsilon\delta_j)} f_1 \left( j \frac{a}{N} \right) \right] - \right.$$

$$\left. - Ke^{-r(T-t)} \sum_{j=0}^N [3 + (-1)^{j+1} - \delta_j] \cdot \Re \left[ \frac{e^{-ij\frac{a}{N}\ln K}}{i(j+\varepsilon\delta_j)} f_2 \left( j \frac{a}{N} \right) \right] \right]$$

**PDE**

**Newton-Cotes schemes compute  $C_t$**   
**Simpson rule**



37



$$C_t \approx S_i \left( \frac{1}{2} + \frac{1}{\pi} \right) a \left[ \frac{1}{N(N-1)} \Re \left( \frac{e^{-ic\ln K}}{i\varepsilon} f_1(\varepsilon) \right) + \Re \left( \frac{e^{-ia\ln K}}{ia} f_1(a) \right) \right] +$$

$$+ \sum_{j=2}^{N-1} \frac{1}{N(N-1)[P_{N-1}(\xi_j)]^2} \cdot \Re \left[ \frac{e^{-i[\frac{1}{2}a(1+\xi_j)]\ln K}}{i(\frac{1}{2}a(1+\xi_j))} f_1 \left( \frac{1}{2}a(1+\xi_j) \right) \right] -$$

$$- Ke^{-r(T-t)} \left( \frac{1}{2} + \frac{1}{\pi} \right) a \left[ \frac{1}{N(N-1)} \Re \left( \frac{e^{-ic\ln K}}{i\varepsilon} f_2(\varepsilon) \right) + \Re \left( \frac{e^{-ia\ln K}}{ia} f_2(a) \right) \right] +$$

$$+ \sum_{j=2}^{N-1} \frac{1}{N(N-1)[P_{N-1}(\xi_j)]^2} \cdot \Re \left[ \frac{e^{-i[\frac{1}{2}a(1+\xi_j)]\ln K}}{i(\frac{1}{2}a(1+\xi_j))} f_2 \left( \frac{1}{2}a(1+\xi_j) \right) \right]$$

**PDE**  
**Gauss schemes compute  $C_t$**   
**Gauss-Lobatto rule**



40



$$C_t \approx -S_i \left( \frac{1}{2N} + \frac{1}{N\pi} \right) \sum_{j=1}^N \frac{1}{L_{N-1}(\xi_j) L'_N(\xi_j)} \Re \left( \frac{e^{-\xi_j(i\ln K-1)}}{i\xi_j} f_1(\xi_j) \right) +$$

$$+ Ke^{-r(T-t)} \left( \frac{1}{2N} + \frac{1}{N\pi} \right) \sum_{j=1}^N \frac{1}{L_{N-1}(\xi_j) L'_N(\xi_j)} \Re \left( \frac{e^{-\xi_j(i\ln K-1)}}{i\xi_j} f_2(\xi_j) \right)$$

**PDE**

**Gauss schemes compute  $C_t$**   
**Gauss-Laguerre rule**



43



**Gauss schemes compute  $C_t$**   
**Gauss-Lobatto rule**



41



**FT**  
**Gauss schemes compute  $C_t$**   
**Gauss-Laguerre rule**

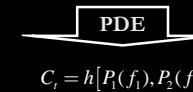
$$C_t \approx -\frac{e^{-\alpha\ln K}}{N\pi} \sum_{j=1}^N \frac{1}{L_{N-1}(\xi_j) L'_{N-1}(\xi_j)} \cdot \Re \left( e^{-\xi_j(i\ln K-1)} f_2(\xi_j) \right)$$



44



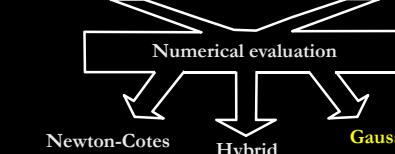
**AJD models compute  $C_t$**



$$C_t = h[P_1(f_1), P_2(f_2)]$$



$$C_t = u(f_2)$$



**Newton-Cotes**

**Hybrid**

**Gauss**

39

**Gauss schemes compute  $C_t$**   
**Gauss-Lobatto rule**

**FT**

$$C_t \approx \frac{e^{-\alpha\ln K}}{\pi} a \left[ \frac{1}{N(N-1)} \Re(f_2(0)) + \Re(e^{-ia\ln K} f_2(a)) + \right.$$

$$\left. + \sum_{j=2}^{N-1} \frac{1}{N(N-1)[P_{N-1}(\xi_j)]^2} \cdot \Re \left( e^{-i[\frac{1}{2}a(1+\xi_j)]\ln K} f_2 \left( \frac{1}{2}a(1+\xi_j) \right) \right) \right]$$

