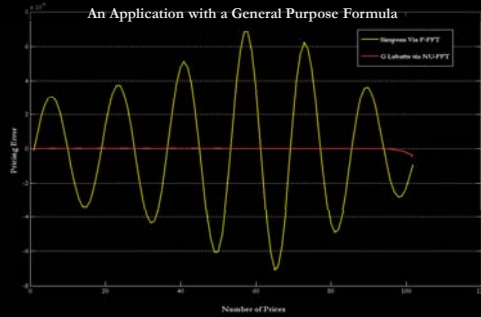


# Fractional vs Non Uniform Discrete Fourier Transform



## Syllabus of the presentation

- Review of Option Pricing via DFT
  - The Lewis Standard Machine
  - DFT Convergence to FT
  - Convergence Theorems for Uniform Grids
  - Convergence Theorems for Non Uniform Gaussian Grids
- Fast Option Pricing
  - Fractional FFT
  - Non Uniform FFT
    - Gaussian Gridding: a matter of interpolation
  - Fractional vs Non Uniform FFT: Empirical Analysis
  - Conclusions

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### The Lewis Standard Machine

$$\tilde{w}(z) = \int_{-\infty}^{+\infty} e^{-izx} w(x) dx$$

is the PayOff functional's Transform

$$\phi_T(z) = E^Q[e^{iz \ln S_T}]$$

under risk-neutral measure



A linear direct mapping from Fourier Spectral Space



### LEWIS REPRESENTATION

### The Lewis Standard Machine

$$\tilde{w}(z) = \int_{-\infty}^{+\infty} e^{-izx} w(x) dx$$

is the PayOff functional's Transform

$$\phi_T(z) = E^Q[e^{iz \ln S_T}]$$

under risk-neutral measure



A linear direct mapping from Fourier Spectral Space



$$V_t = \frac{e^{-r(T-t)}}{2\pi} \int_{i\alpha-\infty}^{i\alpha+\infty} e^{-iz \ln K} \phi_T(-z) \int_{-\infty}^{+\infty} e^{-izx} w(x) dx dz$$

### The Lewis Standard Machine

Knowing  $\tilde{w}(z)$



implies reducing the problem to the calculation of a single integral

### The Lewis Standard Machine

$$z = \xi + i\alpha$$

Financial Claim	$w(x)$	$\tilde{w}(z)$
Call Option	$\max[S_T - K, 0]$	$\frac{K^{-\alpha+1}}{z^\alpha - K}$ , $\alpha > 1$
Put Option	$\max[K - S_T, 0]$	$\frac{K^{-\alpha+1}}{z^\alpha - K}$ , $\alpha < 0$
Covered Call	$\min[S_T, K]$	$\frac{K^{-\alpha+1}}{z^\alpha - K}$ , $0 < \alpha < 1$
Money Market	1	$2\pi\delta(k)$ , $\alpha \in \mathbb{R}$
Self Quanto Call	$\max[S_T - K, 0] \cdot S_T$	$\frac{K^{-\alpha+2}}{(z+1)^{\alpha-1} (z+2)^{\alpha-1}}$ , $\alpha < -2$
Power Call	$\max[S_T - K, 0]^d$	$\frac{K^{-\alpha+d} \Gamma(\alpha) \Gamma(d+1)}{\Gamma(\alpha+d+1)}$ , $\alpha < -d$

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## DFT Convergence to FT

Given the General DFT



$$\omega(m) = \sum_{j=0}^{N-1} e^{-i \frac{2\pi}{N} x_j (m-1)} f(x_j) \quad \text{where } m = 1, 2, \dots, M$$

Given the General DFT

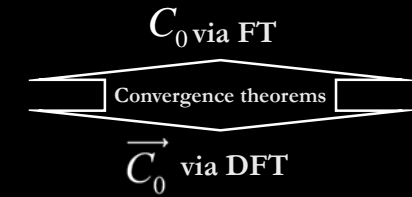
$$\omega(m) = \sum_{j=0}^{N-1} e^{-i \frac{2\pi}{X} x_j (m-1)} f(x_j) \quad \text{where } m = 1, 2, \dots, M$$

