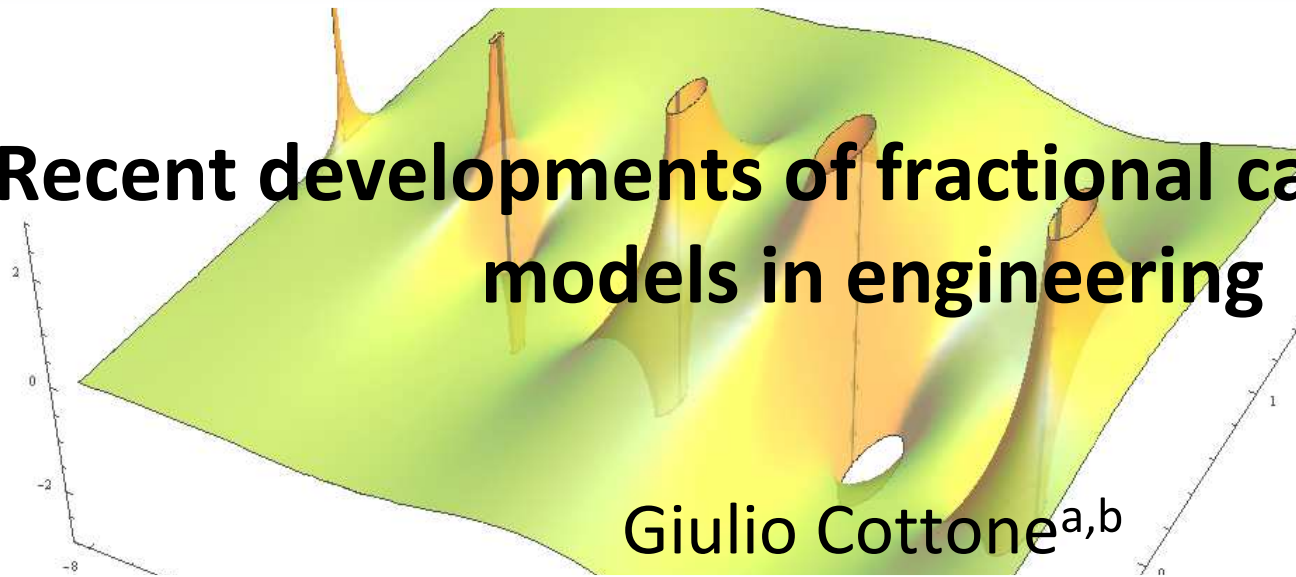


Recent developments of fractional calculus-based models in engineering



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FRACTIONAL CALCULUS: The name of the game: a misnomer!

Fractional operators

have nothing to do with fractions (1/3, 1/2...)!!!

Fractional operators

are derivatives and integrals of **real** (or **complex**) order:

$$f(x), \frac{df(x)}{dx}, \frac{d^2 f(x)}{dx^2}, \dots, \frac{d^j f(x)}{dx^j}, \dots, j \in \mathbb{N}$$

Q: How to define derivatives of order $j = 0.3 + 2.5i$?

Therefore, if someone says:

“I had **Fractional Calculus** in the elementary school”

there are two possibilities:

1. A probable misunderstanding
2. First effects of the Bologna process

FRACTIONAL CALCULUS? Not exotic!

“What if n (in $d^n f / dx^n$) be $\frac{1}{2}$?”

De L'Hospital



1661 - 1704

Leibnitz



1646 - 1716

It will lead to a paradox, from which one day **useful consequences** will be drawn. (1695)

Leibnitz, Euler, Riemann, Liouville, Abel, Feller, Grünwald, Letnikov, Marchaud, Weyl, Caputo, Riesz, Samko

Useful consequences in Mechanics

- Viscoelasticity
- Fracture Mechanics
- Non-local Continuum Mechanics
- Stochastic Dynamics

FRACTIONAL CALCULUS? Never heard! WHY?

- Many definitions and symbols
 - Riemann - Liouville
 - Marchaud
 - Grünwald-Letnikov
 - Riesz
 - Weyl
 - Caputo
 - ...
- Calculations are very hard to be tackled by hand (CAS needed)
- Lack of simple geometrical meaning
- Is it useful?

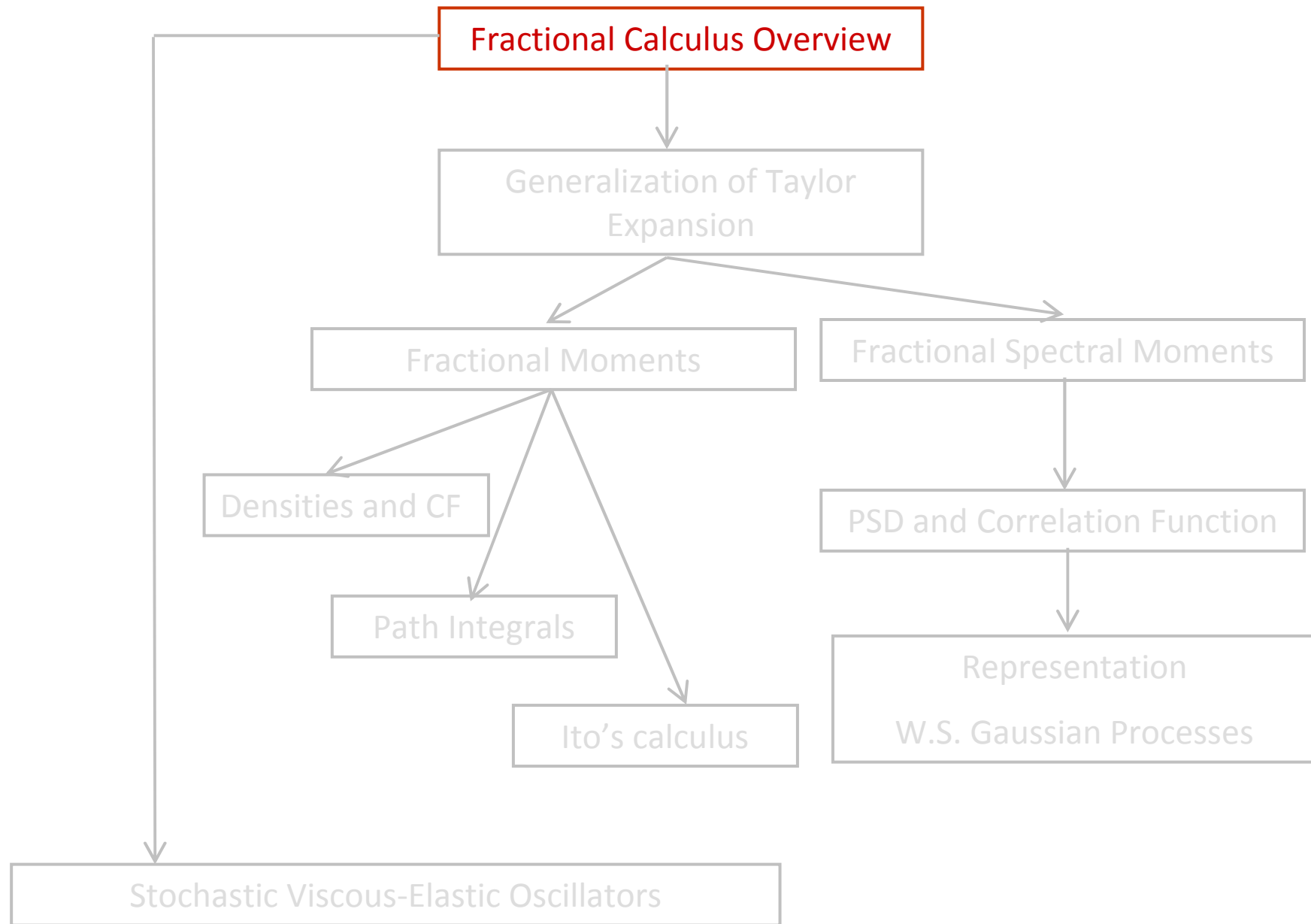
Books:

For engineers: Podlubny

For Physicists: Hilfer

Encyclopedic textbook: Samko, Kilbas and Marichev

OUTLINE



FRACTIONAL CALCULUS OVERVIEW: WHICH IS THE SCOPE?

- Classical derivatives:

$$f(x), \frac{df(x)}{dx}, \frac{d^2 f(x)}{dx^2}, \dots, \frac{d^j f(x)}{dx^j}, \dots, j \in \mathbb{N}$$

- Fractional derivatives:

$$(D^\alpha f)(x) \quad \alpha \in \mathbb{C}, \operatorname{Re}[\alpha] > 0$$

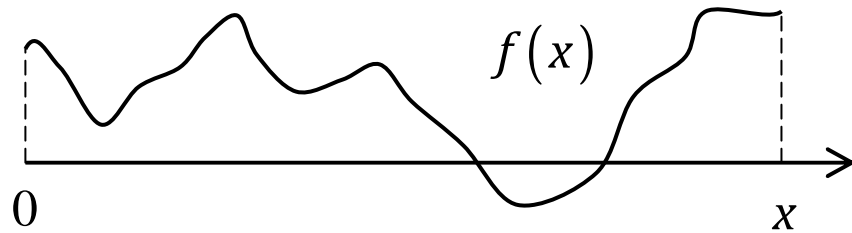
-
- Classical n-folded integrals:

$$f(x), \int f(x) dx, \iint f(x) dx dx, \dots, \underbrace{\int \dots \int f(x) dx \dots dx}_{n\text{-fold}}$$

- Fractional integrals:

$$(I^\alpha f)(x) \quad \alpha \in \mathbb{C}, \operatorname{Re}[\alpha] > 0$$

First step: definition of the Fractional Integral



$$(I_{0+}^1 f)(x) = \int_0^x f(x_1) dx_1$$

$$(I_{0+}^2 f)(x) = \int_0^x \left(\int_0^{x_1} f(x_2) dx_2 \right) dx_1$$

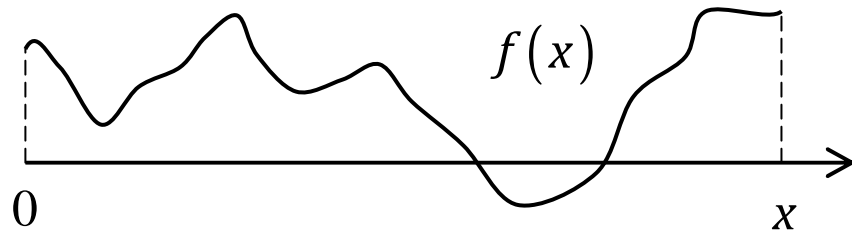
...

Cauchy formula

$$\begin{aligned} (I_{0+}^n f)(x) &= \underbrace{\int_0^x dx_1 \dots \int_0^{x_{n-1}} f(x_n) dx_n}_{n\text{-fold}} = \\ &= \frac{1}{(n-1)!} \int_0^x (x-z)^{n-1} f(z) dz \end{aligned}$$

We want to extend this formula to non-integer number. HOW?

First step: definition of the Fractional Integral



$$(I_{0+}^1 f)(x) = \int_0^x f(x_1) dx_1$$

$$(I_{0+}^2 f)(x) = \int_0^x \left(\int_0^{x_1} f(x_2) dx_2 \right) dx_1$$

...

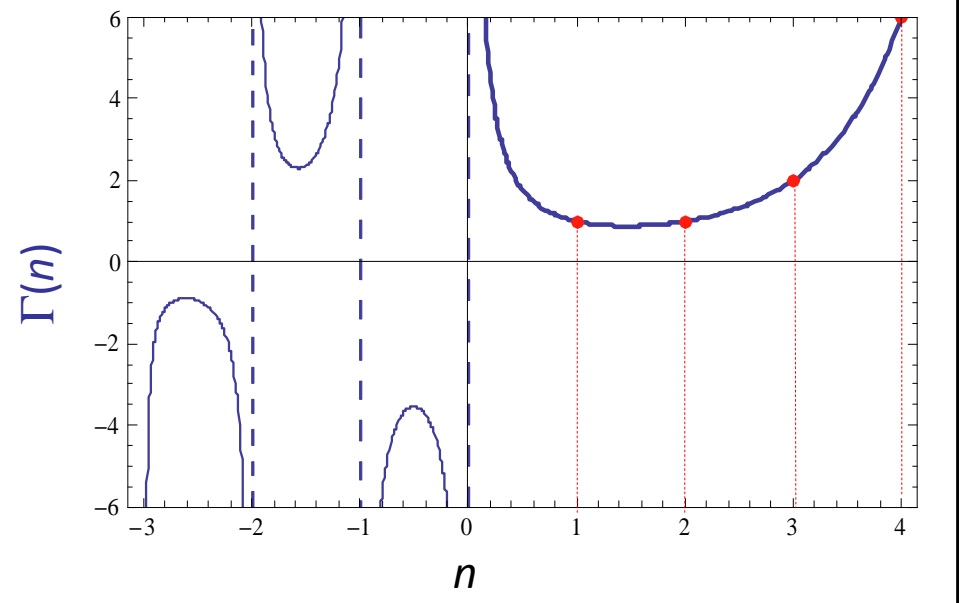
Cauchy formula

$$\begin{aligned} (I_{0+}^n f)(x) &= \underbrace{\int_0^x dx_1 \dots \int_0^{x_{n-1}} f(x_n) dx_n}_{n\text{-fold}} = \\ &= \frac{1}{(n-1)!} \int_0^x (x-z)^{n-1} f(z) dz \end{aligned}$$

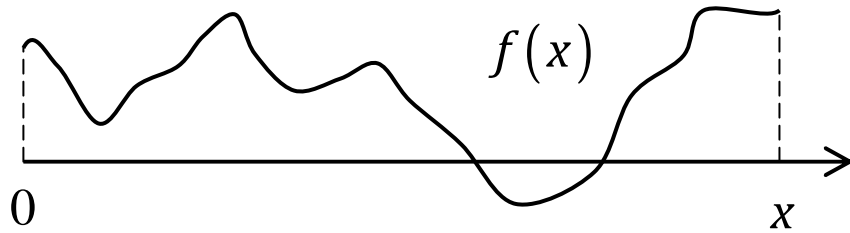
Euler Gamma function

$$\Gamma(\alpha) \stackrel{\text{def}}{=} \int_0^{\infty} e^{-z} z^{\alpha-1} dz$$

$$\Gamma(n) = (n-1)!$$



First step: definition of the Fractional Integral



$$(I_{0+}^1 f)(x) = \int_0^x f(x_1) dx_1$$

$$(I_{0+}^2 f)(x) = \int_0^x \left(\int_0^{x_1} f(x_2) dx_2 \right) dx_1$$

...

$$(I_{0+}^n f)(x) = \underbrace{\int_0^x dx_1 \dots \int_0^{x_{n-1}} f(x_n) dx_n}_{n\text{-fold}} = \frac{1}{(n-1)!} \int_0^x (x-z)^{n-1} f(z) dz$$

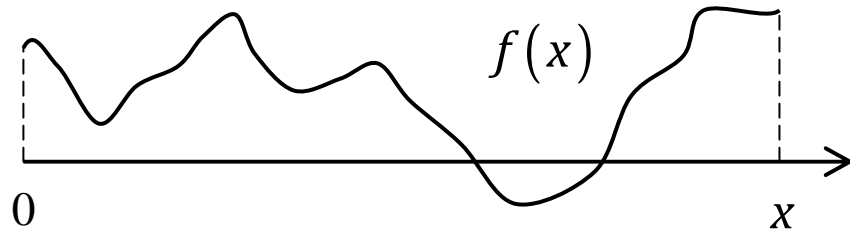
$$\Gamma(n) = (n-1)!$$

Riemann – Liouville fractional integral

$$(I_{0+}^\alpha f)(x) = \frac{1}{\Gamma(\alpha)} \int_0^x (x-z)^{\alpha-1} f(z) dz$$

$$\alpha \in \mathbb{C}, \quad \text{Re}[\alpha] > 0$$

Second step: From the Fractional Integrals to the fractional Derivatives



Riemann – Liouville fractional integral

$$(I_{0+}^{\alpha} f)(x) = \frac{1}{\Gamma(\alpha)} \int_0^x (x-z)^{\alpha-1} f(z) dz$$

$$f(x) = \frac{d}{dx} [(I_{0+}^1 f)(x)]$$

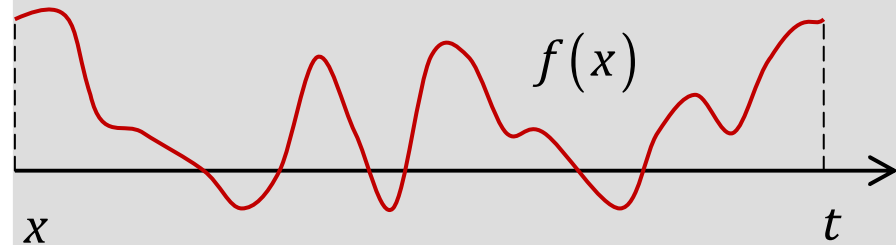
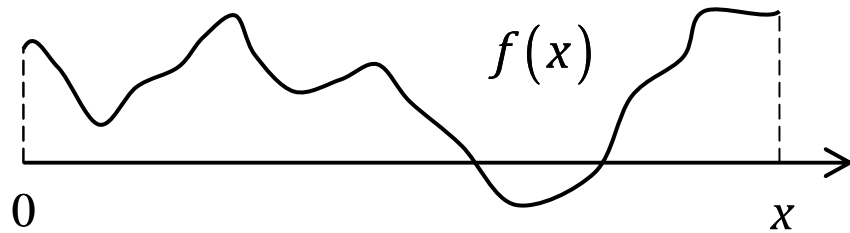
$$(D^1 f)(x) = \frac{d}{dx} \left(\frac{d}{dx} [(I_{0+}^1 f)(x)] \right)$$

Riemann – Liouville fractional derivative

$$(D_{0+}^{\alpha} f)(x) = \frac{d^m}{dx^m} [(I_{0+}^{m-\alpha} f)(x)]$$

$$m = [\alpha] + 1$$

Second step: From the Fractional Integrals to the fractional Derivatives



Riemann – Liouville fractional integral

$$(I_{0+}^{\alpha} f)(x) = \frac{1}{\Gamma(\alpha)} \int_0^x (x-z)^{\alpha-1} f(z) dz$$

$$f(x) = \frac{d}{dx} [(I_{0+}^1 f)(x)]$$

$$(D^1 f)(x) = \frac{d}{dx} \left(\frac{d}{dx} [(I_{0+}^1 f)(x)] \right)$$

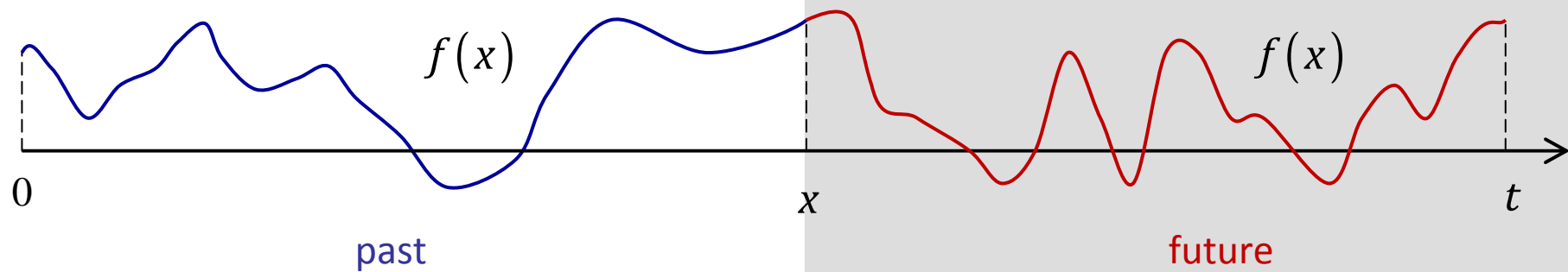
Riemann – Liouville fractional derivative

$$(D_{0+}^{\alpha} f)(x) = \frac{d^m}{dx^m} [(I_{0+}^{m-\alpha} f)(x)]$$

$$m = [\alpha] + 1$$

May we choose another interval?