42

- Unbundling structured products
  - Structured Bonds Italian market
- Review of Option Pricing in Affine Jump Diffusion Models
- Implementation of Fourier Trasform approach
- Implementation of Fast Fourier Trasform approach
  - Rules of Thumb

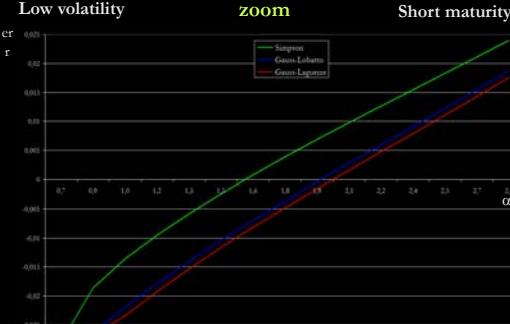
45

$$C_t \text{ via FT}$$

$$C_t = u\{g^{-1}[P_2(\Theta, \textcolor{blue}{a})]\}$$

calibration of  $\alpha$ 

46

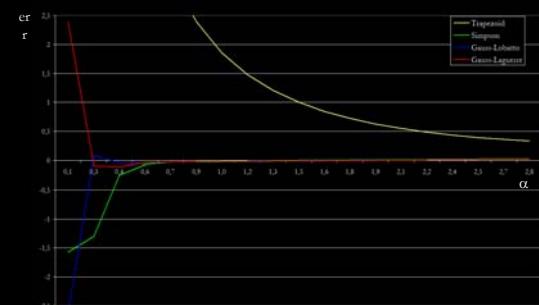

 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models


49


 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models

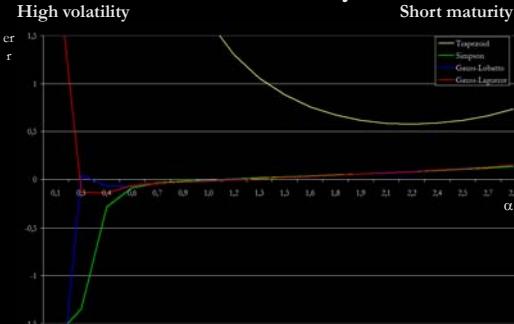
Low volatility

Short maturity

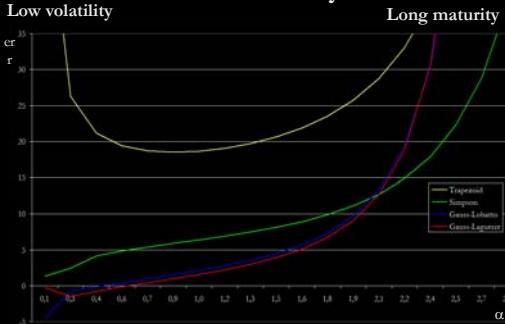


48


 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models

 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models


50


 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models


52



by minimizing

 $C_t \text{ via PDE}$ 

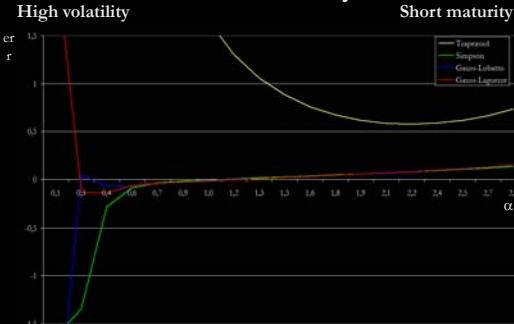
$$C_t = h[P_1(\Theta), P_2(\Theta)]$$

 $C_t \text{ via FT}$ 

$$C_t = u\{g^{-1}[P_2(\Theta, \textcolor{blue}{a})]\}$$

calibration of  $\alpha$ 

47

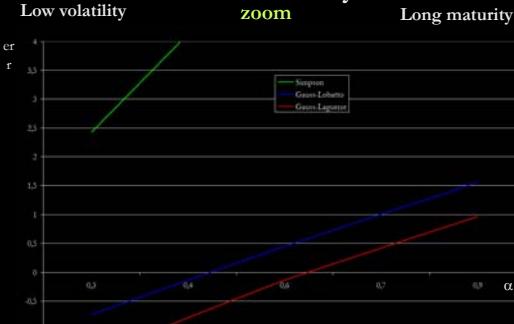

 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models


51

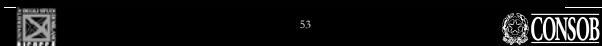


48


 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models

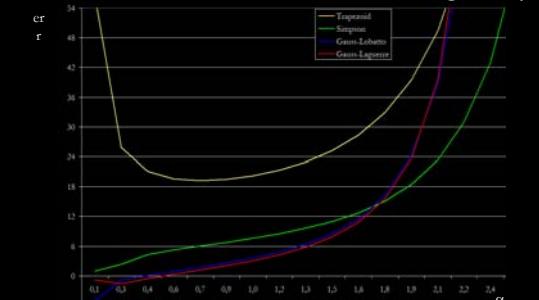
 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models