$M \neq N$

The Convergence Theorem for General DFT's (C Th)

$$\mathcal{F}[f(x)](t_m) = \lim_{N \rightarrow \infty} \sum_{j=1}^N e^{-i \frac{2\pi}{X} x_j (m-1)} f(x_j, X)$$

$$t_m = \frac{2\pi}{X} (m-1)$$



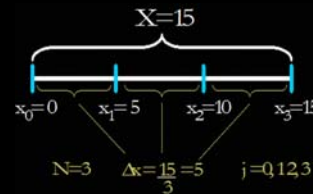
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Convergence Theorems for Uniform Grids

Condition 1

Uniform Discretization Grid



Convergence Theorems for Uniform Grids

Condition 2

$N=M$

DFT specialized

$$\omega(n) = \sum_{j=0}^{N-1} e^{-i \frac{2\pi}{X} x_j (n-1)} f(x_j) \quad \text{where } n=1, 2, \dots, N$$

Convergence Theorems for Uniform Grids

Condition 1

Condition 2



DFT Simplified Formula

$$\omega(n) = \sum_{j=0}^{N-1} e^{-i 2\pi k j \gamma} f(x_j) \quad \text{where } n = 1 \dots N$$

Convergence Theorems for Uniform Grids

Nyquist – Shannon Limit (N-S)

$$\mathcal{F}[f(x)](t_n) = \lim_{N \rightarrow \infty} \frac{X}{N} \omega(n)$$

$$\{t_n\}_{n=1 \dots \frac{N}{2}} \quad \text{for } N \text{ even}$$

$$\{t_n\}_{n=1 \dots \frac{N+1}{2}} \quad \text{for } N \text{ odd}$$

Convergence Theorems for Uniform Grids

$$C_t = \frac{K e^{-\gamma(T-t)}}{2\pi} \int_{i\alpha-\epsilon}^{i\alpha+\epsilon} e^{-iz \ln K} \frac{\phi_T(-z)}{z^2 - iz} dz$$

Uniform Discretization Grids for  $\phi_T$

1.  $\phi_T(\xi_{j-1}) = e^{-i(j-1)\eta[\ln S_t - b]} \Psi_0[(j-1)\eta]$
2.  $\phi_T(\xi_{j-1}) = e^{-i(j-1)\eta[\ln S_t - b]} \Psi_0[(j-1)\eta] \cdot [3 + (-1)^{j+1} - \delta_j - \delta_{N-j}]$

1.

$$C_f = \frac{Ke^{-r(T-t)}}{2\pi} \int_{i\alpha-\epsilon}^{i\alpha+\epsilon} e^{-iz \ln K} \frac{\phi_T(-z)}{z^2 - iz} dz$$

$$\phi_T(\xi_{j-1}) = e^{-i(j-1)\eta[\ln S_T - b]} \Psi_0[(j-1)\eta]$$



$$C_0[\ln K]_u^- \approx \frac{e^{-\alpha[\ln S_T - b + \lambda(u-1)]}}{b} \cdot \Re(\omega(u))$$



2.

$$C_f = \frac{Ke^{-r(T-t)}}{2\pi} \int_{i\alpha-\epsilon}^{i\alpha+\epsilon} e^{-iz \ln K} \frac{\phi_T(-z)}{z^2 - iz} dz$$

$$\phi_T(\xi_{j-1}) = e^{-i(j-1)\eta[\ln S_T - b]} \Psi_0[(j-1)\eta] \cdot [3 + (-1)^{j+1} - \delta_j - \delta_{N-j}]$$



$$C_0[\ln K]_u^- \approx \frac{e^{-\alpha[\ln S_T - b + \lambda(u-1)]}}{3b} \cdot \Re(\omega(u))$$