The memory of Fractional Integrals and Derivatives



Left Riemann – Liouville fractional integral

$$(I_{0+}^{\alpha} f)(x) = \frac{1}{\Gamma(\alpha)} \int_0^x (x-z)^{\alpha-1} f(z) dz$$

Left Riemann-Liouville fractional derivative

$$(D_{0+}^{\alpha} f)(x) = \frac{d^m}{dx^m} \left[(I_{0+}^{m-\alpha} f)(x) \right]$$

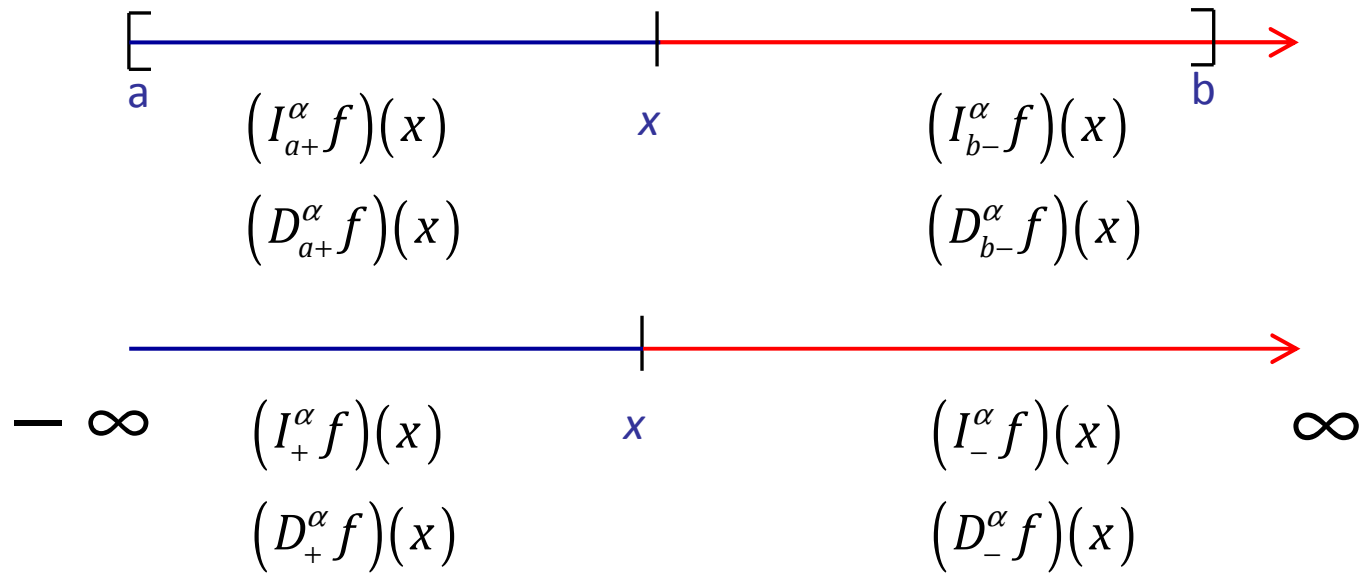
Right Riemann – Liouville fractional integral

$$(I_{t-}^{\alpha} f)(x) = \frac{1}{\Gamma(\alpha)} \int_x^t (x-z)^{\alpha-1} f(z) dz$$

Right Riemann-Liouville fractional derivative

$$(D_{t-}^{\alpha} f)(x) = (-1)^m \frac{d^m}{dx^m} \left[(I_{t-}^{m-\alpha} f)(x) \right]$$

Possible intervals for Fractional Derivatives



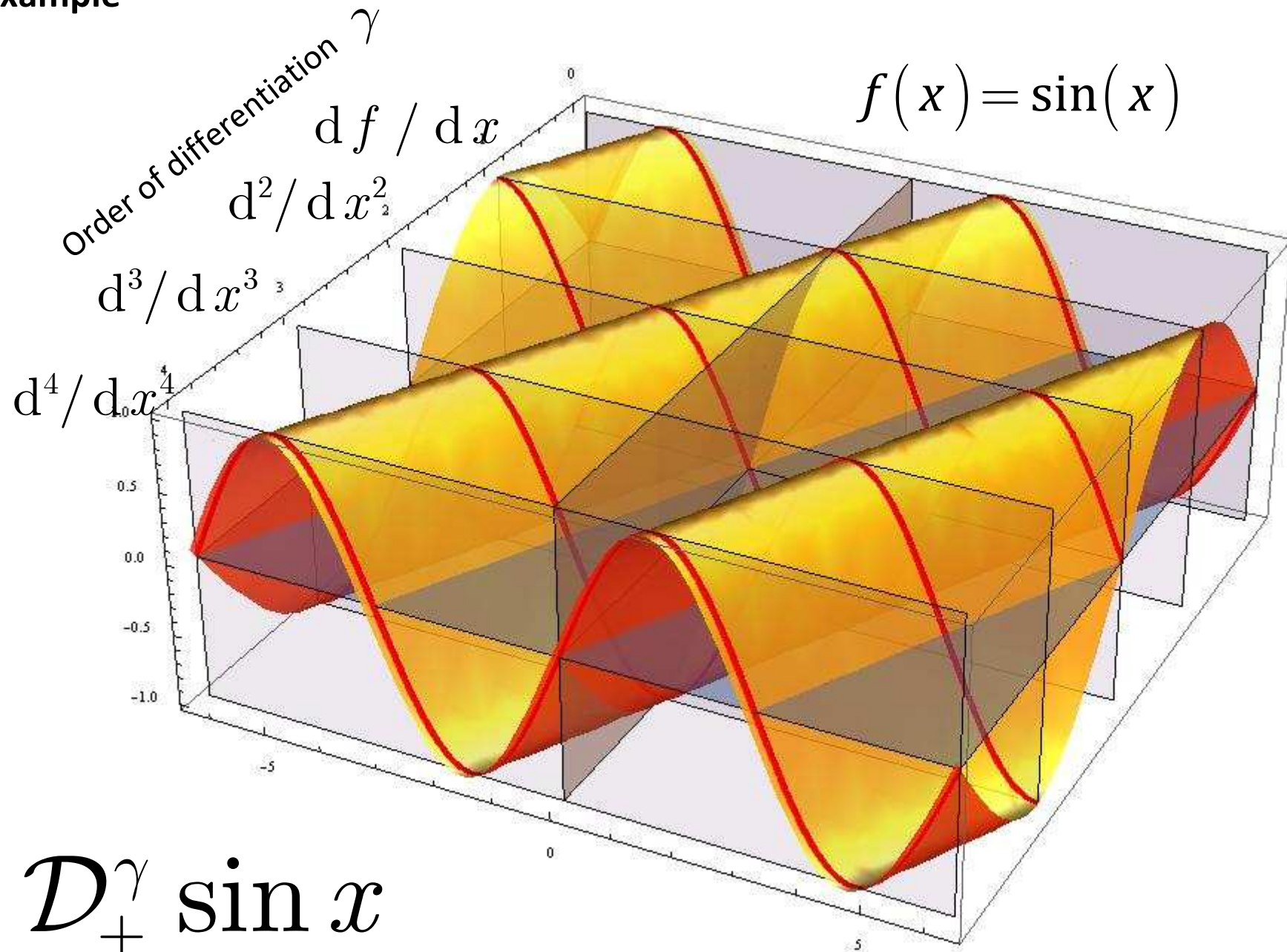
Properties

Composition

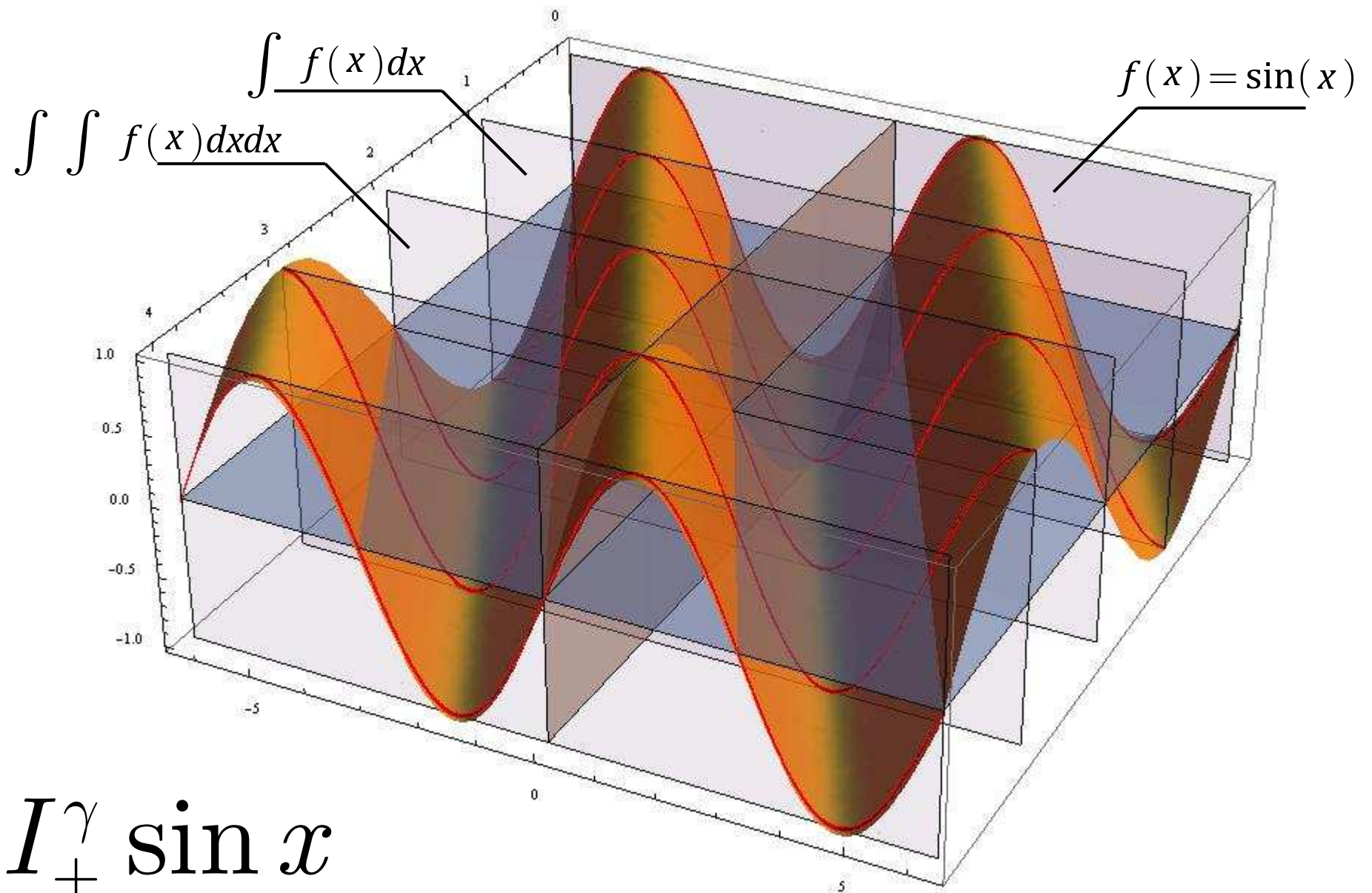
$$D_{a\pm}^\alpha I_{a\pm}^\beta f = D_{a\pm}^{\alpha-\beta} f$$

- *Leibnitz's rule*
- *Integration by parts*
- *Fourier Transform*
- *Taylor's expansion*

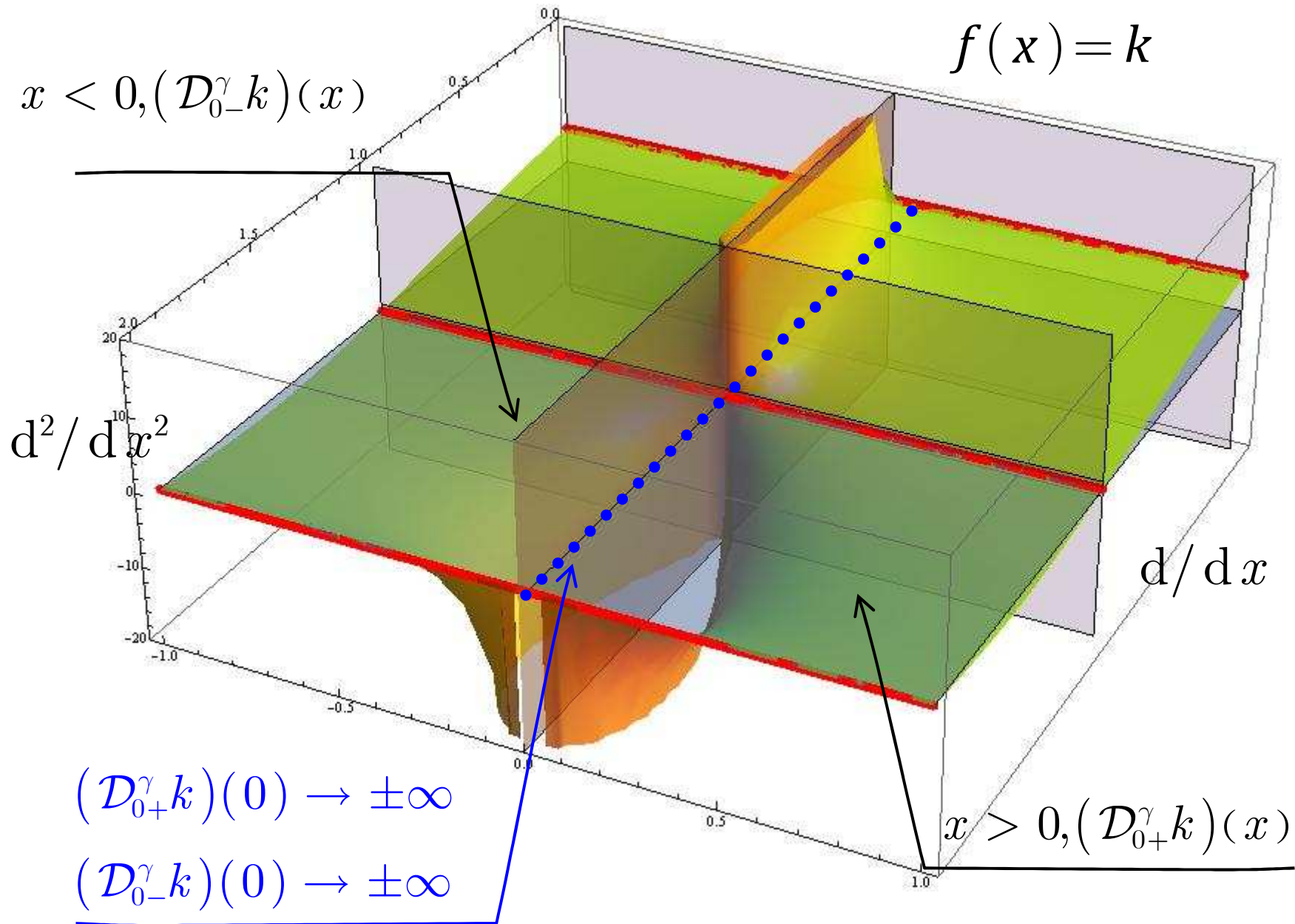
Example



Example



Example



Classical Derivatives

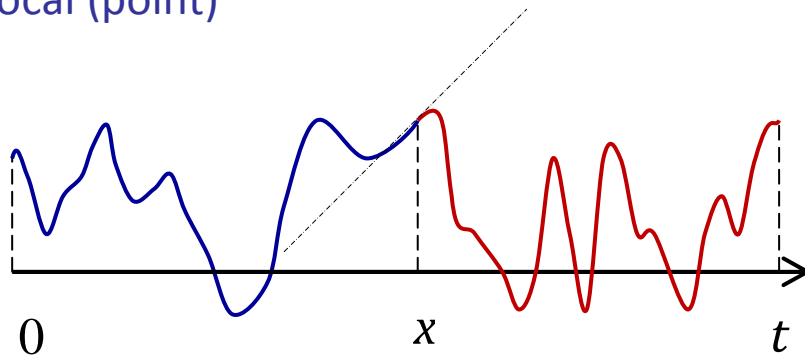
VERSUS

Fractional Derivatives

- Integer order
- One definition

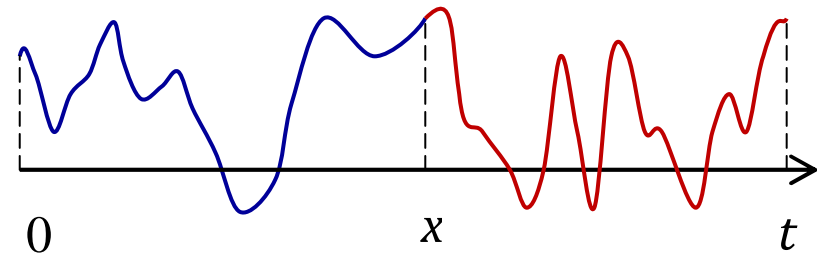
- Real or Complex order
- Different definitions
- Generalize the classical derivatives
- Interpolate the classical derivatives

- Local (point)



- Taylor series: approximation in a point's neighborhood
- Models via ODE and PDE

- Global (interval)

