

Mittag-Leffler Functions, Fourier Transforms, and Stochastic Processes

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1 Mittag-Leffler Functions

1.1 One-Parameter Mittag-Leffler Function

The one-parameter Mittag-Leffler function $E_\alpha(z)$ is defined as:

$$E_\alpha(z) = \sum_{k=0}^{\infty} \frac{z^k}{\Gamma(\alpha k + 1)} \quad (1)$$

where $\alpha > 0$ and z is a complex number.

1.2 Two-Parameter Mittag-Leffler Function

The two-parameter Mittag-Leffler function $E_{\alpha,\beta}(z)$ is defined as:

$$E_{\alpha,\beta}(z) = \sum_{k=0}^{\infty} \frac{z^k}{\Gamma(\alpha k + \beta)} \quad (2)$$

where $\alpha > 0$, β is a complex number, and z is a complex number.

2 Fourier Transforms

2.1 One-Parameter Mittag-Leffler Function

For $0 < \alpha \leq 2$, the Fourier transform of $E_\alpha(-t^\alpha)$ is:

$$\mathcal{F}\{E_\alpha(-t^\alpha)\}(\omega) = \frac{1}{1 + |\omega|^\alpha e^{i\pi\alpha/2\text{sgn}(\omega)}} \quad (3)$$

2.2 Two-Parameter Mittag-Leffler Function

For $0 < \alpha \leq 2$ and β real, the Fourier transform of $t^{\beta-1} E_{\alpha,\beta}(-t^\alpha)$ is:

$$\mathcal{F}\{t^{\beta-1} E_{\alpha,\beta}(-t^\alpha)\}(\omega) = \frac{1}{|\omega|^\alpha} H_{1,2}^{2,1} \left[\frac{1}{|\omega|^\alpha} \middle| \begin{matrix} (1-\beta, \alpha) \\ (0, 1), (1-\beta, 1) \end{matrix} \right] \quad (4)$$

where $H_{1,2}^{2,1}$ is the H-function.

3 Special Cases and Connections

3.1 Cauchy Distribution

When $\alpha = 1$ in the one-parameter case, we get the Cauchy distribution:

$$\mathcal{F}\{E_1(-t)\}(\omega) = \frac{1}{1 + |\omega|} \quad (5)$$

This is related to the Ornstein-Uhlenbeck process.

3.2 Gaussian Distribution

When $\alpha = 2$ in the one-parameter case, we get a Gaussian distribution:

$$\mathcal{F}\{E_2(-t^2)\}(\omega) = \frac{1}{1 + \omega^2} \quad (6)$$

This is related to standard Brownian motion.

4 Orthogonal Polynomials

For the one-parameter case, we can define a weight function:

$$w(\omega) = \frac{1}{1 + |\omega|^\alpha} \quad (7)$$

And an inner product:

$$(f, g) = \int_{-\infty}^{\infty} f(\omega) g(\omega) w(\omega) d\omega \quad (8)$$

Orthogonal polynomials can be constructed using the Gram-Schmidt process on the monomials $1, \omega, \omega^2, \omega^3, \dots$

5 Implications for Stochastic Processes

The Mittag-Leffler function and its Fourier transform provide a framework for modeling a spectrum of processes:

- $\alpha = 1$: Ornstein-Uhlenbeck-like processes
- $1 < \alpha < 2$: Anomalous diffusion
- $\alpha = 2$: Brownian motion
- $0 < \alpha < 1$: Processes with long-range dependence

This framework is particularly useful in fractional calculus and for modeling phenomena that exhibit behavior intermediate between classical diffusion and pure randomness.