**Theorems of Equivalence**



The Call Price computed via Convergence Theorem is equal to the Call Price computed via Trapezoid/Simpson Quadrature Rule



Syllabus of the presentation

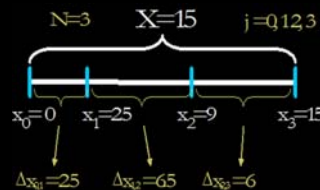
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Convergence Theorems for Non Uniform Gaussian Grids

Condition 1

**Non Uniform Discretization Grid**



Convergence Theorems for Non Uniform Gaussian Grids

Condition 1

**Gaussian Grids**



Optimal choice of discretization points



Gauss Laguerre



Gander Gautschi



Convergence Theorems for Non Uniform Gaussian Grids

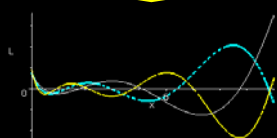
Condition 1

**Gaussian Grids**



Optimal choice of discretization points

**Gauss Laguerre**



Zeros of Laguerre Polynomials



Convergence Theorems for Non Uniform Gaussian Grids

Condition 1

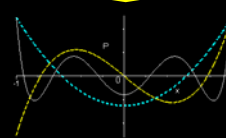
**Gaussian Grids**



Optimal choice of discretization points

**Gauss Lobatto**

Zeros of Legendre Polynomials



Convergence Theorems for Non Uniform Gaussian Grids

Condition 1

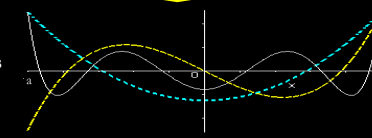
**Gaussian Grids**



Optimal choice of discretization points

**Gander Gautschi**

Zeros of rescaled Legendre Polynomials



Condition 2

**N≠M**



**General DFT**

$$\omega(m) = \sum_{j=0}^{N-1} e^{-i \frac{2\pi}{X} x_j (m-1)} f(x_j) \text{ where } m=1,2,\dots,2M$$



**The Convergence Theorem for General DFT's (C Th)**



$$\mathcal{F}[f(x)](t_m) = \lim_{N \rightarrow \infty} \sum_{j=1}^N e^{-i \frac{2\pi}{X} x_j (m-1)} f(x_j, X)$$

$$t_m = \frac{2\pi}{X} (m-1)$$



$$C_t = \frac{K e^{-r(T-t)}}{2\pi} \int_{i\alpha-\epsilon}^{i\alpha+\epsilon} e^{-iz \ln K} \frac{\phi_T(-z)}{z^2 - iz} dz$$



**Gaussian Grids for  $\phi_T$**

- $\phi_T(\xi_{j-1}) = e^{\left[1 + i \left(\frac{M\pi}{a} \ln S_t\right)\right] \xi_{j-1}} \Psi_0[\xi_{j-1}] \cdot \frac{1}{L_{N+1}(\xi_{j-1}) L'_N(\xi_{j-1})}$
- $\phi_T\left(\frac{1}{2}a(1 + \xi_{j-1})\right) = e^{\left[-i \left(\frac{1}{2}a(1 + \xi_{j-1})\right) \left(\ln S_t - \frac{M\pi}{a}\right)\right] \xi_{j-1}} \Psi_0\left[\frac{1}{2}a(1 + \xi_{j-1})\right] \cdot \frac{1}{\left[P_{N-1}(\xi_{j-1})\right]^2}$



1.

$$C_t = \frac{K e^{-r(T-t)}}{2\pi} \int_{i\alpha-\epsilon}^{i\alpha+\epsilon} e^{-iz \ln K} \frac{\phi_T(-z)}{z^2 - iz} dz$$

$$\phi_T(\xi_{j-1}) = e^{\left[1 + i \left(\frac{M\pi}{a} \ln S_t\right)\right] \xi_{j-1}} \Psi_0[\xi_{j-1}] \cdot \frac{1}{L_{N+1}(\xi_{j-1}) L'_N(\xi_{j-1})}$$