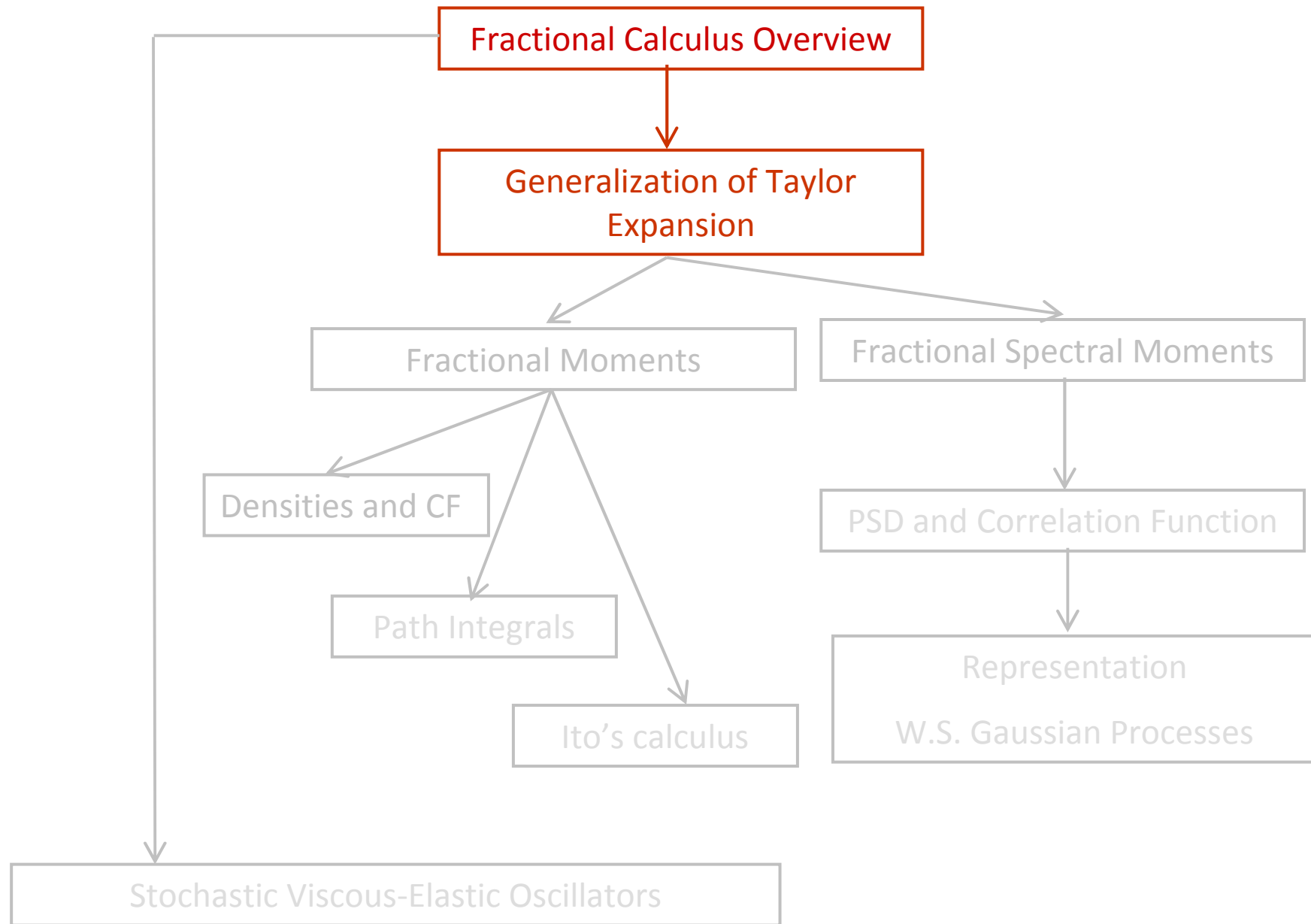


- Generalized Taylor series: approximation in the whole interval
- Models via FDE and PFDE

MODELLING WITH FRACTIONAL DERIVATIVES. *WHEN?*

- Phenomena with memory
 - Viscoelasticity
 - Nonlinear behaviour
 - Long correlated processes (inverse power-law decay)
 - Long-range interactions (inverse power-law decay)
- Alternative to higher-order models (i.e. gradient non-local theory)
- Failure of Taylor series:
 - Does not catch the necessary information (non-local)
 - Cannot be calculated (not derivable functions)

OUTLINE

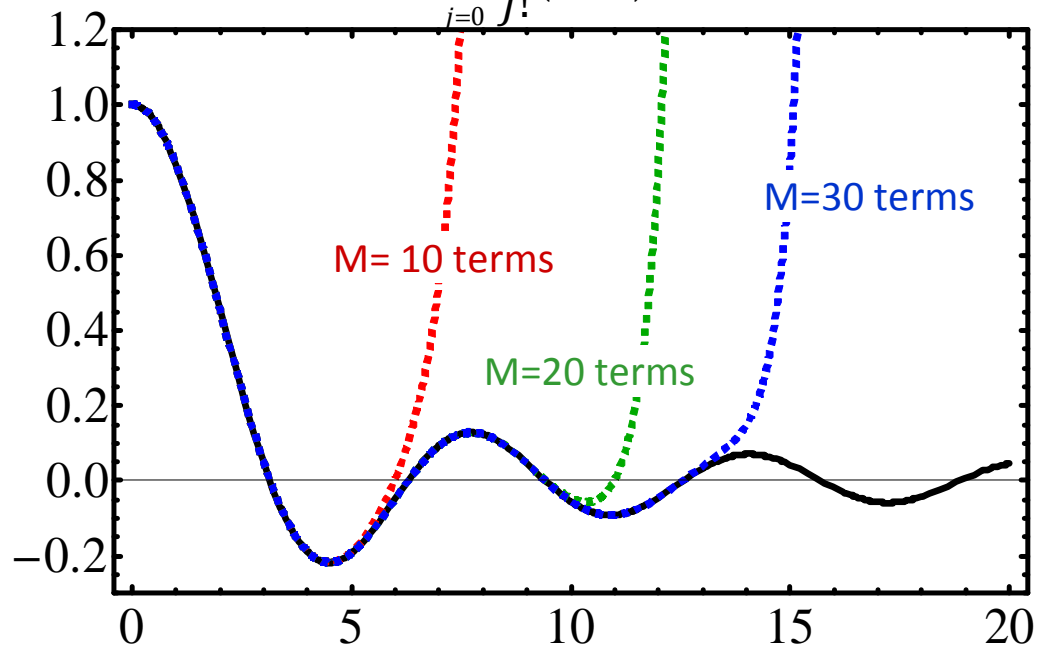


Classical Taylor expansion

$$f(x) = \sum_{j=0}^{\infty} \frac{1}{j!} (D^j f)(0) x^j$$

$$(D^j f)(0) = \left. \frac{d^j f(x)}{dx^j} \right|_{x=0}$$

$$f(x) = \sum_{j=0}^M \frac{1}{j!} (D^j f)(0) x^j$$



$$f(x) = \frac{\sin(x)}{x}$$

Generalization of the Taylor expansion by

- Riemann (Hardy proved the asymptotic convergence)

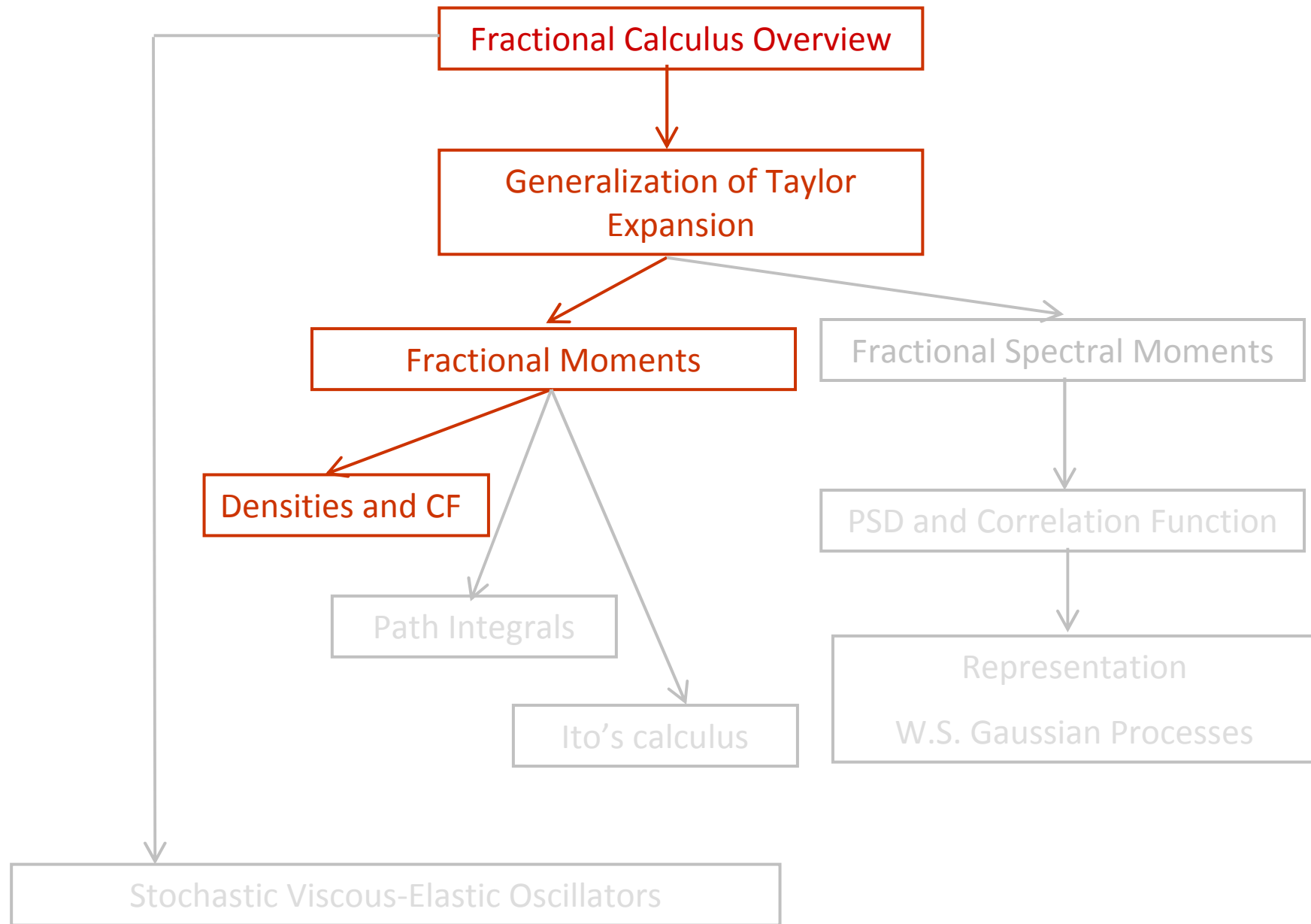
$$f(x+h) = \sum_{m=-\infty}^{\infty} \frac{h^{m+r}}{\Gamma(m+r+1)} (\mathcal{D}_+^{m+r} f)(x) \quad (\text{post. 1876})$$

- Dzherbashyan and Nerseyan (1958)
- Osler (1972)
- Samko, Kilbas, Marichev* (1993)
- Trujillo, Bonilla, Rivero (1999)
- Jumarie (2006)

Integral Taylor form (Samko et al., Cottone et al.)

$$f(\pm\xi) = \frac{1}{2\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Gamma(\gamma) (I_{\pm}^{\gamma} f)(0) |\xi|^{-\gamma} d\gamma$$

OUTLINE



Characteristic Function – Integer Moments relation

$$\phi_X (\vartheta) = E [\exp (i\vartheta X)] = \sum_{j=0}^{\infty} \frac{(i\vartheta)^j E[X^j]}{j!}$$

Generalization of the Taylor expansion

For general function

$$f (\pm\xi) = \frac{1}{2\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Gamma (\gamma) (I_{\pm}^{\gamma} f) (0) |\xi|^{-\gamma} d\gamma$$

For the Fourier transform of the probability density function (characteristic function)

$$\phi_X (\vartheta) = \frac{1}{2\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Gamma (\gamma) E \left[(\mp iX)^{-\gamma} \right] |\vartheta|^{-\gamma} d\gamma$$

“Taylor series like” form

$$\phi_X (\vartheta) \cong \frac{\Delta}{2\pi} \sum_{k=-M}^M \Gamma (\gamma_k) E \left[(-iX)^{-\gamma_k} \right] |\vartheta|^{-\gamma_k} \quad \gamma_k = \rho + ik\Delta,$$

What are the fractional moments? (keep in mind the misnomer...)

Definition, knowing the density

$$E \left[(\pm iX)^{\mp \gamma} \right] = \int_{-\infty}^{\infty} p(x) (\pm ix)^{\mp \gamma} dx$$

Definition, knowing the characteristic function

$$E \left[(\pm iX)^{\gamma} \right] = \left(D_{\mp}^{\gamma} \phi_X \right) (0) \quad \text{Re}[\gamma] > 0$$

$$E \left[(\pm iX)^{-\gamma} \right] = \left(I_{\mp}^{\gamma} \phi_X \right) (0)$$

How to calculate fractional moments from data?

$$X_1, X_2, \dots, X_n$$

$$E \left[(iX)^{0.5+2.3i} \right] = \frac{1}{n} \sum_{j=1}^n \left(iX_j \right)^{0.5+2.3i} \quad \text{Not Exotic!}$$

EXAMPLE

Standard Gaussian random variable

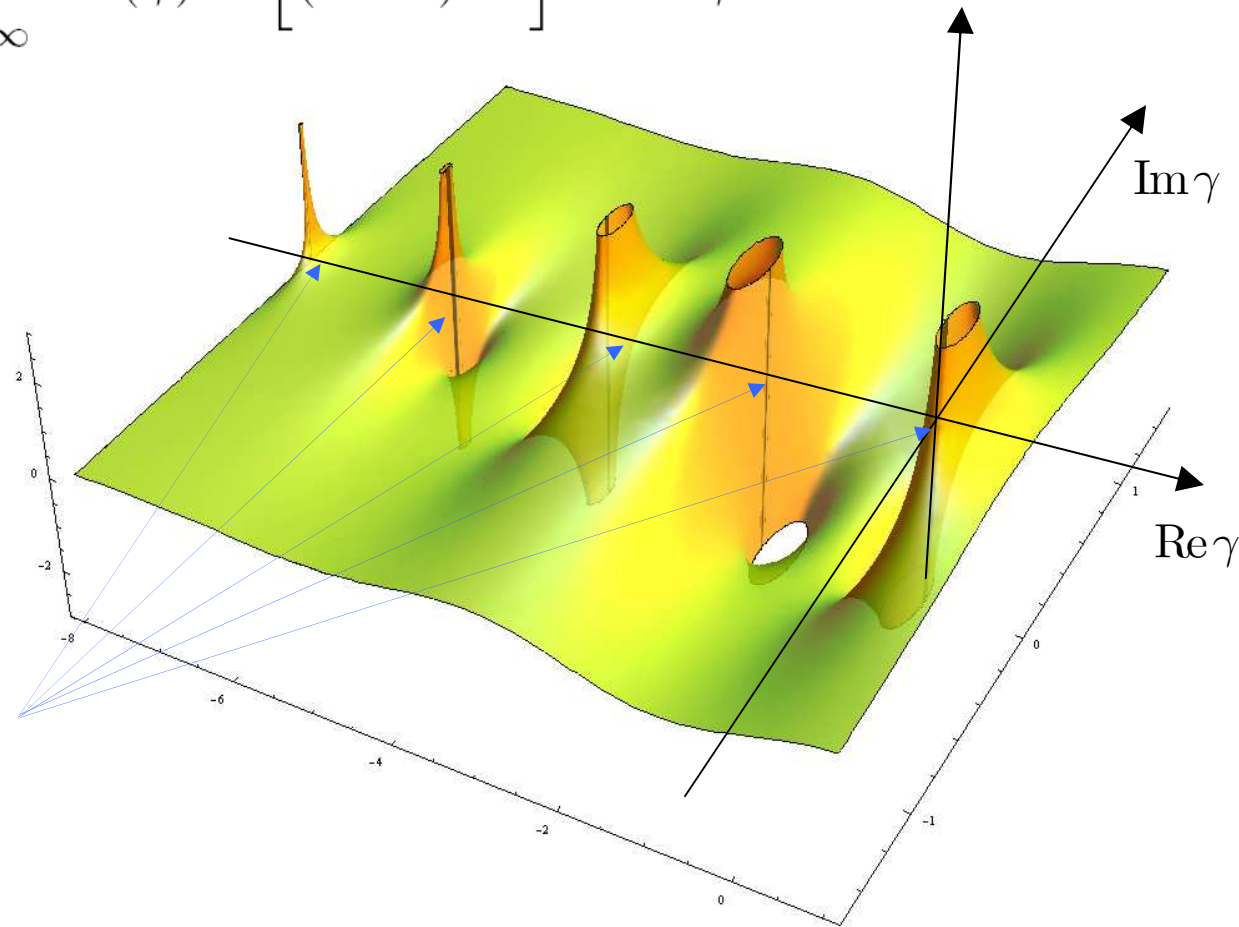
$$\phi_X(\vartheta) = e^{-\frac{\vartheta^2}{2}}$$

For a fixed value of $\vartheta = 1.5$ we plot the REAL part of the integrand

$$\phi_X(\vartheta) = \frac{1}{2\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Gamma(\gamma) E[(-iX)^{-\gamma}] \vartheta^{-\gamma} d\gamma \quad \text{Re} \left[\Gamma(\gamma) E[(-iX)^{-\gamma}]^{-\gamma} \vartheta^{-\gamma} \right]$$

**ISOLATED
SINGULARITIES**

at **0,-2,-4...**



EXAMPLE Standard Gaussian random variable

In order to evaluate the integral, the value of ρ must be properly selected

$$\phi_X(\vartheta) = \frac{1}{2\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Gamma(\gamma) E[(-iX)^{-\gamma}] \vartheta^{-\gamma} d\gamma \quad \gamma = \rho + i\eta$$

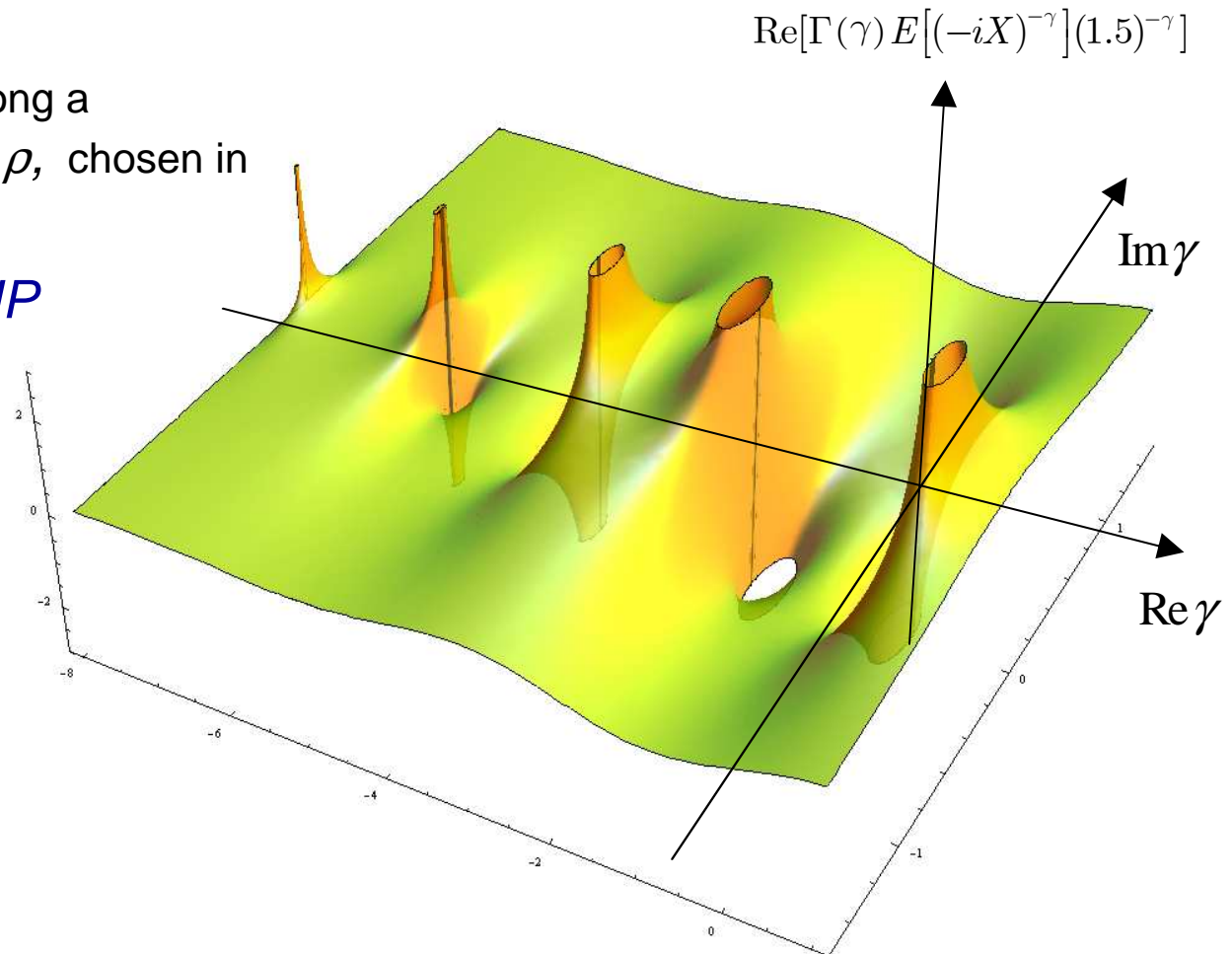
The integral is performed along a imaginary axis with real part ρ , chosen in the