53


 $C_t \text{ via FT - spanning } \Theta, \alpha$   
 Stochastic volatility models

High volatility

Long maturity

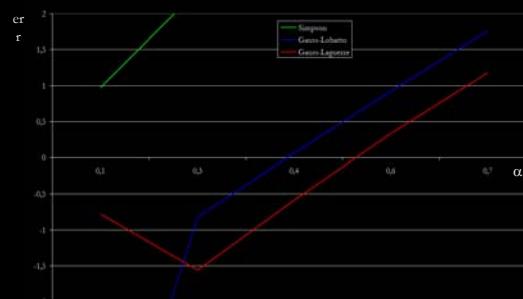


54



$C_t$  via FT - spanning  $\Theta, \alpha$   
Stochastic volatility models

High volatility zoom Long maturity

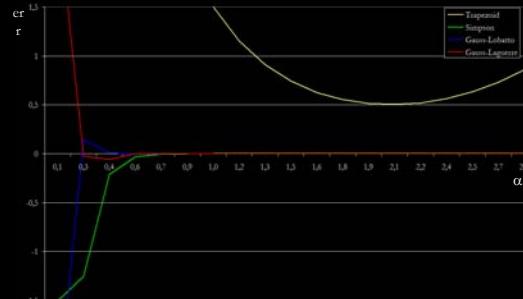


55



$C_t$  via FT - spanning  $\Theta, \alpha$   
Jump Diffusion models

High volatility zoom Short maturity

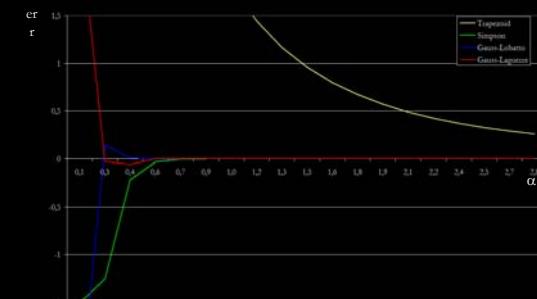


58



$C_t$  via FT - spanning  $\Theta, \alpha$   
Jump Diffusion models

Low volatility zoom Short maturity

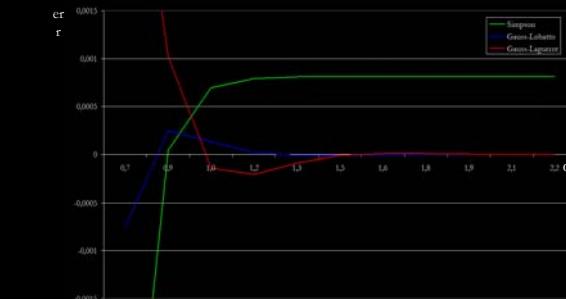


56



$C_t$  via FT - spanning  $\Theta, \alpha$   
Jump Diffusion models

Low volatility zoom Short maturity

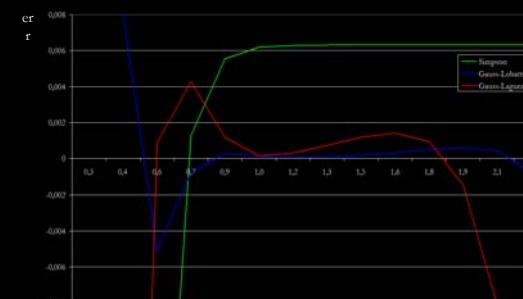


57



$C_t$  via FT - spanning  $\Theta, \alpha$   
Jump Diffusion models

Low volatility zoom Long maturity

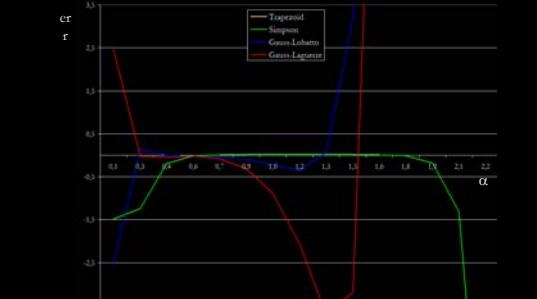


61



$C_t$  via FT - spanning  $\Theta, \alpha$   
Jump Diffusion models

High volatility zoom Long maturity

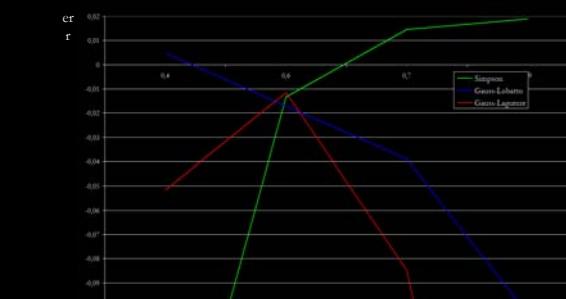


62



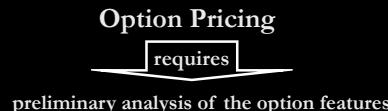