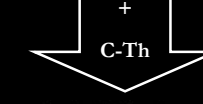
$$C_0([\ln K]_u^*) \approx -\Re \left[ \frac{e^{-a \left(\ln S_t - \frac{M\pi}{a} + \frac{2\pi}{a} (v-1)\right)}}{\pi} \cdot \frac{1}{N+1} \cdot \omega^*(u) \right]$$



2.

$$C_t = \frac{K e^{-r(T-t)}}{2\pi} \int_{i\alpha-\epsilon}^{i\alpha+\epsilon} e^{-iz \ln K} \frac{\phi_T(-z)}{z^2 - iz} dz$$

$$\phi_T\left(\frac{1}{2}a(1 + \xi_{j-1})\right) = e^{\left[-i \left(\frac{1}{2}a(1 + \xi_{j-1})\right) \left(\ln S_t - \frac{M\pi}{a}\right)\right] \xi_{j-1}} \Psi_0\left[\frac{1}{2}a(1 + \xi_{j-1})\right] \cdot \frac{1}{\left[P_{N-1}(\xi_{j-1})\right]^2}$$



$$C_0([\ln K]_u^*) \approx \Re \left[ \frac{e^{-a \left(\ln S_t - \frac{M\pi}{a} + \frac{2\pi}{a} (v-1)\right)}}{\pi} \cdot \frac{1}{M(N-1)} \cdot \omega^*\left(\frac{1}{2}a(1 + v_{j-1})\right) \right]$$



**Theorems of Equivalence**



The Call Price computed via Convergence Theorem is equal to the Call Price computed via Gauss Laguerre/Gander Gautschi Quadrature Rule



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  - Conclusions



Fast Option Pricing

$\vec{C}_t$  via DFT



Fast Fourier Transform Algorithms



Fast Option Pricing

$\vec{C}_t$  via DFT



Fractional FFT



$\vec{C}_i$  via DFT



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Bayley-Swarztrauber F-DFT Characterization



$$\omega(n) = \sum_{j=0}^{N-1} e^{-i2\pi kj\gamma} f(x_j) \text{ where } n = 1 \dots N$$



Bayley-Swarztrauber F-DFT Characterization



$$\omega(n) = \sum_{j=0}^{N-1} e^{-i2\pi kj\gamma} f(x_j) \text{ where } n = 1 \dots N$$

with  $\gamma$  that can be any complex number

If  $\gamma = \frac{1}{N}$



$$\omega(n) = \sum_{j=0}^{N-1} e^{-i\frac{2\pi}{N} j(n-1)} f(x_j) \text{ where } n = 1, 2, \dots, N$$

**The standard DFT definition**

Choosing two independent uniform grids



$$x_j = jg\left(\frac{a}{N}\right) \text{ for } j = 1 \dots N$$

**Spectral Grid**

$$[\ln K]_u^+ = \ln S_r - b + \lambda_u \text{ for } u = 1, \dots, N$$

**Log-Strike Grid**



Choosing two independent uniform grids



Implies choosing a specific value of  $\gamma$



$$\gamma = \frac{\lambda g\left(\frac{a}{N}\right)}{2\pi}$$

Fast Fractional Reconstruction



Step 1



Calculate sequences of  $2N$  points

Fast Fractional Reconstruction



Step 1



$$y = \left\{ \left( f\left( (j-1)g\left(\frac{a}{N}\right) \right) e^{-i\pi j^2 \gamma} \right)_{j=0}^{N-1}, (0)_{j=0}^{N-1} \right\}$$

$$z = \left\{ \left( e^{i\pi j^2 \gamma} \right)_{j=0}^{N-1}, \left( e^{i\pi (N-j)^2 \gamma} \right)_{j=0}^{N-1} \right\}$$



Fast Fractional Reconstruction



Step 2



Calculate

$$w = \psi_0 \left( (j-1)g\left(\frac{a}{N}\right) \right) \odot \left( e^{i\pi(N-j)^2\gamma} \right)_{j=0}^{N-1}$$