FUNDAMENTAL STRIP

$$0 < \rho < 1$$

In particular we select

$$\rho = 1/2$$

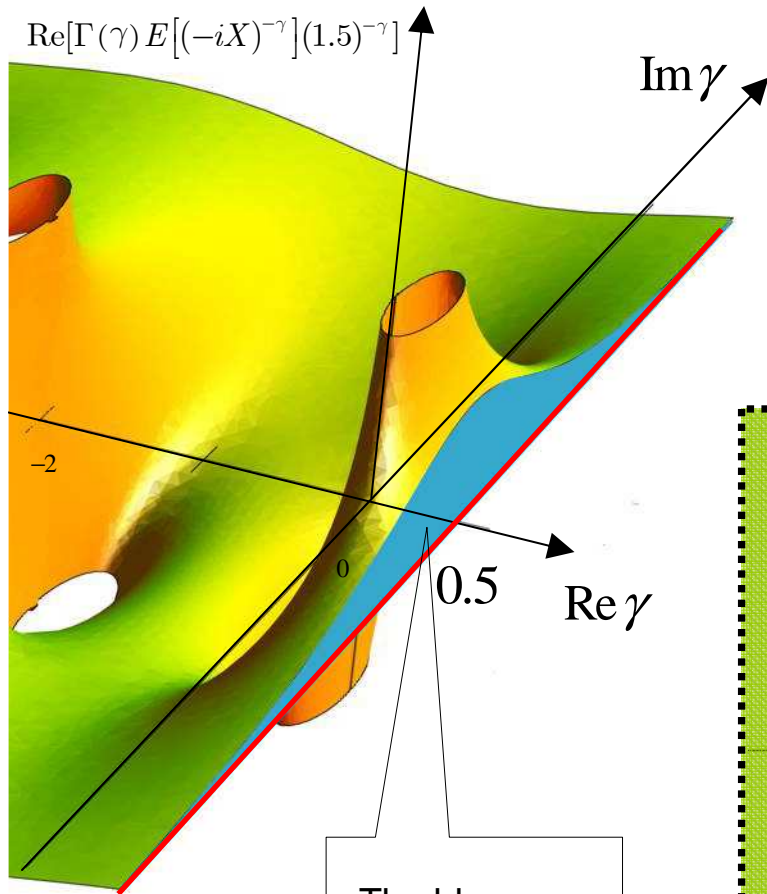


EXAMPLE Standard Gaussian random variable

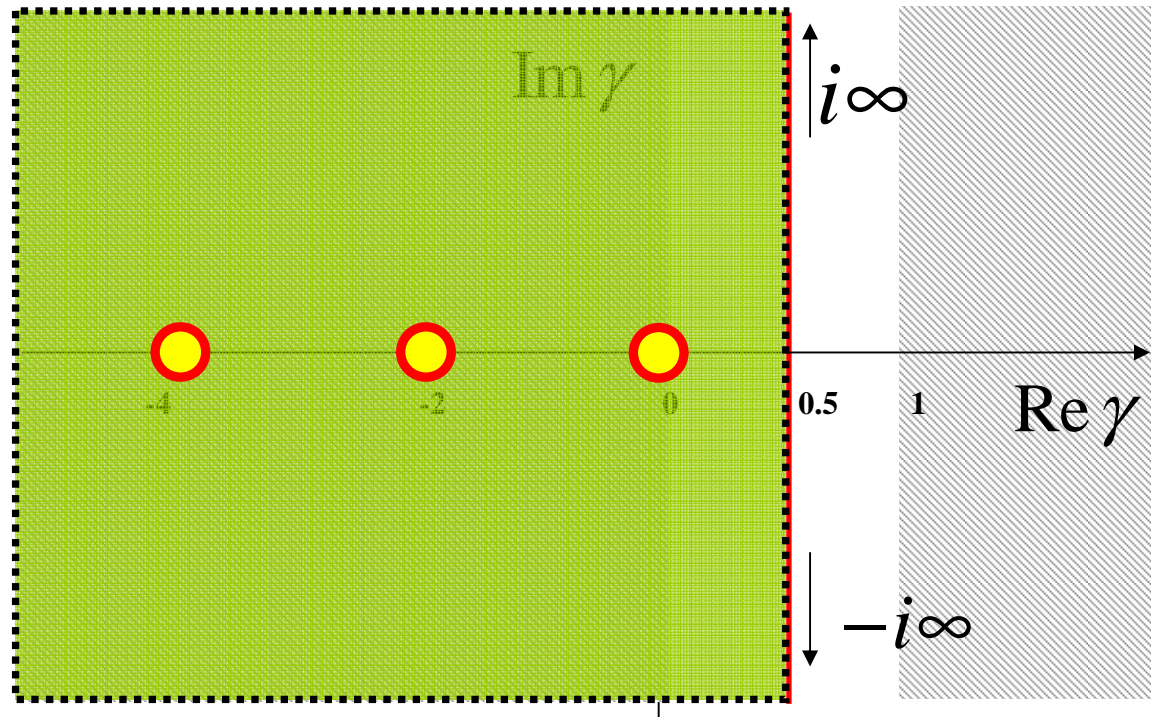
$$\phi_X(\vartheta) = \frac{1}{2\pi i} \int_{1/2-i\infty}^{1/2+i\infty} \Gamma(\gamma) E[(-iX)^{-\gamma}] \vartheta^{-\gamma} d\gamma = \text{RESIDUE THEOREM}$$

$$= \sum_{n=-\infty}^0 \text{Res} \left\{ \Gamma(n) E[(-iX)^{-n}] \vartheta^{-n} \right\} =$$

$$= \sum_{n=0}^{\infty} \frac{(i\vartheta)^n}{n!} E[X^n]$$

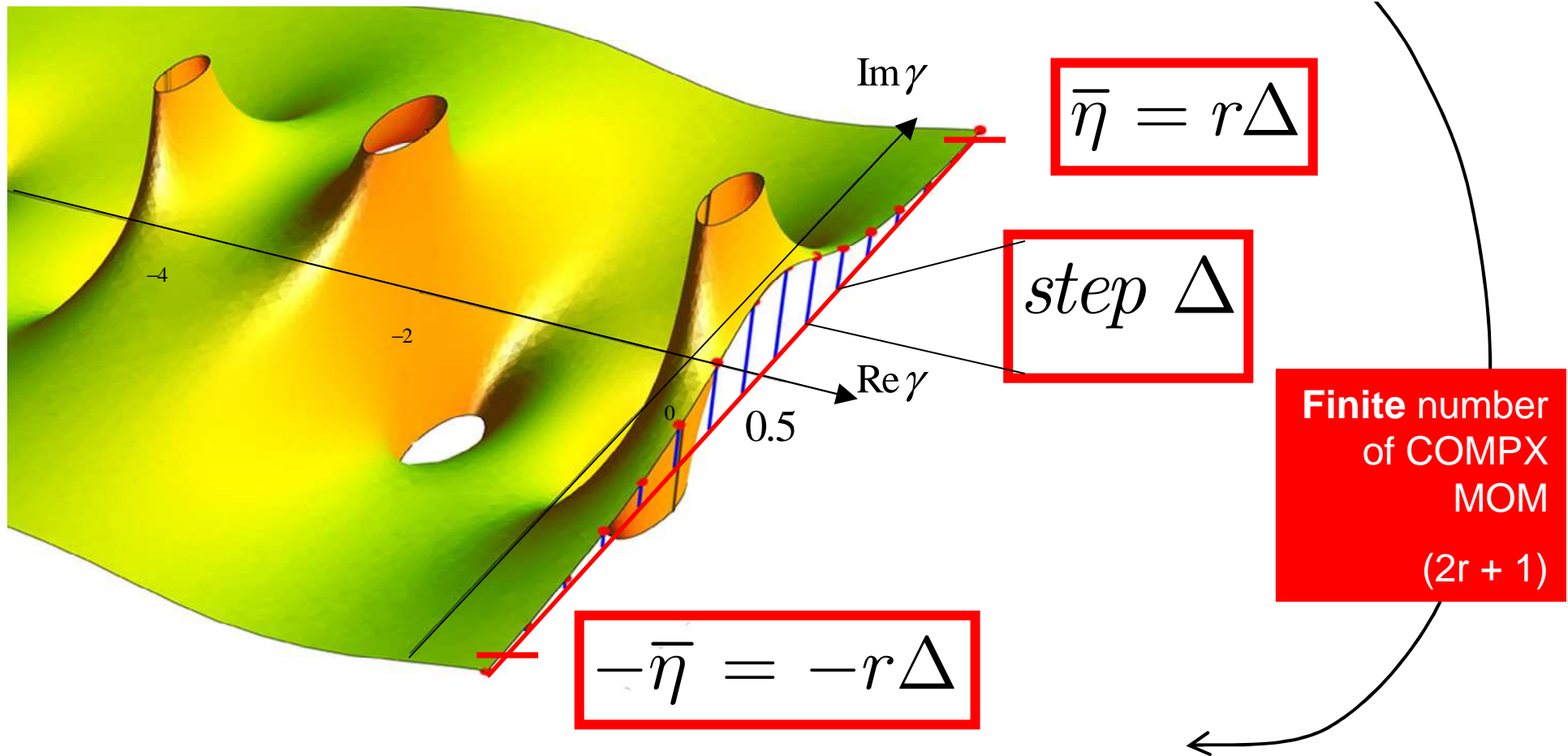


The blue area has to be evaluated



EXAMPLE Standard Gaussian random variable

$$\phi_X(\vartheta) = \frac{1}{2\pi i} \int_{0.5-\bar{\eta}}^{0.5+\bar{\eta}} \Gamma(\gamma) E\left[(-iX)^{-\gamma}\right] \vartheta^{-\gamma} d\gamma$$



$$\phi_X(\vartheta) = \frac{\Delta}{2\pi} \sum_{k=-r}^r \Gamma(0.5 + ik\Delta) E\left[(-iX)^{-(0.5+ik\Delta)}\right] (\vartheta)^{-(0.5+ik\Delta)}$$

Comparison between the integer and fractional moments series

$$\phi_X(\vartheta) = \exp(-\vartheta^2 / 2)$$

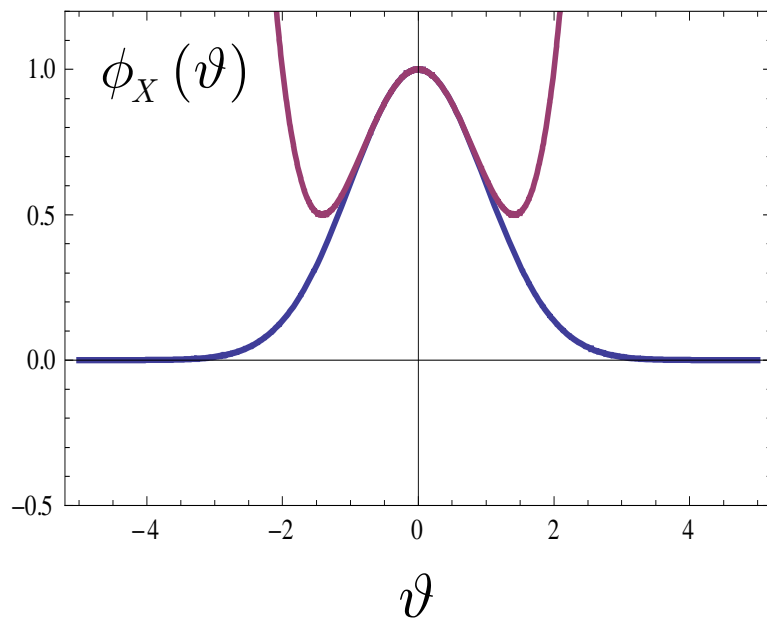
$$\phi_X(\vartheta) \approx \sum_{n=0}^4 \frac{(i\vartheta)^n}{n!} E[X^n]$$

$$\phi_X(\vartheta) = \frac{\Delta}{2\pi} \sum_{k=-2}^2 \Gamma(\gamma_k) E\left[(-iX)^{-\gamma_k}\right] (\vartheta)^{-\gamma_k}$$

$$\gamma_k = 1/2 + ik\Delta, \quad \Delta = 0.4$$

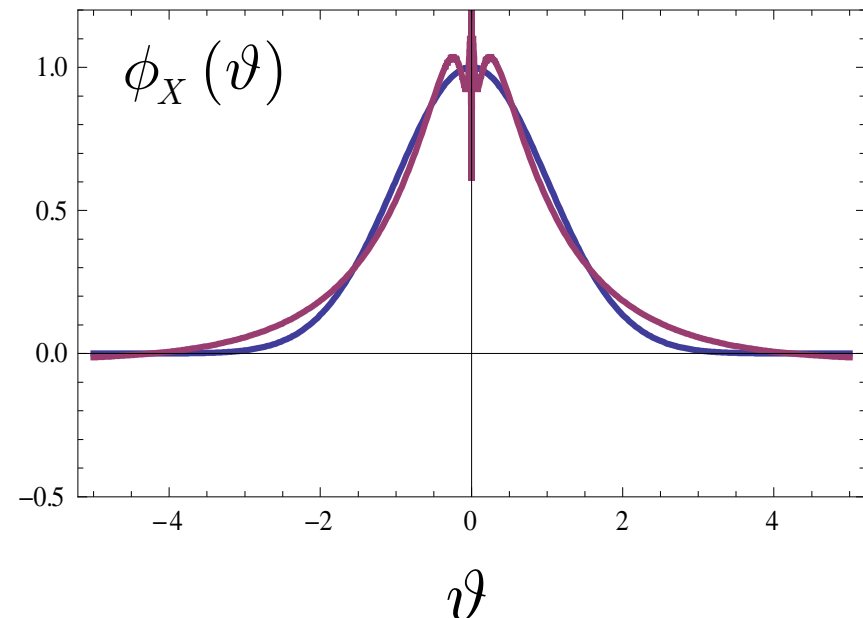
5 INTEGER MOMENTS

— exact
— approximated



5 FRACTIONAL MOMENTS

— exact
— approximated



Comparison between the integer and fractional moments series

$$\phi_X(\vartheta) = \exp(-\vartheta^2 / 2)$$

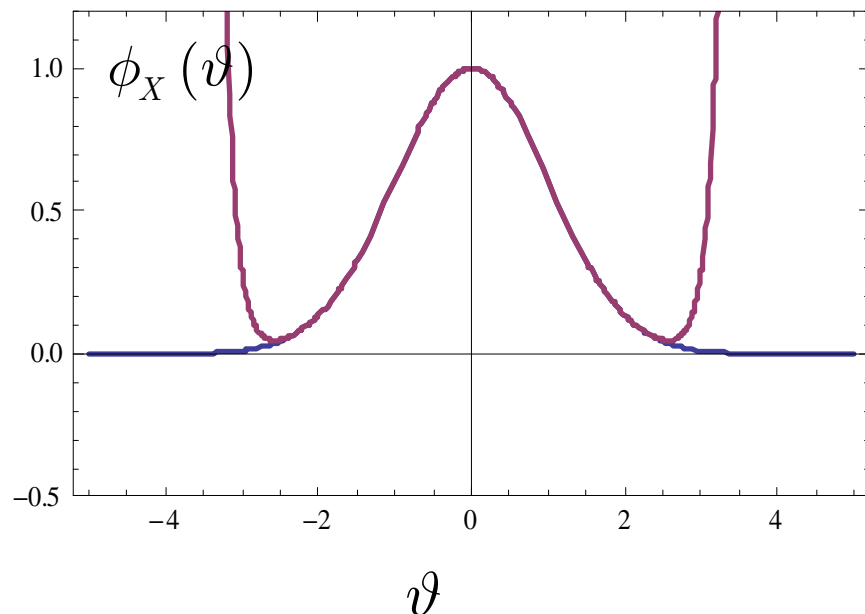
$$\phi_X(\vartheta) \approx \sum_{n=0}^{20} \frac{(i\vartheta)^n}{n!} E[X^n]$$

$$\phi_X(\vartheta) = \frac{\Delta}{2\pi} \sum_{k=-10}^{10} \Gamma(\gamma_k) E\left[(-iX)^{-\gamma_k}\right] (\vartheta)^{-\gamma_k}$$

$$\gamma_k = 1/2 + ik\Delta, \quad \Delta = 0.4$$

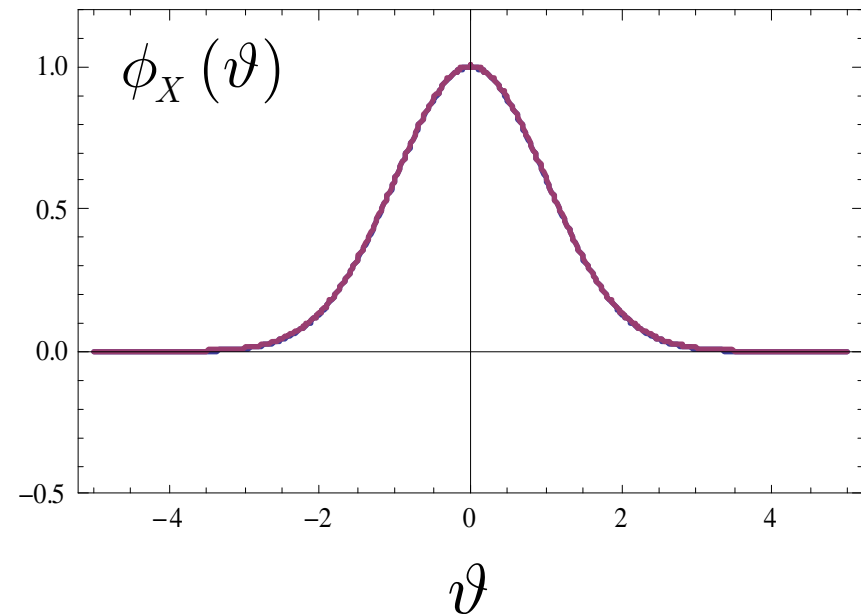
21 INTEGER MOMENTS

— exact
— approximated



21 FRACTIONAL MOMENTS

— exact
— approximated



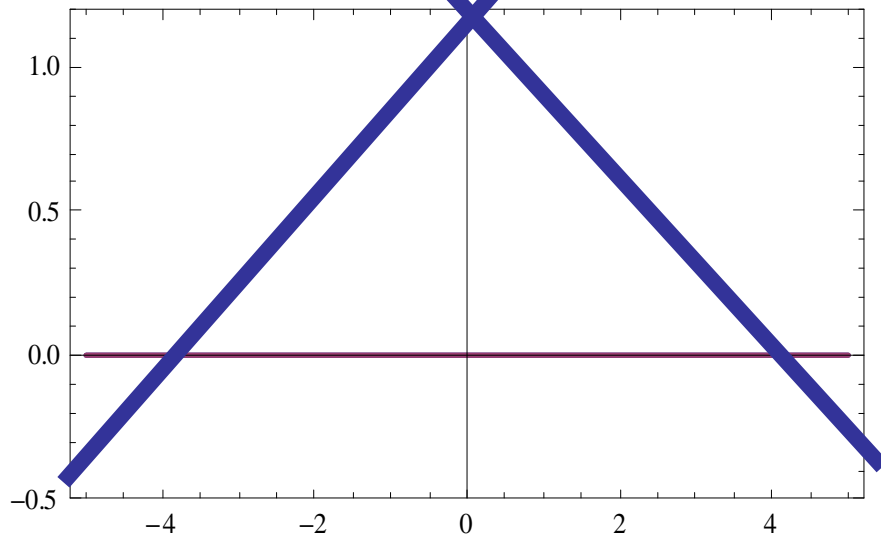
Lévy – Smirnov distribution

REAL PART

$$\phi_X(\vartheta) = \text{Re} \left[e^{-|\vartheta|^{1/2} (1 - i \text{Sign}(\vartheta))} \right]$$