$C_t$  via FT - spanning  $\Theta, \alpha$   
Jump Diffusion models

High volatility zoom Long maturity

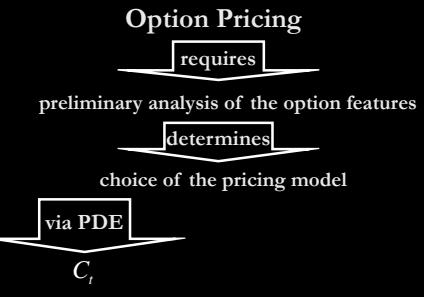


63



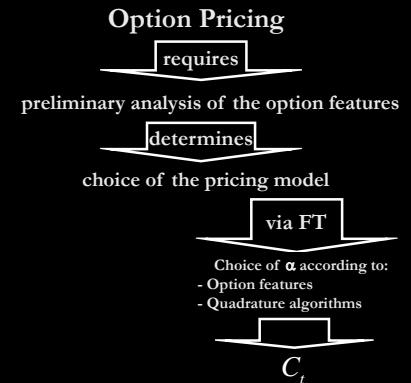


preliminary analysis of the option features



choice of the pricing model

$C_t$



Choice of  $\alpha$  according to:  
- Option features  
- Quadrature algorithms

$C_t$



64



65

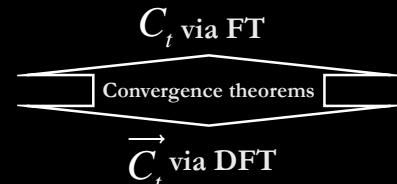


66

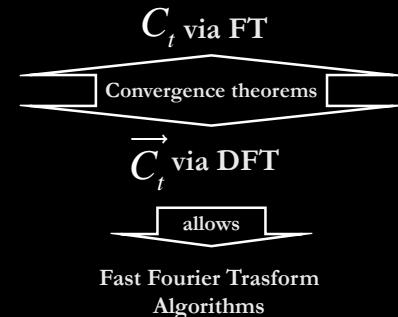


#### Syllabus of the presentation

- Unbundling structured products
  - Structured Bonds Italian market
- Review of Option Pricing in Affine Jump Diffusion Models
- Implementation of Fourier Trasform approach
- **Implementation of Fast Fourier Trasform approach**
  - Rules of Thumb



$\vec{C}_t$  via DFT



Fast Fourier Trasform  
Algorithms



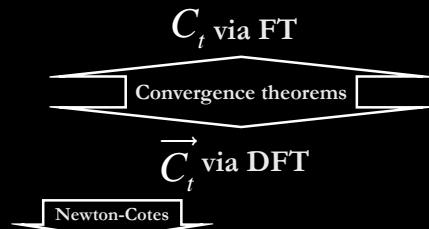
67



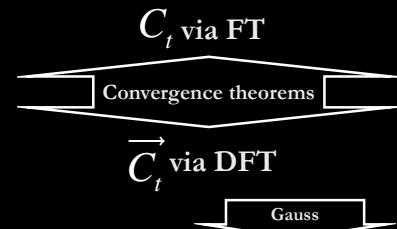
68



69



Cooley-Tuckey FFT



Fessler-Sutton  
NonUniform FFT

$$C_0 \left( [\ln K]_u \right) \approx \frac{e^{-\alpha \left( \ln S_t - \frac{N\pi}{a} + \frac{2\pi}{a}(u-1) \right)}}{2\pi} \frac{a}{3N} \omega^*(u)$$