Fast Fractional Reconstruction



Step 3



Calculate via standard FFT

$$\bar{w} = FFT(w), \bar{z} = FFT(z)$$



Fast Fractional Reconstruction



Step 4



Calculate

$$q = \bar{w} \odot \bar{z}$$



Fast Fractional Reconstruction



Step 4



Calculate

$$\omega(n) = IFFT(q) \oslash \left( e^{i\pi j^2\gamma} \right)_{j=0}^{N-1}$$



Fast Fractional Reconstruction



The total computational cost drops



$O(N^2)$



$O(6N \log_2 N)$



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Gaussian Gridding



Step 1

Gaussian Convolution of the non uniformly sampled characteristic function



$$f_\tau(x) = \sum_{j=0}^{N-1} f(x_j) \sum_{k=-\infty}^{\infty} e^{-\frac{(x_j-x-2k\pi)^2}{4\tau}}$$



Gaussian Gridding



Step 1

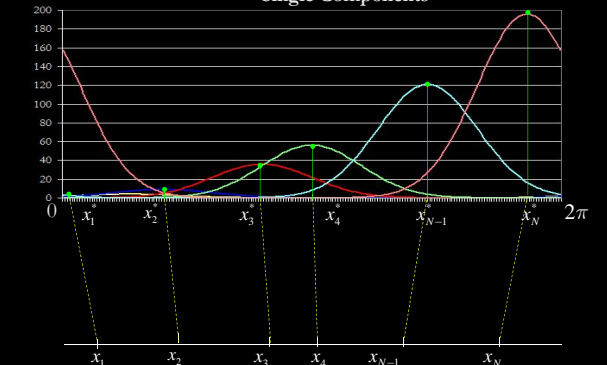
Gaussian Convolution of the non uniformly sampled characteristic function



$$f_\tau(x) = \sum_{j=0}^{N-1} f(x_j) \sum_{k=-\infty}^{\infty} e^{-\frac{(x_j-x-2k\pi)^2}{4\tau}}$$



Single Components



### Gaussian Gridding

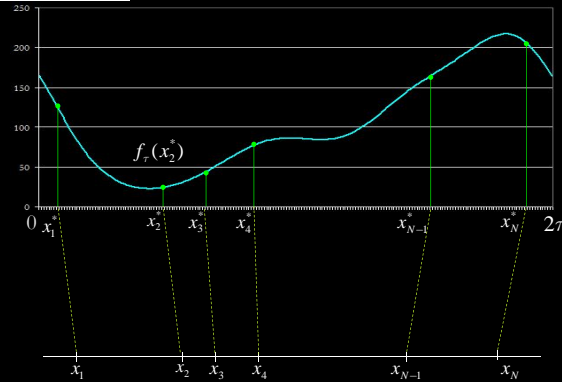
Step 1

Gaussian Convolution of the non uniformly sampled characteristic function

$$f_\tau(x) = \sum_{j=0}^{N-1} f(x_j) \sum_{k=-\infty}^{\infty} e^{-\frac{(x_j-x-2k\pi)^2}{4\tau}}$$



55



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### Gaussian Gridding

Step 2

Discretization on an uniform oversampled grid of  $f_\tau(x)$

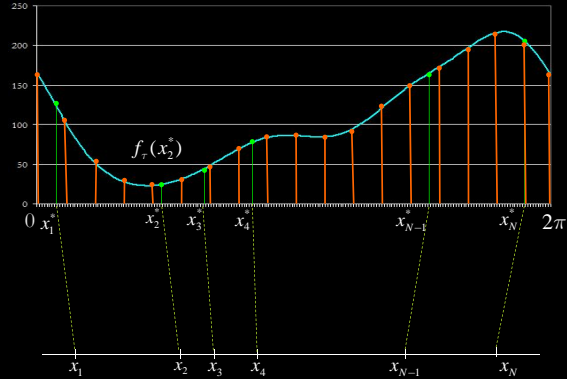
$$\tilde{f}_\tau(y_m) = \sum_{j=0}^{N-1} f(x_j) \tilde{K}(y_m, x_j, \sqrt{2\tau}; 2\pi)$$