$$\phi_X(\vartheta) \approx \sum_n \frac{(i\vartheta)^n}{n!} E[X^n]$$

**INTEGER MOMENTS DO NOT
EXIST !!!**



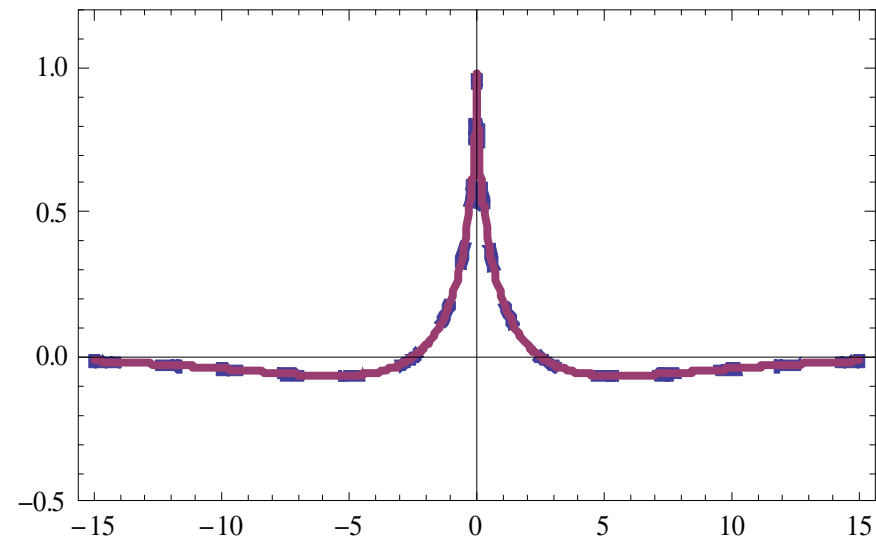
$$\phi_X(\vartheta) = \frac{\Delta}{2\pi} \sum_{k=-10}^{10} \Gamma(\gamma_k) E[(-iX)^{-(\gamma_k)}] (\vartheta)^{-\gamma_k}$$

$$\gamma_k = 1/2 + ik\Delta, \quad \Delta = 0.4$$

21 FRACTIONAL MOMENTS

— exact

— approximated

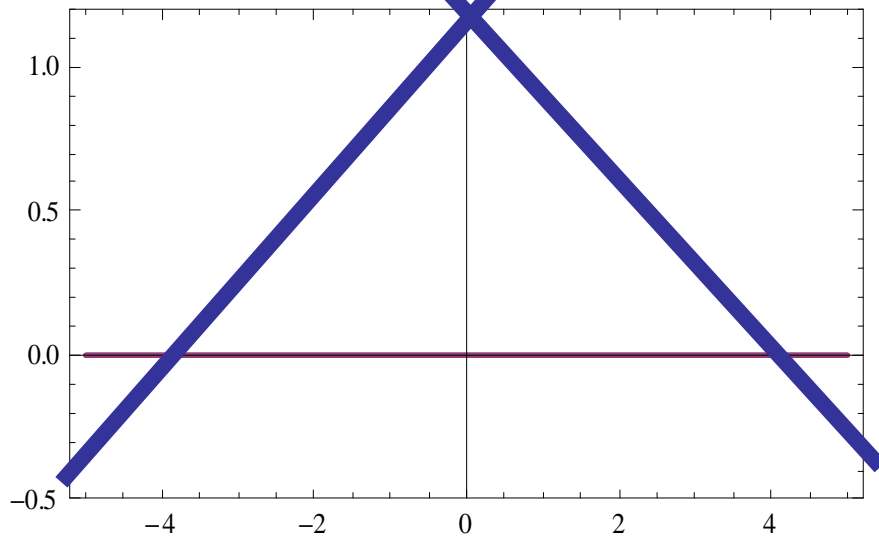


Lévy – Smirnov distribution

$$\phi_X(\vartheta) = \text{Im} \left[e^{-|\vartheta|^{1/2} (1 - i \text{Sign}(\vartheta))} \right]$$

$$\phi_X(\vartheta) \approx \sum_{n=0}^4 \frac{(i\vartheta)^n}{n!} E[X^n]$$

INTEGER MOMENTS DO NOT EXIST !!!



IMAGINARY PART

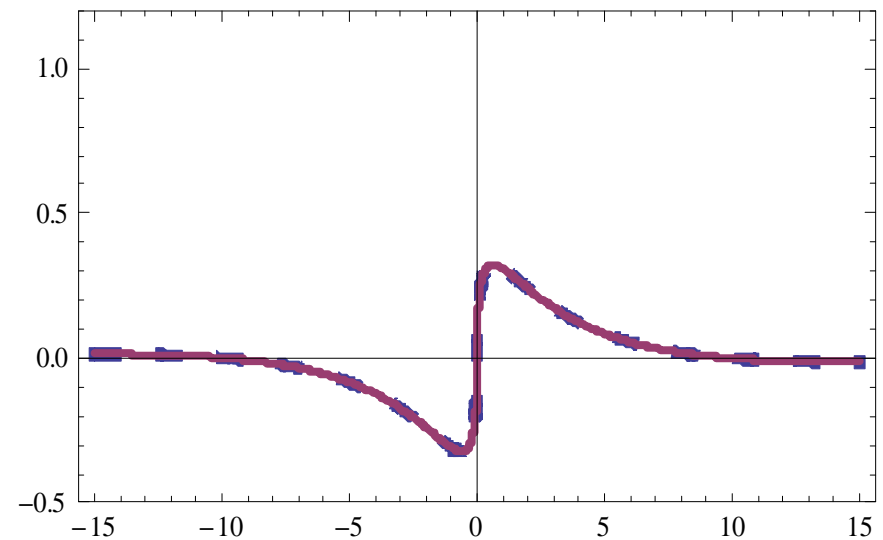
$$\phi_X(\vartheta) = \frac{\Delta}{2\pi} \sum_{k=-2}^2 \Gamma(\gamma_k) E[(-iX)^{-(\gamma_k)}] (\vartheta)^{-\gamma_k}$$

$$\gamma_k = 1/2 + ik\Delta, \quad \Delta = 0.4$$

21 FRACTIONAL MOMENTS

— exact

— approximated



Density by complex moments

In the case of integer moments the well-known expressions are valid

$$\phi_X(\vartheta) = \sum_{j=0}^{\infty} \frac{(i\vartheta)^j E[X^j]}{j!} \quad \boxed{\text{FT}} \quad p_X(x) = \sum_{j=0}^{\infty} (-1)^j \frac{E[X^j]}{j!} \frac{d^j \delta(x)}{dx^j}$$

In the case of fractional moments of complex order

$$\phi_X(\pm\vartheta) = \frac{1}{2\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Gamma(\gamma) E[(\mp iX)^{-\gamma}] |\vartheta|^{-\gamma} d\gamma$$

By making Inverse Fourier transform

$$p_X(x) = \frac{1}{(2\pi)^2 i} \int_{\rho-i\infty}^{\rho+i\infty} \Gamma(\gamma) \Gamma(1-\gamma) \times \\ \times \left\{ E[(-iX)^{-\gamma}] (ix)^{\gamma-1} + E[(iX)^{-\gamma}] (-ix)^{\gamma-1} \right\} d\gamma$$

As the CF has symmetry properties, it simplifies

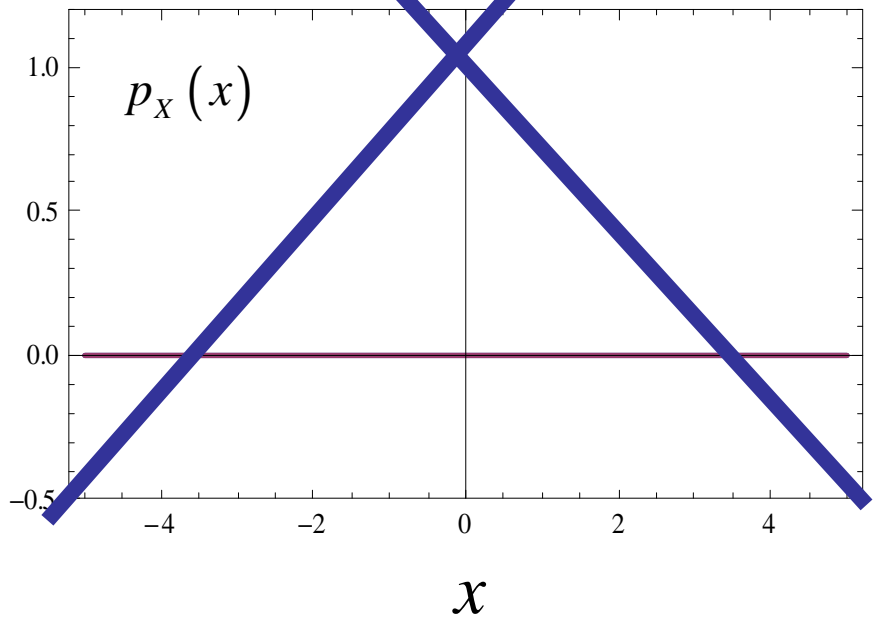
$$p_X(x) = \frac{1}{2\pi^2 i} \operatorname{Re} \left\{ \int_{\rho-i\infty}^{\rho+i\infty} \Gamma(\gamma) \Gamma(1-\gamma) E[(-iX)^{-\gamma}] (ix)^{\gamma-1} d\gamma \right\}$$

PDF Lévy-Smirnov random variable

$$p_X(x) = (1/2\pi)^{1/2} (x)^{-3/2} e^{-\frac{1}{2x}}$$

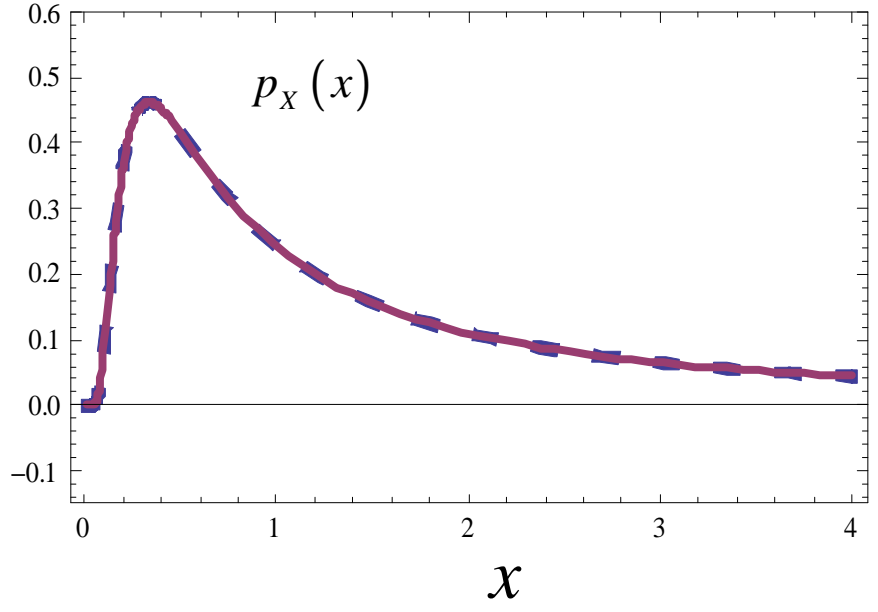
Integer moments do not exist!

$$p_X(x) = \sum_{j=0}^{\infty} (-1)^j \frac{E[X^j]}{j!} \frac{d^j \delta(x)}{dx^j}$$



21 FRACTIONAL MOMENTS

— exact
— approximated



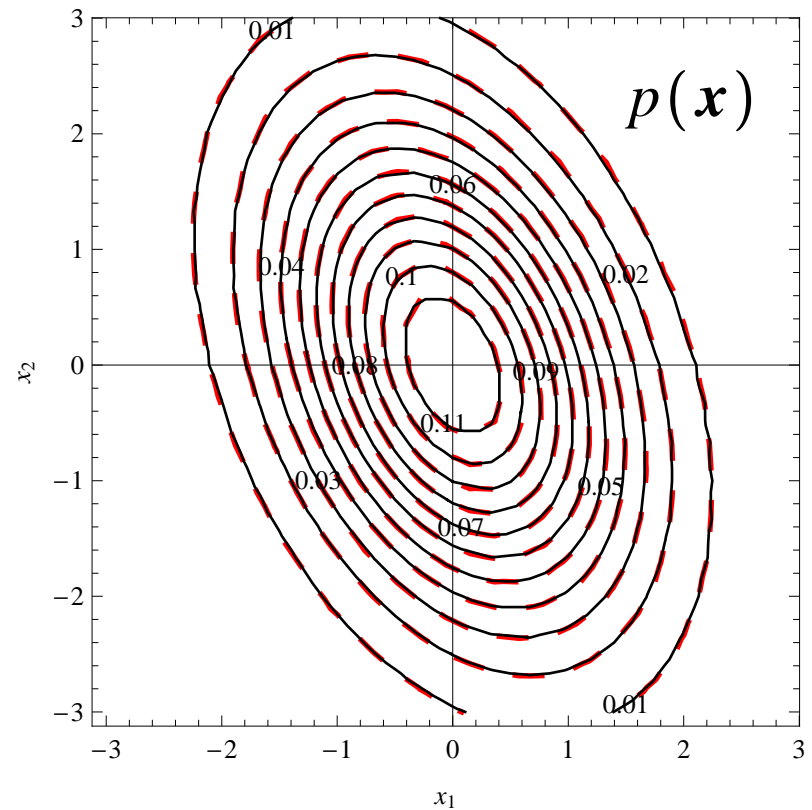
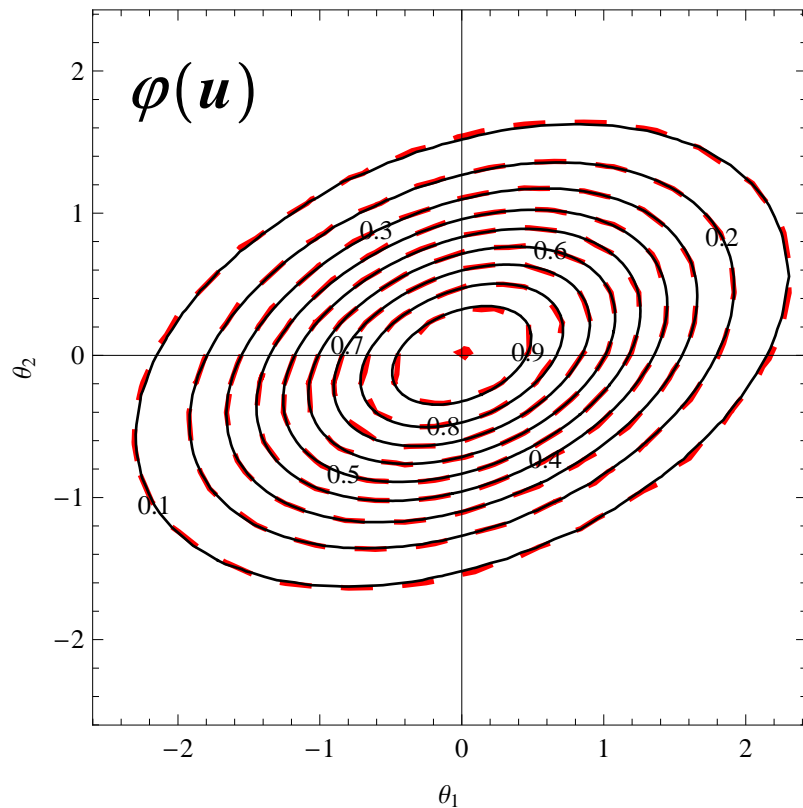
Bivariate Gaussian Random Vector

$$p(\mathbf{x}) = \left[(2\pi)^2 \det(\boldsymbol{\Sigma}) \right]^{-1/2} \exp \left[-\frac{1}{2} \mathbf{x}^T (\boldsymbol{\Sigma})^{-1} \mathbf{x} \right]$$

$$\varphi(\mathbf{u}) = \exp \left[-\frac{1}{2} \mathbf{u}^T \boldsymbol{\Sigma} \mathbf{u} \right]$$

$$\boldsymbol{\Sigma} = \begin{pmatrix} 1 & .5 \\ .5 & 2 \end{pmatrix}; \boldsymbol{\mu} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

