Solutions of nonlinear stochastic differential equations with long-range power-law distributions

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Our research is related to the 1/f noise problem and long-range processes.

1/f noise

a type of noise whose power spectral density S(f) behaves like

$$S(f) \sim 1/f^{eta}$$
, eta is close to 1

Fluctuations of signals exhibiting 1/f behavior of the power spectral density at low frequencies have been observed in a wide variety of physical, geophysical, biological, financial, traffic, Internet, astrophysical and other systems.

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Example of 1/f noise



Power spectral density of trading activity (number of trades per 1 \min). for ABT stock on NYSE

- A pure 1/f power spectrum is physically impossible because the total power would be infinity.
- We search for a model where the spectrum of signal has 1/f^β behavior only in some intermediate region of frequencies, f_{min} ≪ f ≪ f_{max}, whereas for small frequencies f ≪ f_{min} the spectrum is bounded.
- The behavior of spectrum at frequencies $f_{\min} \ll f \ll f_{\max}$ is connected with the behavior of the autocorrelation function at times $1/f_{\max} \ll t \ll 1/f_{\min}$.

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- Often 1/f noise is defined by a long-memory process, characterized by S(f) ~ 1/f^β as f → 0.
- This long-range dependence property is equivalent to similar behavior of autocorrelation function C(t) as t → ∞
- This behavior of the autocorrelation function is not necessary for obtaining required form of the power spectrum in a finite interval of the frequencies which does not include zero

- 1/f noise is intermediate between white noise, $S(f) \sim 1/f^0$ and Brownian motion $S(f) \sim 1/f^2$
- In contrast to the Brownian motion generated by the linear stochastic equations, the signals and processes with 1/f spectrum cannot be understood and modeled in such a way.

Goal

to find a simple nonlinear stochastic differential equation (SDE) generating signals exibiting 1/f noise

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- Time series of financial data exhibit highly nontrivial statistical properties. Many of these properties appear to be universal.
- Trading activity, trading volume, and volatility are stochastic variables with the long-range correlation. The autocorrelation of the volatility decays only slowly as a power law.
- Probability distribution functions (PDFs) of return and trading activity have fat tails exhibiting power-law decay.
- Proposed equations can exhibit both power-law PDF and power-law spectrum.

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• If $S(f) \sim f^{-\beta}$ then power spectral density has a scaling property $S(af) = a^{-\beta}S(f)$

• Wiener-Khintchine theorem

$$\mathcal{C}(t) = \int_{-\infty}^{+\infty} \mathcal{S}(f) \cos(2\pi f t) \,\mathrm{d} f$$

• Autocorrelation function C(t) has scaling property

$$C(at) \sim a^{\beta-1}C(t)$$

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• Autocorrelation function can be written as

$$C(t) = \int \mathrm{d}x \int \mathrm{d}x' \, xx' P_0(x) P_x(x',t|x,0)$$

- *P*₀(*x*) is the steady state PDF
- $P_x(x', t|x, 0)$ is the transition probability
- The transition probability can be obtained from the solution of the Fokker-Planck equation with the initial condition $P_x(x', 0|x, 0) = \delta(x' x)$.

Let us assume that

• Steady state PDF has power-law form

$$P_0(x) \sim x^{-\nu}$$

• Trasnsition probability has a scaling property

$$P(ax', t|ax, 0) = a^{-1}P(x', a^{2(\eta-1)}t|x, 0)$$

• Then the autocorrelation function will have the required scaling with

$$\beta = 1 + \frac{\nu - 3}{2(\eta - 1)}$$

To get the required scaling of transition probability:

- SDE will contain only powers of x
- The diffusion coefficient will be of the form $x^{2\eta}$
- The drift term is fixed by the requirement that the steady-state PDF should be $x^{-\nu}$

Proposed SDE

$$\mathrm{d} x = \sigma^2 (\eta - \nu/2) x^{2\eta - 1} \mathrm{d} t + \sigma x^\eta \mathrm{d} W_t$$

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B. Kaulakys and J. Ruseckas, Phys. Rev. E 70, 020101(R) (2004).

B. Kaulakys and J. Ruseckas, V. Gontis, and M. Alaburda, Physica A 365, 217 (2006).

- Because of the divergence of the power-law distribution and the requirement of the stationarity of the process, the SDE should be analyzed together with the appropriate restrictions of the diffusion in some finite interval.
- When diffusion is restricted, scaling properties are only approximate, but 1/f spectrum remains in a wide interval of frequencies.