where

$$\tilde{K}(y_m, x_j, \sqrt{2\tau}; 2\pi) = \begin{cases} \sum_{k=-\infty}^{\infty} e^{-\frac{(y_m - x_j - 2k\pi)^2}{4\tau}} & \text{if } y_m \text{ is in the support of } f_\tau \\ \sum_{k=-\infty}^{\infty} e^{-\frac{(y_m - x_j - 2k\pi)^2}{4\tau}} & \text{otherwise} \end{cases}$$



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### Gaussian Gridding

Step 3

Computation of the Fourier Coefficient of  $f_\tau(x)$  discretised

$$F_\tau(n) = \lim_{M_\tau \rightarrow \infty} \frac{1}{M_\tau} \sum_{m=0}^{M_\tau-1} \tilde{f}_\tau\left(\frac{2\pi m}{M_\tau}\right) e^{-im\frac{2\pi}{M_\tau}(n-1)}$$



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### Gaussian Gridding

Step 4

NU-DFT representation of the Fourier Coefficient  $F_\tau(n)$

$$\tilde{\omega}(n) = \sqrt{\frac{\pi}{\tau}} e^{n^2\tau} F_\tau(n)$$



60



### Gaussian Gridding

Step 5

DFT representation of the Fourier Coefficient  $F_\tau(n)$

$$F_\tau(n) = \lim_{M_\tau \rightarrow \infty} \frac{1}{M_\tau} \omega(n)$$

for  $n = 1, 2, \dots, \frac{M}{2}$



61



### Gaussian Gridding

Step 6

NU-DFT derivation as a function of DFT

$$\tilde{\omega}(n) = \lim_{M_\tau \rightarrow \infty} \sqrt{\frac{\pi}{\tau^*}} e^{n^2\tau^*} \frac{1}{M_\tau} \omega(n)$$

for  $n = 1, 2, \dots, \frac{M}{2}$



62



### Gaussian Gridding

Step 7

NU-FFT computation

$$\tilde{\omega}(n) = \lim_{M_\tau \rightarrow \infty} \sqrt{\frac{\pi}{\tau^*}} e^{n^2\tau^*} \frac{1}{M_\tau} \omega(n)$$

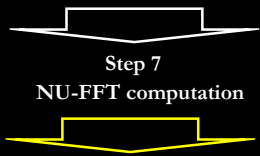
FFT



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### Gaussian Gridding

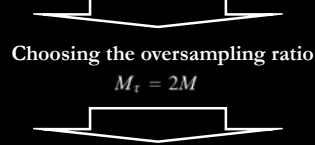


$$\tilde{\omega}(n) = \lim_{M_\tau \rightarrow \infty} \sqrt{\frac{\pi}{\tau^*}} e^{n^2 \tau^*} \frac{1}{M_\tau^2} \omega(n)$$