— exact
- - - - - Approximated



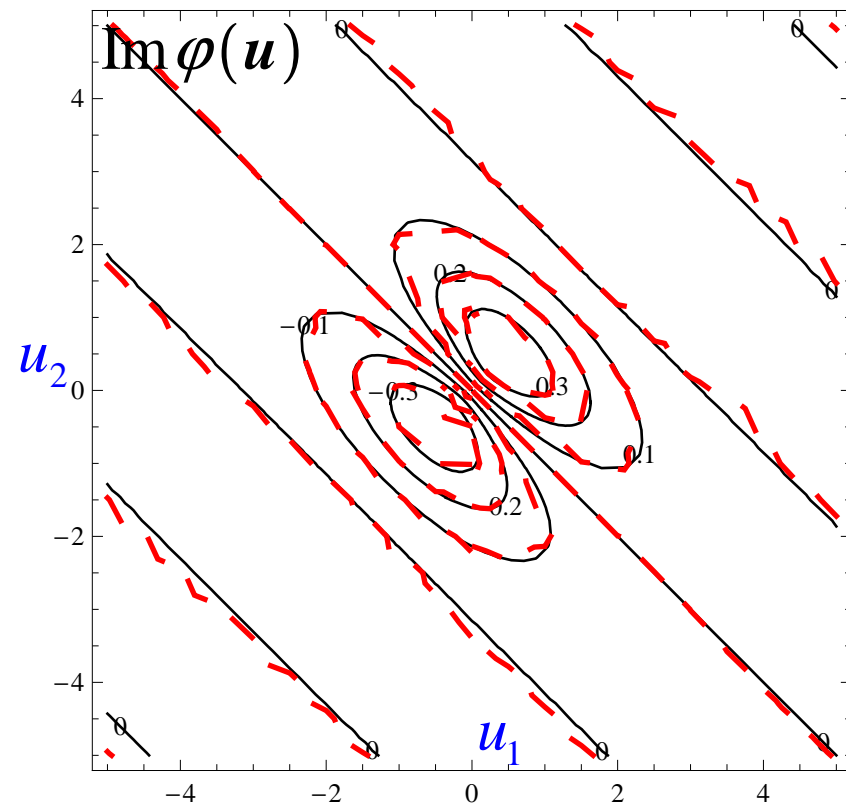
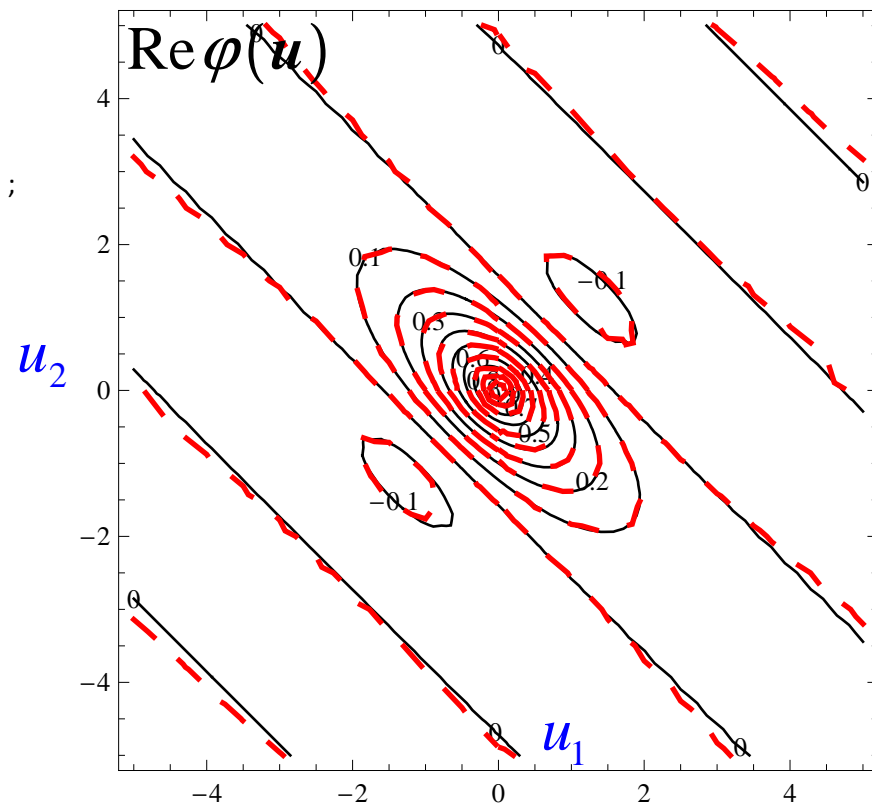
Bivariate Cauchy Random Vector

$$p(\mathbf{x}) = \frac{c \det(\boldsymbol{\Sigma})^{-1/2}}{(1 + (\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1} (\mathbf{x} - \boldsymbol{\mu}))^{3/2}}$$

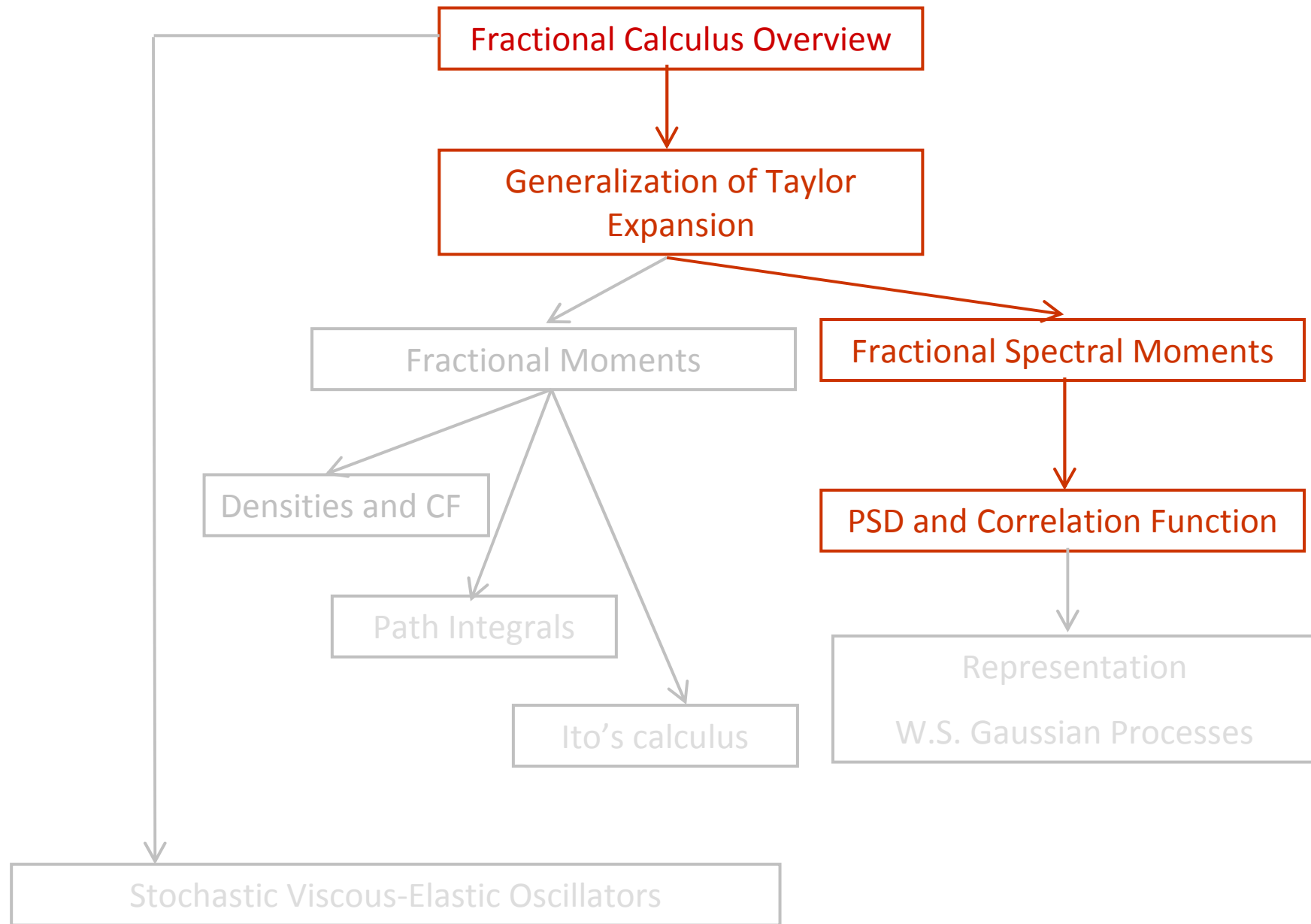
$$\boldsymbol{\Sigma} = \begin{pmatrix} 1 & 0.2 \\ 0.2 & 1 \end{pmatrix}; \boldsymbol{\mu} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\varphi(\mathbf{u}) = \exp(-\mathbf{u}^T \boldsymbol{\Sigma} \mathbf{u})^{1/2} + i \mathbf{u}^T \boldsymbol{\mu}$$

— exact
 - - - approximated



OUTLINE



Application of Fractional calculus to the representation of Stochastic processes

Define the analytical process

$$X(t) = [Y(t) + i\hat{Y}(t)] / \sqrt{2}$$

Hilbert transform

$$\hat{Y}(t) = \frac{1}{\pi} \mathcal{P} \int_{-\infty}^{\infty} \frac{Y(\rho)}{t - \rho} d\rho$$

Correlation Function of $X(t)$

$$R_X(\tau) = R_Y(\tau) + i\hat{R}_Y(\tau)$$

One-sided Power Spectral Density

$$S_X(\omega) = 2U(\omega) S_Y(\omega)$$

Spectral moments of the analytical process (Vanmarke)

$$\lambda_X^j = \int_0^\infty 2U(\omega) S_Y(\omega) \omega^j d\omega = \int_0^\infty S_X(\omega) \omega^j d\omega$$

$$\lambda_X^{2n} = E \left[\frac{d^n X(t)}{dt^n} \frac{d^n X^*(t)}{dt^n} \right]$$

$$i\lambda_X^{2n+1} = E \left[\frac{d^{n-1} X(t)}{dt^{n-1}} \frac{d^n X^*(t)}{dt^n} \right]$$

Fractional spectral moments of the analytical process

$$\Lambda_X(\gamma) = \int_0^\infty S_X(\omega) \omega^\gamma d\omega \quad \gamma = \rho + i\eta$$

 Mellin transform of $S_X(\omega)$

INVERTIBLE

Two important results:

$$(\mathcal{I}^\gamma R_Y)(0) = \int_0^\infty \omega^{-\gamma} S_X(\omega) d\omega = \Lambda_X(-\gamma)$$

$$(\mathcal{D}^\gamma R_Y)(0) = \int_0^\infty \omega^\gamma S_X(\omega) d\omega = \Lambda_X(\gamma)$$

Correlation function

$$R_Y(s) = \frac{1}{2\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \nu(\gamma) \Lambda_X(-\gamma) |s|^{-\gamma} d\gamma$$

Power Spectral Density

$$S_Y(\omega) = \frac{1}{4\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Lambda_X(-\gamma) |\omega|^{\gamma-1} d\gamma$$

Exact Form

Approximate Form

$$R_Y(s) \cong \frac{\Delta\eta}{2\pi} \sum_{k=-m}^m \nu(\gamma_k) \Lambda_X(-\gamma_k) |s|^{-\gamma_k}$$

$$S_Y(\omega) \cong \frac{\Delta\eta}{4\pi} \sum_{k=-m}^m \Lambda_X(-\gamma_k) |\omega|^{\gamma_k-1}$$

Applications to Wind Engineering

Davenport's spectrum

$$S_V(\omega) = \frac{4\pi k_0 V_{ref}^2}{|\omega|} \frac{q(\omega)^2}{(1 + q(\omega)^2)^{4/3}}$$

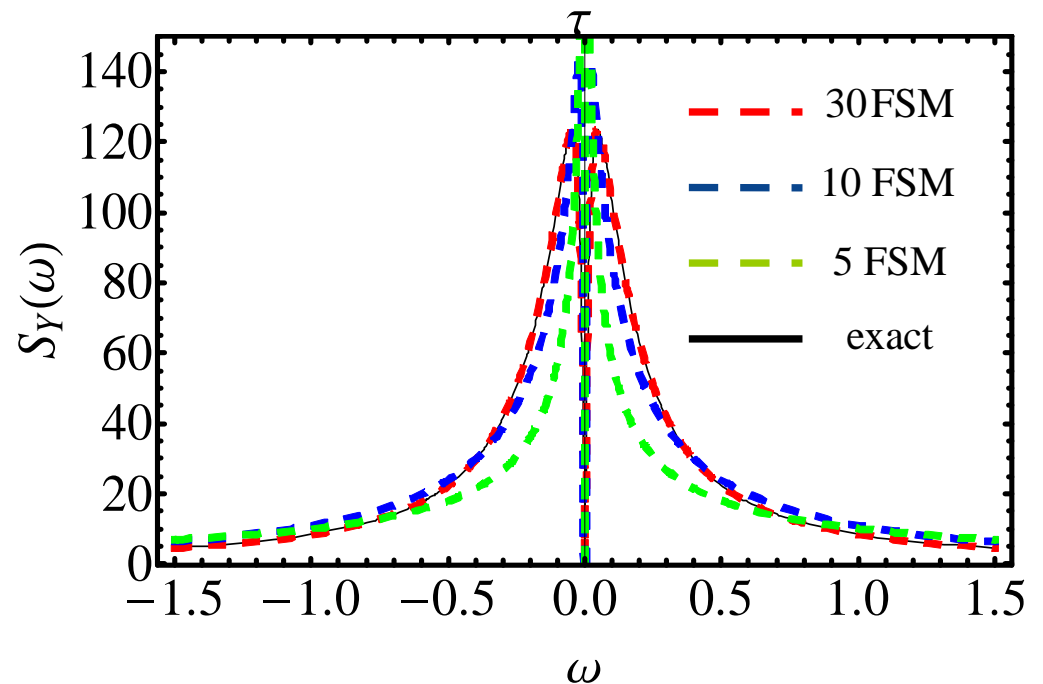
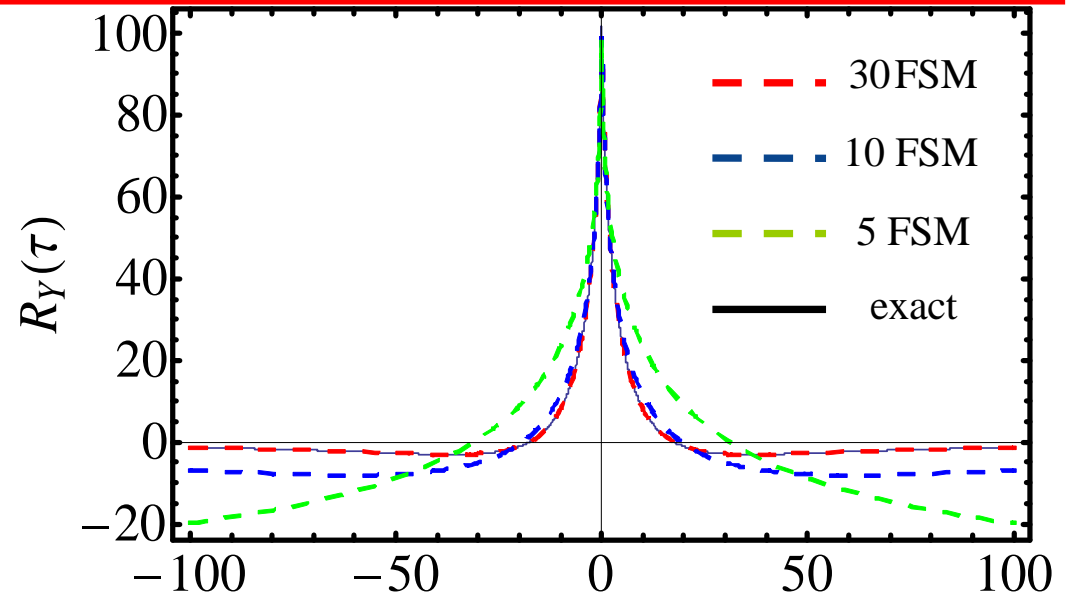
V_{ref} : mean wind speed at the reference level

k_0 : roughness characteristic of the analyzed site

$$q(\omega) = 1200\omega / (2\pi V_{ref})$$

$$R_Y(s) \cong \frac{\Delta\eta}{2\pi} \sum_{k=-m}^m \nu(\gamma_k) \Lambda_X(-\gamma_k) |s|^{-\gamma_k}$$

$$S_Y(\omega) \cong \frac{\Delta\eta}{4\pi} \sum_{k=-m}^m \Lambda_X(-\gamma_k) |\omega|^{\gamma_k-1}$$



Applications to Wind Engineering

Kaimal's spectrum

$$S_V(\omega) = \frac{200v^* z_0}{4\pi v} \left(\frac{1}{1 + 50|\omega|z_0 / (2\pi v)} \right)^{5/3}$$

z_0 : height of the measured mean wind speed

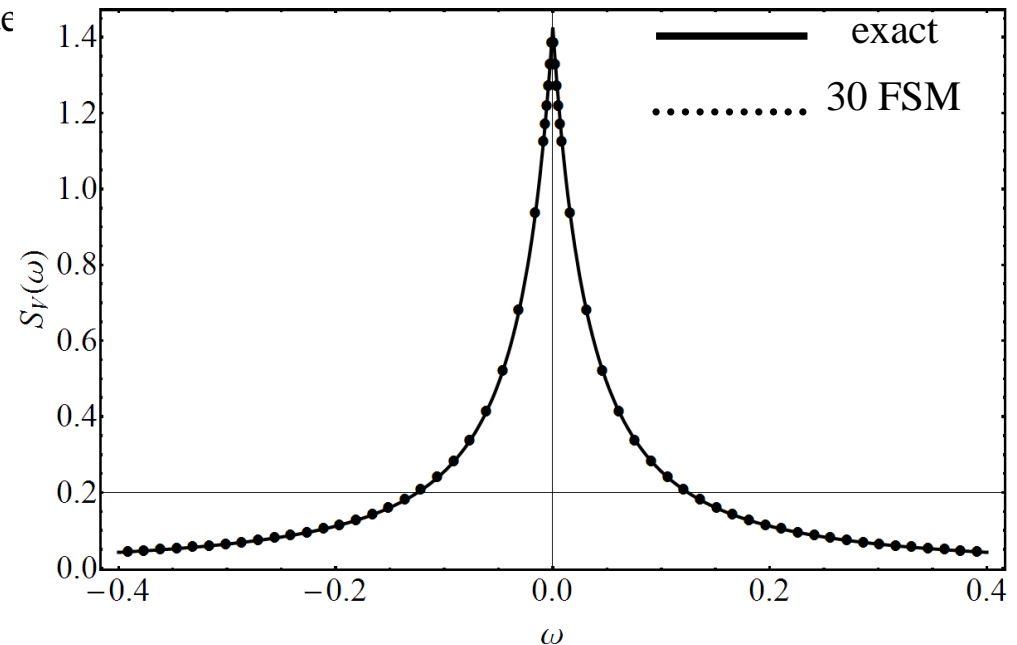
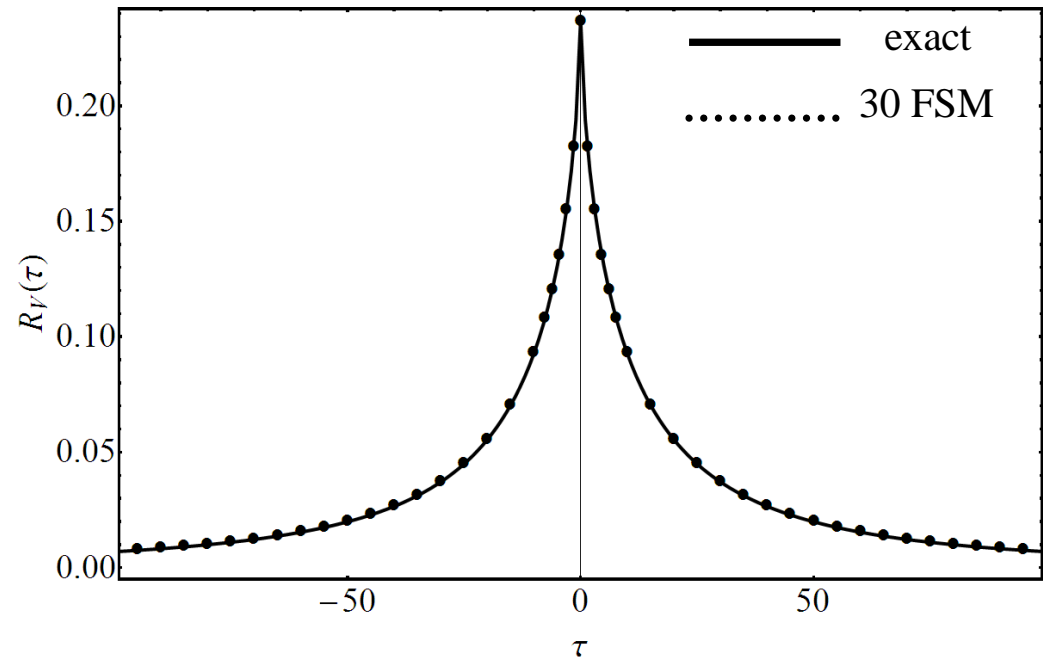
v : mean wind speed

v^* : $0.4v / \ln(z_0 / k)$

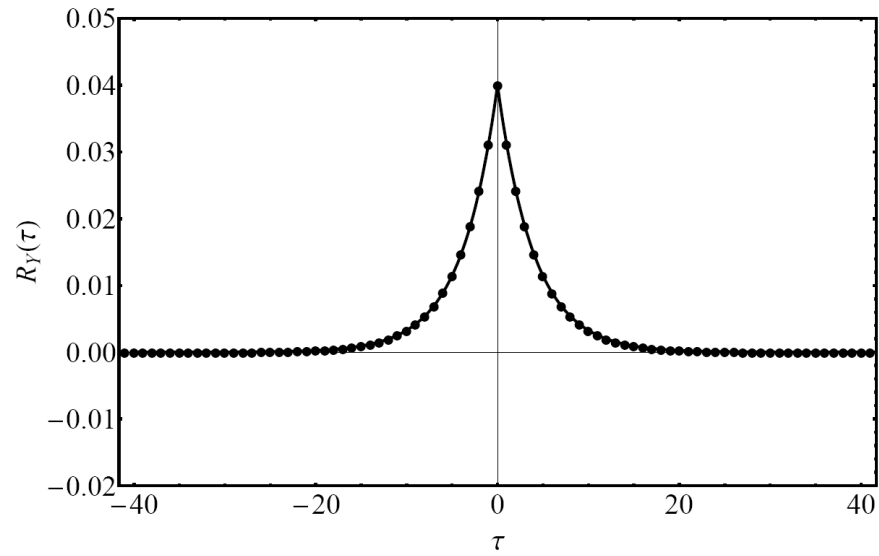
k : roughness coefficient related to the exposed site

$$R_Y(s) \cong \frac{\Delta\eta}{2\pi} \sum_{k=-m}^m v(\gamma_k) \Lambda_X(-\gamma_k) |s|^{-\gamma_k}$$