C-T FFT

Newton-Cotes Convergence Theorem  
characterization compute  $\vec{C}_t$   
Simpson



70

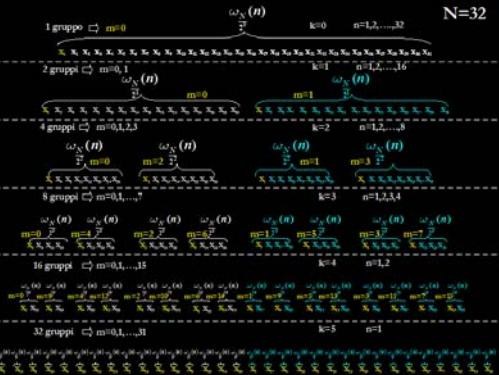


71



72





73



## Syllabus of the presentation

- Unbundling structured products
  - Structured Bonds Italian market
- Review of Option Pricing in Affine Jump Diffusion Models
- Implementation of Fourier Trasform approach
- Implementation of Fast Fourier Trasform approach
  - Rules of Thumb



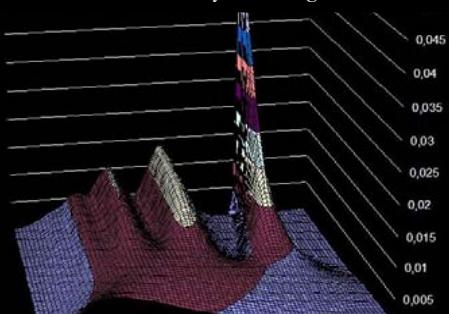
76



S u l f l q j # v u x f w u h g # u r g x f w t #

Fourier Transform vs Discrete Fourier Trasform

From Theory to Trading Desk



Marcello Minenna - Paolo Verzella  
Structured Products Italia 2006 - Milan



Gauss Convergence Theorem

characterization compute  $\bar{C}_t$

**Gauss-Laguerre**

F-S NU FFT

$$C_0 \left( \left[ \ln K \right]_u \right) \approx \frac{e^{-\alpha \left( \ln S_t - \frac{N\pi}{a} + \frac{2\pi}{a}(u-1) \right)}}{2\pi} \omega(v_u)$$

J. Fessler and B. P. Sutton, Non-uniform fast Fourier transforms using min-max interpolation.  
Signal Processing, 2001

$C_t$  via FT

Convergence theorems

$C_t$  via DFT

determines

Choice of  $\alpha$  according to:

- Option features
- Quadrature algorithms



74



$C_t$  via FT - spanning  $\Theta, \alpha$   
Stochastic volatility models

VOLATILITY

		LOW	HIGH
MATURITY	SHORT	Simpson [ $\alpha \approx (1,5;1,6)$ ] Gauss-Lobatto [ $\alpha \approx (1,9;2)$ ] Gauss-Laguerre [ $\alpha \approx (1,9;2)$ ]	Simpson [ $\alpha \approx (1,05;1,1)$ ] Gauss-Lobatto [ $\alpha \approx (1,15;1,2)$ ] Gauss-Laguerre [ $\alpha \approx (1,15;1,2)$ ]
	LONG	Gauss-Lobatto [ $\alpha \approx (0,35;0,45)$ ] Gauss-Laguerre [ $\alpha \approx (0,6;0,65)$ ]	Gauss-Lobatto [ $\alpha \approx (0,4;0,45)$ ] Gauss-Laguerre [ $\alpha \approx (0,5;0,55)$ ]



77



$C_t$  via FT - spanning  $\Theta, \alpha$   
Jump diffusion models

VOLATILITY

		LOW	HIGH
MATURITY	SHORT	Gauss-Lobatto [ $\alpha \approx (1,4;2)$ ] Gauss-Laguerre [ $\alpha \approx (1,55;2)$ ]	Gauss-Lobatto [ $\alpha \approx (1,8;2,2)$ ] Gauss-Laguerre [ $\alpha \approx (1,8;2,2)$ ]
	LONG	Gauss-Lobatto [ $\alpha \approx (0,9;1,2)$ ] Gauss-Laguerre [ $\alpha \approx (0,95;1,05)$ ]	Simpson [ $\alpha \approx (0,6;0,7)$ ] Gauss-Lobatto [ $\alpha \approx (0,35;0,45)$ ] Gauss-Laguerre [ $\alpha \approx (0,6)$ ]



78