NU-FFT ——— FFT

### Computational Cost

The major computational cost of the Procedure is the FFT on the oversampled grid

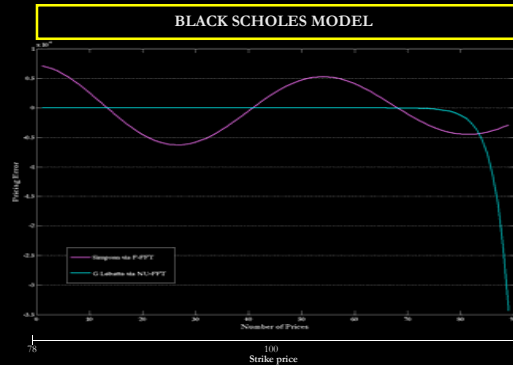


The total cost of the procedure is  $\approx 2M \log 2M$

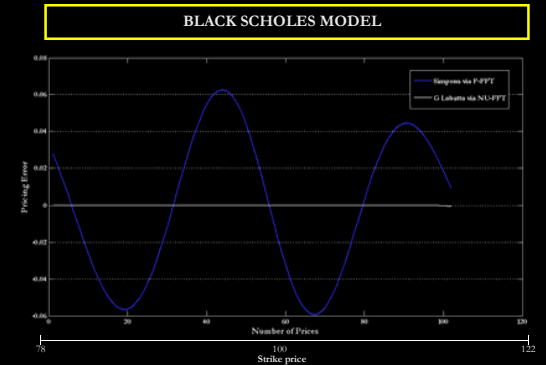
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# ACCURACY

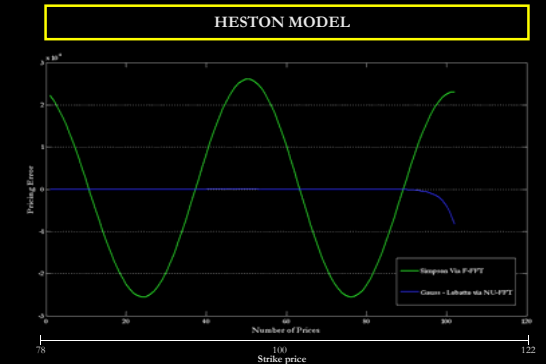
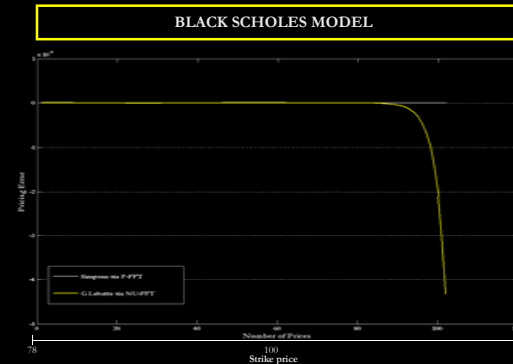
$\sigma = 0.3$



$\sigma = 0.1$



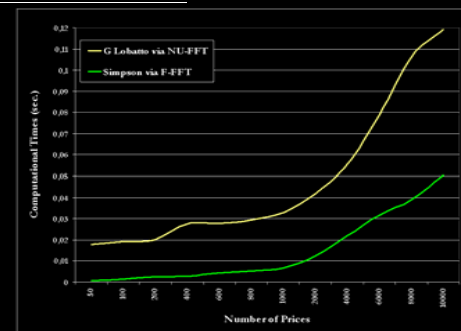
# STABILITY





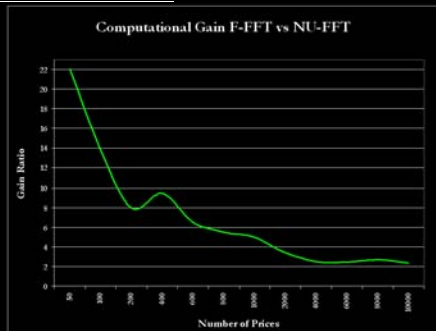


# SPEED



Centrino 1600Mhz – 1gb RAM  
 Mean Value over 1000 runs

The Computational Framework



Centrino 1600Mhz – 1gb RAM  
 Mean Value over 1000 runs

At very low time scales, the differences are negligible

Empirical Analysis

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Conclusions

### Use of Gaussian Grids

F-FFT	NO
NU – FFT	YES

Conclusions

### Indifference to Nyquist-Shannon Limit

F-FFT	YES
NU – FFT	YES

Conclusions

### Independent Price Grids

F-FFT	YES
NU – FFT	YES

FFT's like - Accuracy

F-FFT	YES
NU - FFT	YES



Stability of Pricing

F-FFT	NO
NU - FFT	YES



Speed of Pricing

F-FFT	YES
NU - FFT	YES



	F-FFT	NU - FFT
Gaussian Grids		
NS Limit		
Indipendent Grids		
Accuracy		
Stability		
Speed		