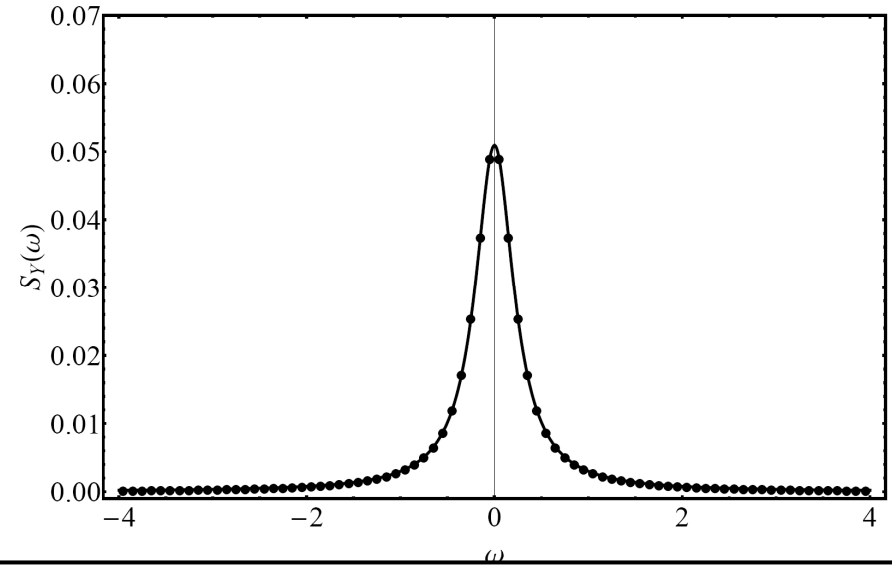
$$S_Y(\omega) \cong \frac{\Delta\eta}{4\pi} \sum_{k=-m}^m \Lambda_X(-\gamma_k) |\omega|^{\gamma_k-1}$$



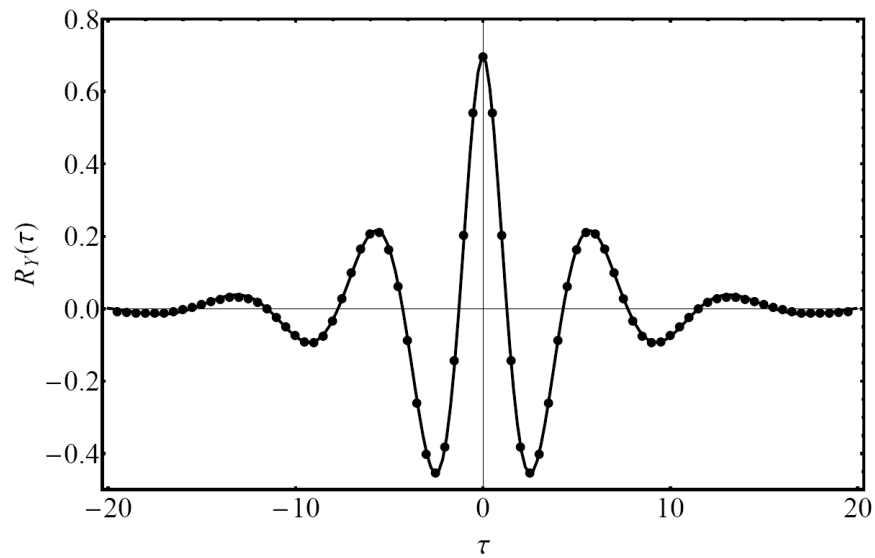
Exp -Correlation Function



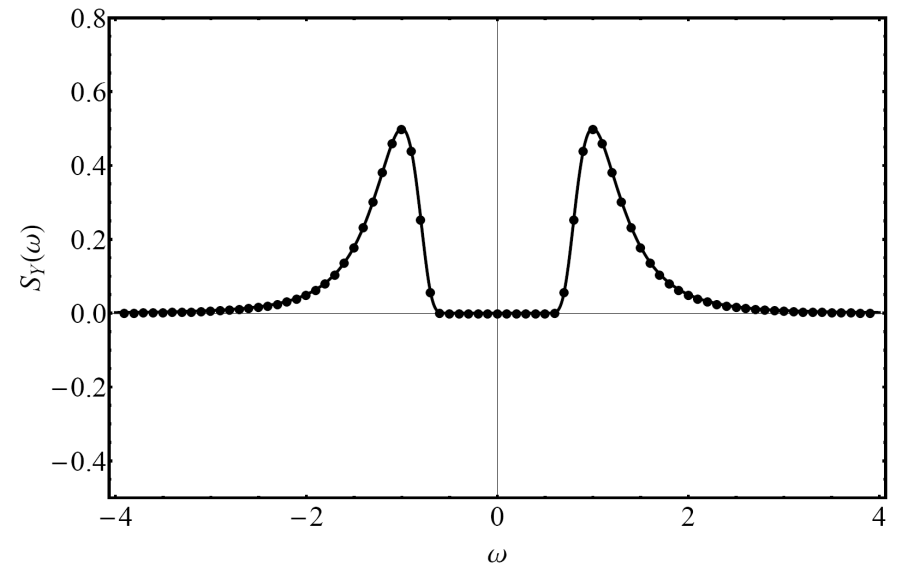
Power Spectral Density



PM -Correlation Function



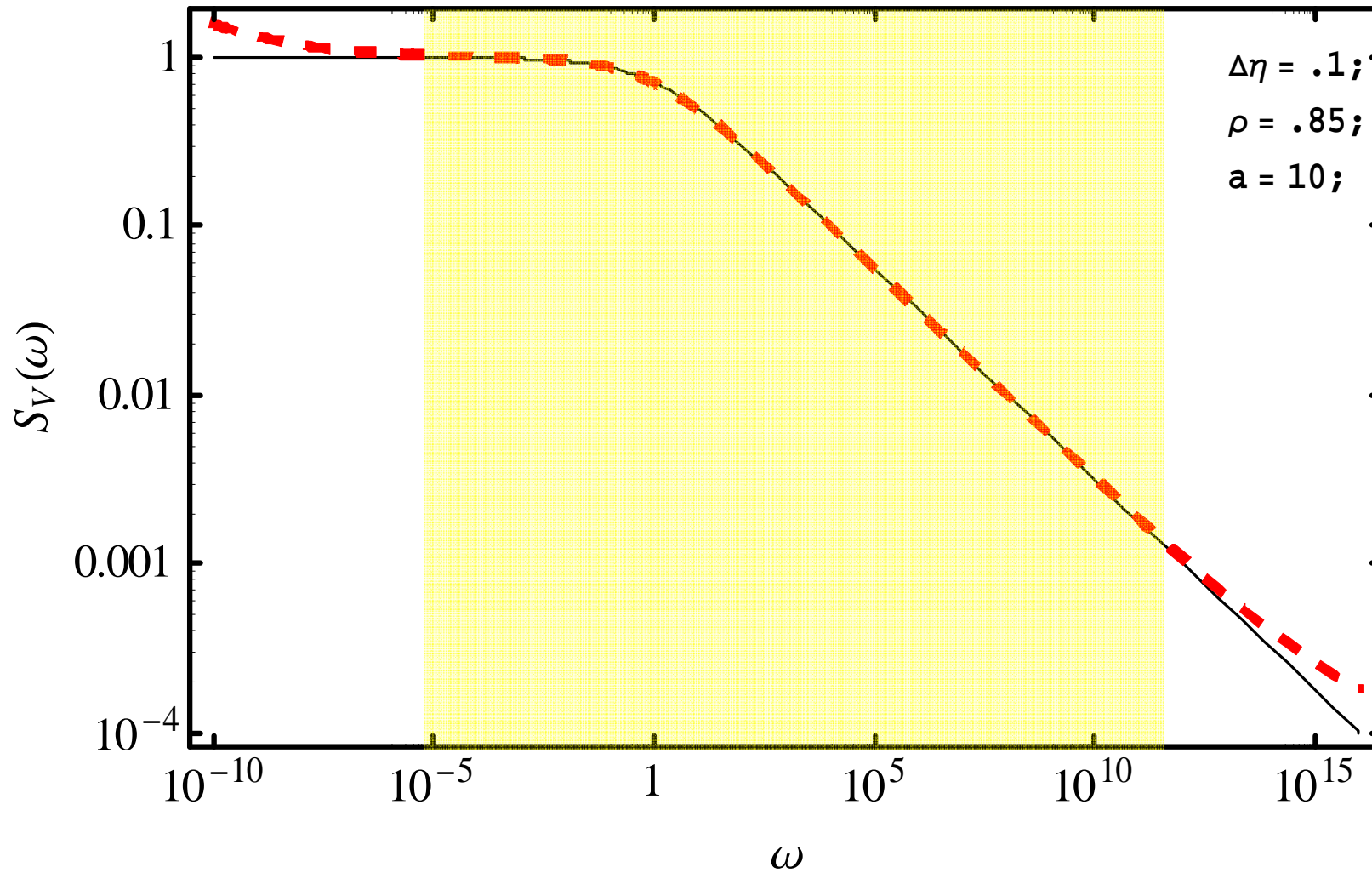
Power Spectral Density



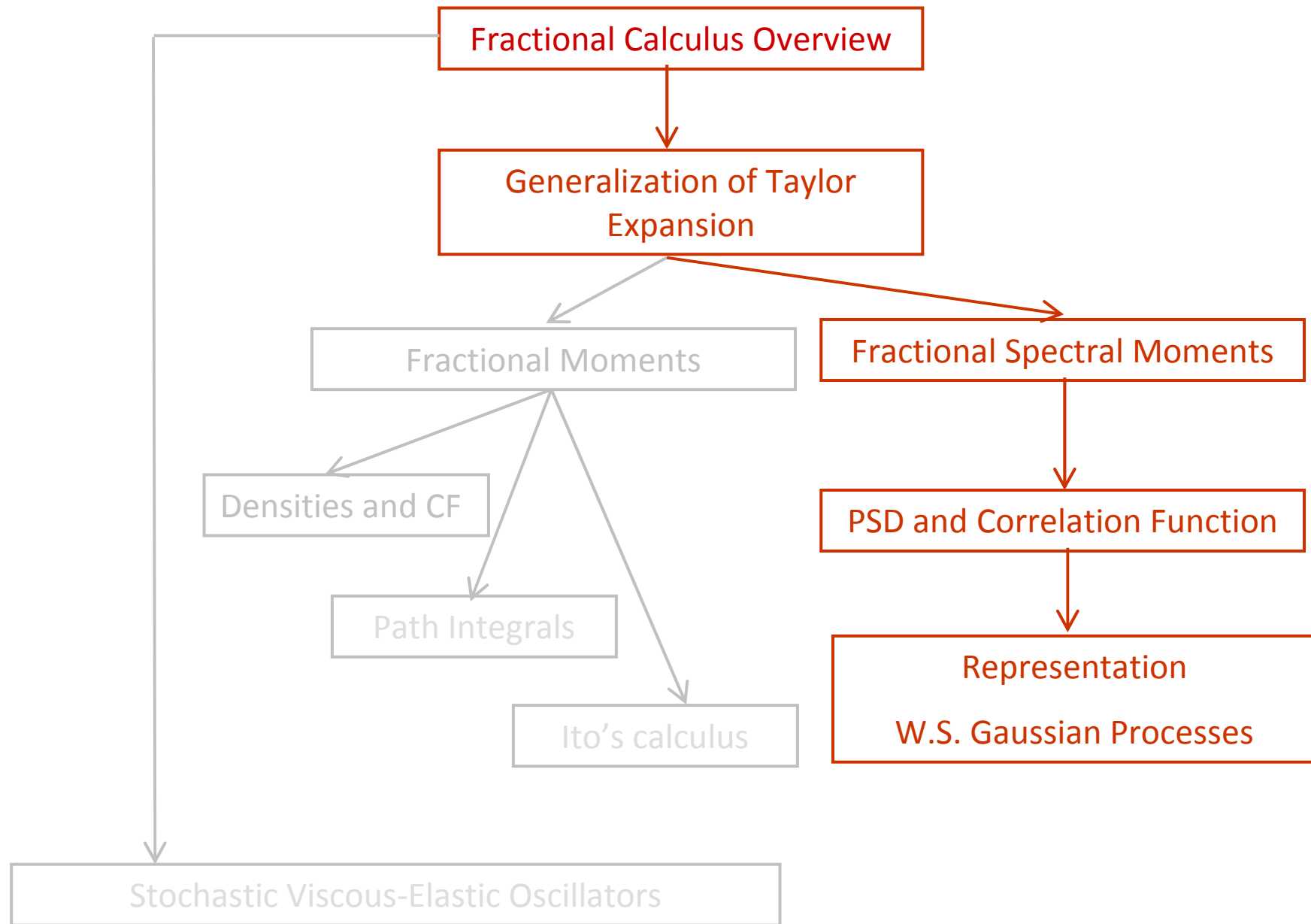
Long-Term Correlation Function

$$S_V(\omega) = \frac{1}{1 + |\omega|^{1/2}}$$

$$R(\tau) \approx \tau^{-1/2}$$



OUTLINE



Generation of weakly stationary Gaussian coloured noises (H-FSM)

GOAL: Process $F(t)$ with target $S_F(\omega)$ as output of linear system

$$\mathcal{L}(F(t)) = W(t)$$

→ **Gaussian White Noise**

In the frequency domain the Power Spectral Density of the process $F(t)$ can be obtained as:

$$S_F(\omega) = |H(\omega)|^2 S_W(\omega) = \frac{q}{2\pi} |H(\omega)|^2$$

The transfer function is the Fourier transform of the Impulse response function $h(t)$

Generation of weakly stationary Gaussian coloured noises (H-FSM)

Fractional Spectral Moments of the transfer function:

$$\Pi_H(\gamma) \stackrel{\text{def}}{=} 2 \int_0^\infty |\omega|^\gamma H(\omega) d\omega, \quad \text{Re}\gamma > 0$$

H-FSM are Riesz fractional integrals and derivatives of the impulse response function $h(t)$, evaluated in zero.

By inverse Mellin transform:

$$H(\omega) = \frac{1}{4\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Pi_H(-\gamma) |\omega|^{\gamma-1} d\gamma, \quad \gamma = \rho + i\eta, \quad 0 < \rho < 1$$

Approximating by truncated series:

$$H(\omega) \cong \frac{\Delta\eta}{4\pi} \sum_{k=-m}^m \Pi_H(-\gamma_k) |\omega|^{\gamma_k-1} \quad \gamma_k = \rho + ik\Delta\eta$$

Generation of weakly stationary Gaussian coloured noises (H-FSM)

From System's Linearity and Inverse Fourier it follows

$$F(t) = \frac{1}{4\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \Pi_H(-\gamma) (I^{1-\gamma} W)(t) d\gamma$$

Fractional moments of
the system's transfer function

Fractional Brownian motions

Riesz Fractional Integral

$$(I^\gamma W)(t) \propto (I_-^\gamma W + I_+^\gamma W)(t)$$

Discrete form

$$F(t) = \frac{\Delta\eta}{4\pi} \sum_{k=-m}^m \Pi_H(-\gamma_k) (I^{1-\gamma_k} W)(t)$$

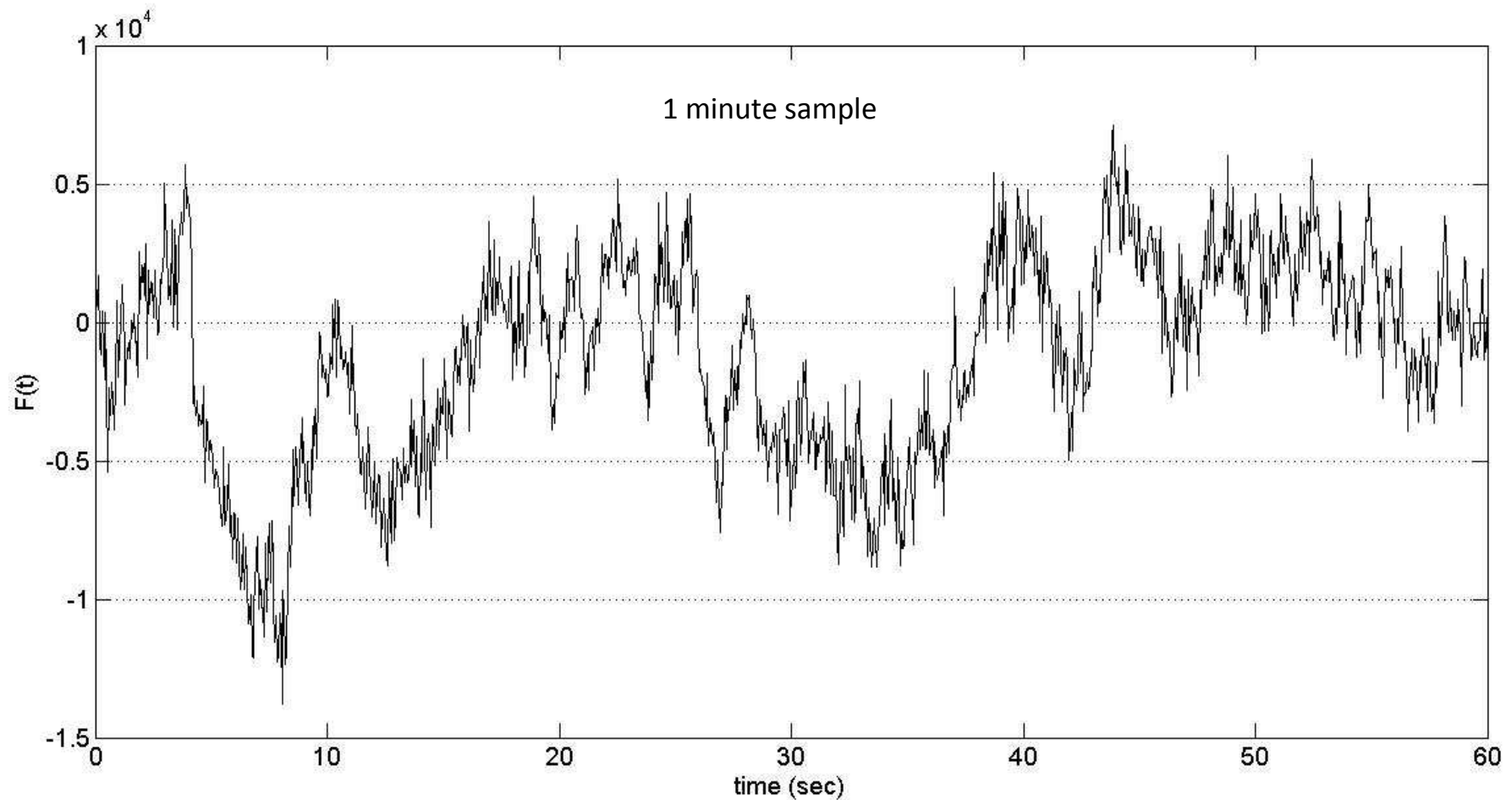
Application to wind loads

$$\mathbf{F} = \sum_{k=-m}^m \beta_k \mathbf{B}_{1-\gamma_k}$$

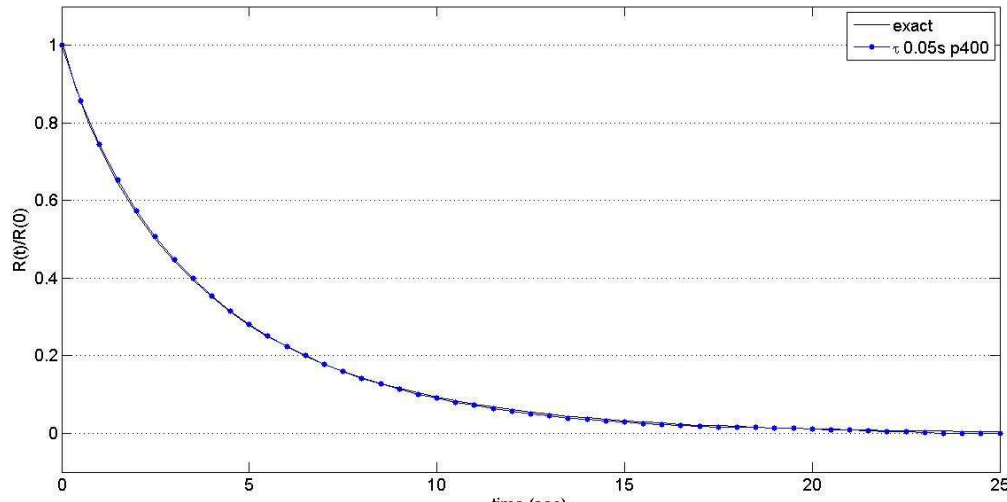
$\gamma_k = \rho + ik\Delta\eta$

↓
H-FSM

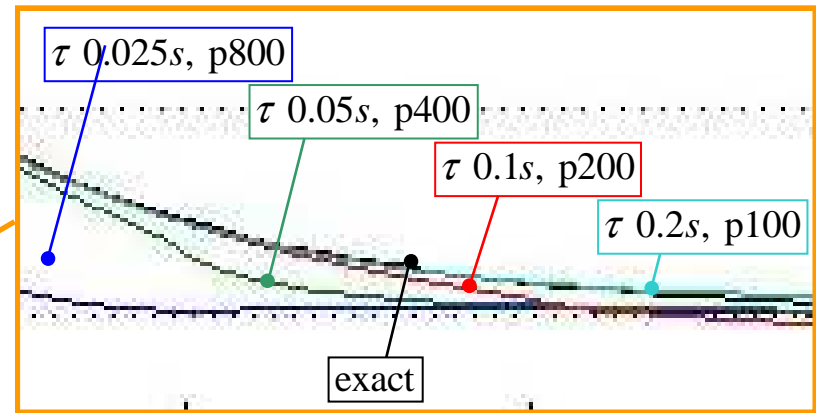
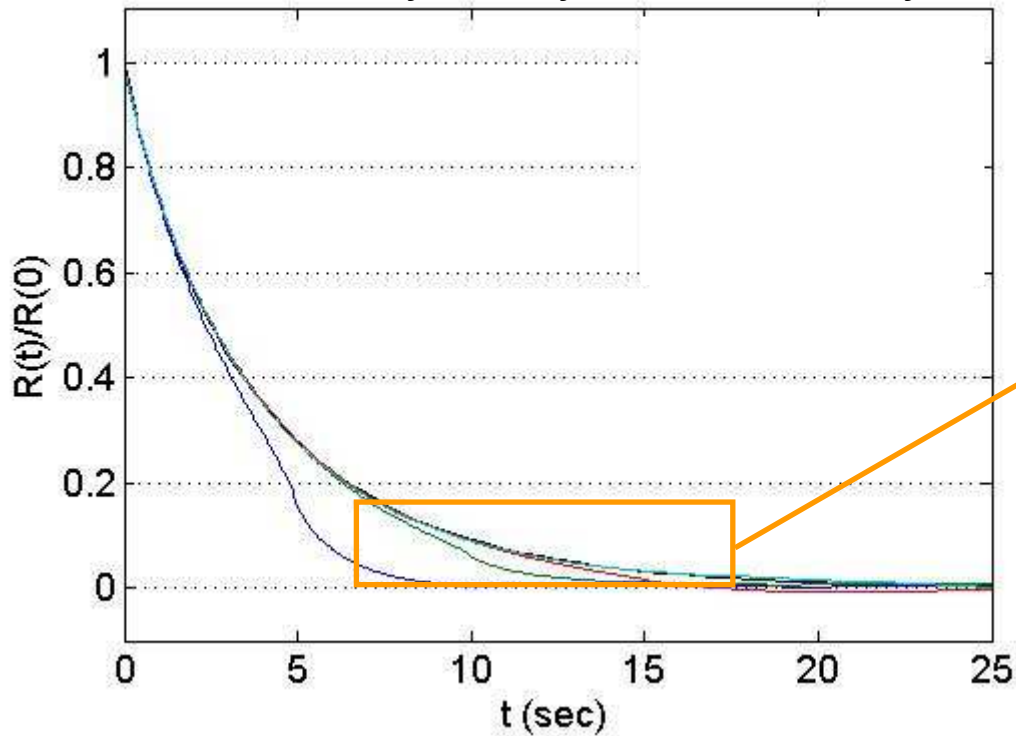
↘ fBm



Auto-correlation of generated wind data: Memory Propagation



Autocorrelation function for constant value of t and different non zero coefficients α_k



Extension: Multivariate processes

$$\mathbf{S}_V(\omega) = \begin{bmatrix} S_{V_1 V_1}(\omega) & S_{V_1 V_2}(\omega) & \dots & S_{V_1 V_N}(\omega) \\ S_{V_1 V_2}(\omega) & S_{V_2 V_2}(\omega) & \dots & \dots \\ \dots & \dots & \dots & \dots \\ S_{V_1 V_N}(\omega) & S_{V_2 V_N}(\omega) & \dots & S_{V_N V_N}(\omega) \end{bmatrix}$$

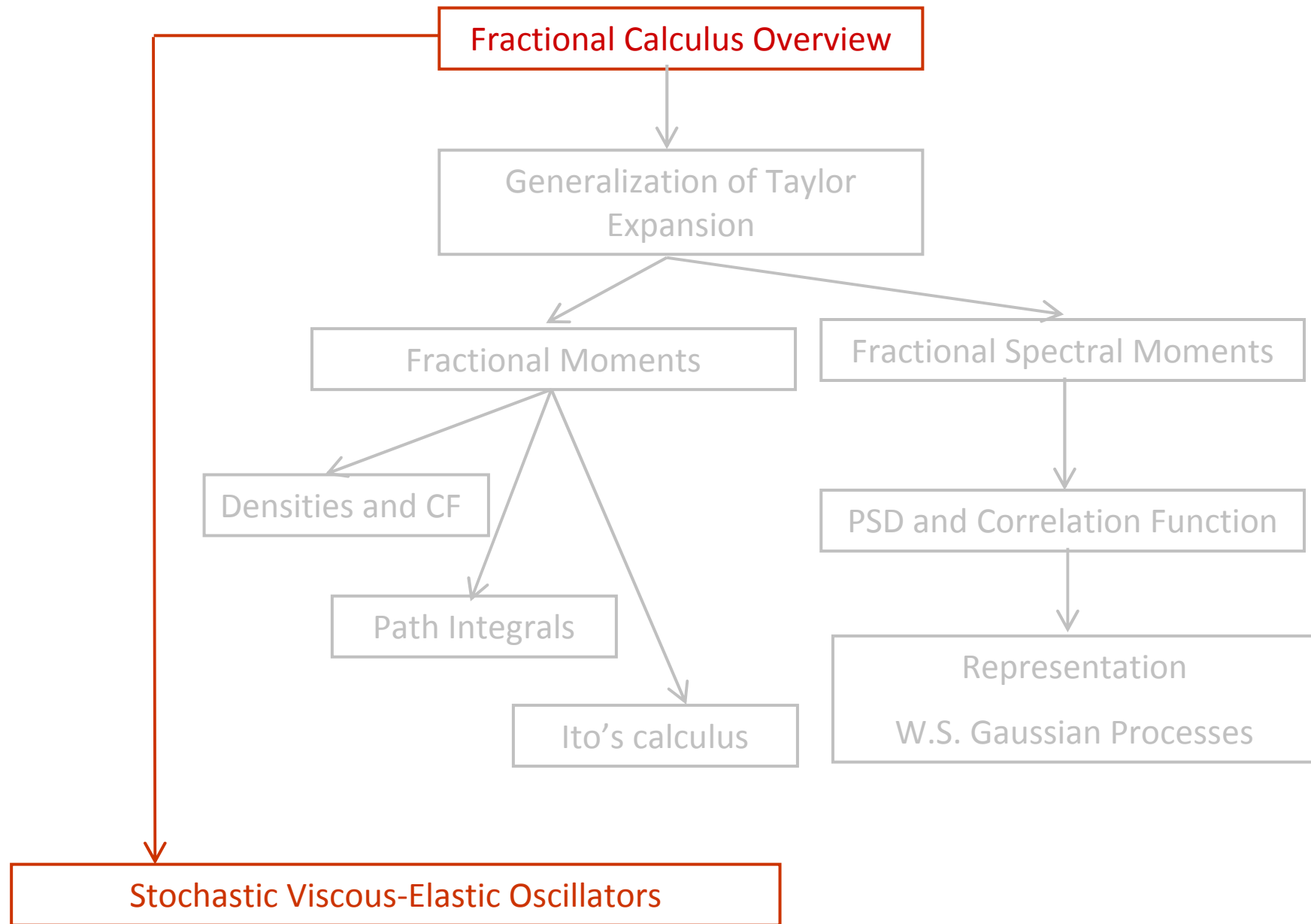
$$\mathbf{S}_V(\omega) = \mathbf{\Psi}(\omega) \mathbf{L}^{1/2}(\omega) \mathbf{L}^{1/2}(\omega) \mathbf{\Psi}^{*T}(\omega)$$

$$\mathbf{H}(\omega) = \mathbf{\Psi}(\omega) \mathbf{L}^{1/2}(\omega)$$

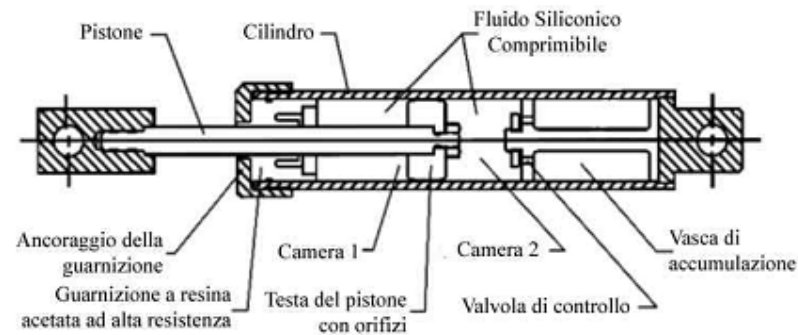
$$\mathbf{\Pi}(\gamma) \stackrel{def}{=} \int_{-\infty}^{\infty} |\omega|^\gamma \mathbf{H}(\omega) d\omega$$

$$\mathbf{V}(t) = \frac{1}{4\pi i} \int_{\rho-i\infty}^{\rho+i\infty} \mathbf{\Pi}(-\gamma) (I^{1-\gamma} \mathbf{W})(t) d\gamma$$

OUTLINE



APPLICATION: STOCHASTIC VISCOUS-ELASTIC SYSTEMS



Constitutive equation of viscous-elastic dampers:

$$f(\dot{X}(t)) = c_d |\dot{X}|^\gamma \text{sign}(\dot{X}) \quad \gamma \in \mathbb{R}$$

Equation of motion

$$m\ddot{X}(t) + f(\dot{X}(t)) + kX(t) = W_\alpha(t)$$

Non-linear Stochastic Differential Equation subjected to Lévy noise

APPLICATION: STOCHASTIC VISCOUS-ELASTIC SYSTEMS

$$m \ddot{X}(t) + c_d \left| \dot{X}(t) \right|^{\gamma} \text{sign}(\dot{X}(t)) + k X(t) = W_{\alpha}(t) \quad \begin{array}{l} \gamma \in \mathbb{R} \\ \gamma \neq 0, 2, 4, \dots \end{array}$$

Spectral Einstein-Smoluchowski equation

$$\frac{\partial \phi_{\mathbf{Z}}(\boldsymbol{\theta}, t)}{\partial t} = \vartheta_1 \frac{\partial \phi_{\mathbf{Z}}(\boldsymbol{\theta}, t)}{\partial \vartheta_2} - \frac{k}{m} \vartheta_2 \frac{\partial \phi_{\mathbf{Z}}(\boldsymbol{\theta}, t)}{\partial \vartheta_1} + \frac{c_d}{m} \vartheta_2^{\gamma} H^{-\gamma}(\phi_{\mathbf{Z}}(\boldsymbol{\theta}, t)) - \left(\frac{|\vartheta_2|}{m} \right)^{\alpha} \phi_{\mathbf{Z}}(\boldsymbol{\theta}, t)$$

Einstein-Smoluchowski equation

$$\frac{\partial p_{\mathbf{Z}}(\mathbf{z}, t)}{\partial t} = -z_2 \frac{\partial p_{\mathbf{Z}}(\mathbf{z}, t)}{\partial z_1} + \frac{k}{m} z_1 \frac{\partial p_{\mathbf{Z}}(\mathbf{z}, t)}{\partial z_2} + \frac{c_d}{m} \frac{\partial}{\partial z_2} \left(|z_2|^{\gamma} \text{sign}(z_2) p_{\mathbf{Z}}(\mathbf{z}, t) \right) + \frac{1}{m^{\alpha}} \left({}_{z_2} \mathcal{D}^{\alpha} p_{\mathbf{Z}}(\mathbf{z}, t) \right)$$

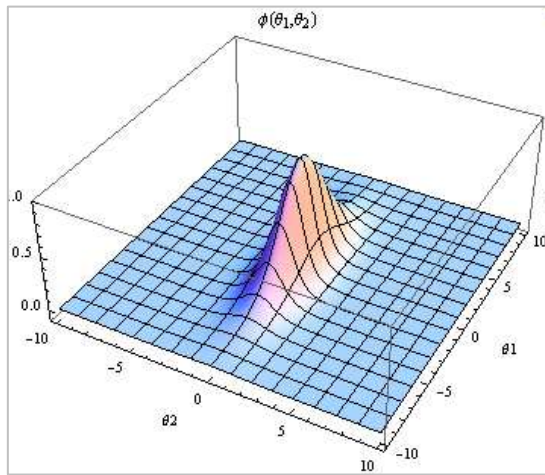
where $({}_{z_2} \mathcal{D}^{\alpha} p_{\mathbf{Z}}(\mathbf{z}, t))$ is the partial Riesz derivative of the joint PDF

Deterministic Fractional Differential Equations

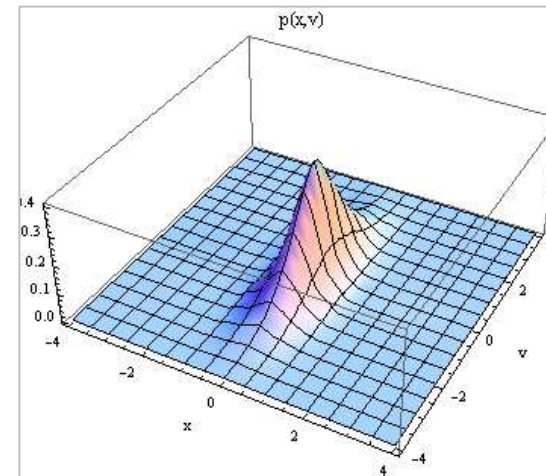
APPLICATION: STOCHASTIC VISCOUS-ELASTIC SYSTEMS

$$\alpha = 2, \quad \gamma = 0.5$$

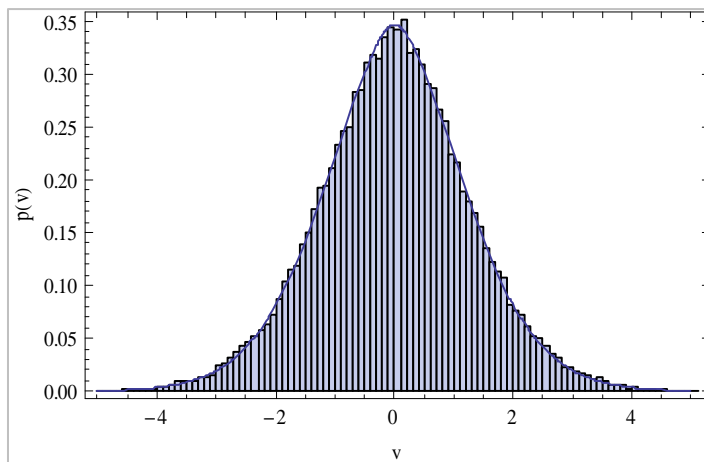
Characteristic function



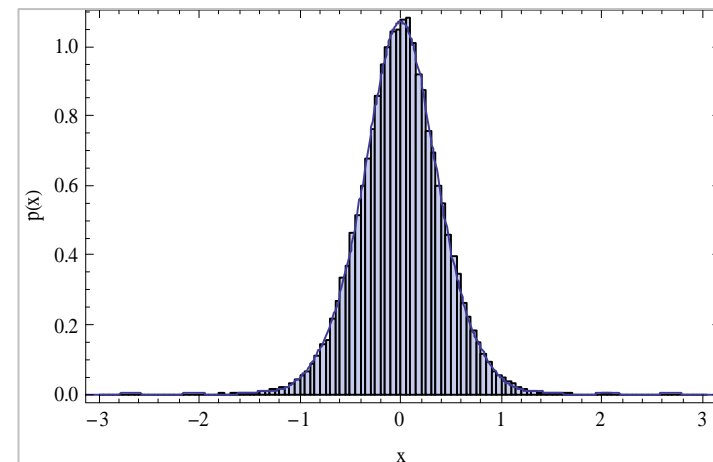
Probability Density function



Marginal probability density function



Marginal probability density function



Thank you ! *

Giulio Cottone

*any inaccuracy should be ascribed to the effect of German Coffee on Italian brain

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References at: www.researcherid.com/rid/B-3565-2009