Possible forms of restriction:

- Reflective boundary conditions at $x = x_{\min}$ and $x = x_{\max}$
- Exponential restriction of the diffusion

$$\mathrm{d}x = \sigma^2 \left(\eta - \frac{\nu}{2} + \frac{m}{2} \left(\frac{x_{\min}}{x} \right)^m - \frac{m}{2} \left(\frac{x}{x_{\max}} \right)^m \right) x^{2\eta - 1} \mathrm{d}t + \sigma x^{\eta} \mathrm{d}W_t$$

Steady state PDF:

$$P_0(x) \sim x^{-\nu} \exp\left(-\left(\frac{x_{\min}}{x}\right)^m - \left(\frac{x}{x_{\max}}\right)^m\right)$$

Restriction of diffusion

q-exponential steady-state PDF

$$\mathrm{d}x = \sigma^2 (\eta - \nu/2) (x + x_0)^{2\eta - 1} \mathrm{d}t + \sigma (x + x_0)^{\eta} \mathrm{d}W_t$$

$$P_0(x) \sim \exp_{1+1/\nu}(-\nu x/x_0)$$

Reflective boundary condition at x = 0

B. Kaulakys and M. Alaburda, J. Stat. Mech. 2009, P02051 (2009).

q-Gaussian steady-state PDF

$$\mathrm{d}x = \sigma^2 (\eta - \nu/2) (x^2 + x_0^2)^{\eta - 1} x \mathrm{d}t + \sigma (x^2 + x_0^2)^{\eta/2} \mathrm{d}W_t$$

 $P_0(x) \sim \exp_{1 + 2/\nu} (-\nu x^2/2x_0^2)$

B. Kaulakys, M. Alaburda, and V. Gontis, AIP Conf. Proc. 1129, 13 (2009).
V. Gontis, B. Kaulakys, and J. Ruseckas, AIP Conf. Proc. 1129, 563 (2009).
V. Gontis, J. Ruseckas, and A. Kononovičius, Physica A, 389, 100 (2010).

q-exponential function: $\exp_q(x) \equiv (1 + (1 - q)x)^{1/(1-q)}$

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Trading return

The distribution of normalized return r per 1 min is close to q-Gaussian.



Normalized trading return per 1 min for ABT stock on NYSE

J. Ruseckas et al. (Lithuania)

Nonlinear stochastic differential equations

For some choces of parameters our SDE takes the form of well-known SDE's considered in econopysics and finance.

• $\eta = 0$ and $\sigma = 1$ corresponds to the Bessel process

$$\mathrm{d}x = \frac{\delta - 1}{2} \frac{1}{x} \mathrm{d}t + \mathrm{d}W_t$$

of dimension $\delta = \mathbf{1} - \nu$

• $\eta = 1/2$, $\sigma = 2$ corresponds to the squared Bessel process

$$\mathrm{d}x = \delta \mathrm{d}t + 2\sqrt{x}\,\mathrm{d}W_t$$

of dimension $\delta = 2(1 - \nu)$

Connection with other equations

 SDE with exponential restriction with η = 1/2, x_{min} = 0 and m = 1 gives Cox-Ingersoll-Ross (CIR) process

$$\mathrm{d}\mathbf{x} = \mathbf{k}(\theta - \mathbf{x})\mathrm{d}t + \sigma\sqrt{\mathbf{x}}\,\mathrm{d}\mathbf{W}_t$$

where
$$k = \sigma^2/2x_{\text{max}}$$
, $\theta = x_{\text{max}}(1 - \nu)$

• When $\nu = 2\eta$, $x_{max} = \infty$ and $m = 2\eta - 2$ then we get the Constant Elasticity of Variance (CEV) process

$$\mathrm{d}\boldsymbol{x} = \mu \boldsymbol{x} \mathrm{d}\boldsymbol{t} + \sigma \boldsymbol{x}^{\eta} \mathrm{d}\boldsymbol{W}_{t}$$

where
$$\mu = \sigma^{2}(\eta - 1)x_{\min}^{2(\eta - 1)}$$

Numerical simulation







Power spectral density



Distribution of x

Used parameters: $\nu = 3$, $\eta = 5/2$, $x_{\min} = 1.0$, $x_{\max} = 10^3$. 1/f spectrum.

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Analytically solvable case

A CEV process:

$$\mathrm{d}\boldsymbol{x} = \mu \boldsymbol{x} \mathrm{d}\boldsymbol{t} + \sigma \boldsymbol{x}^{\frac{3}{2}} \mathrm{d}\boldsymbol{W}_t$$

where $\mu = \sigma^2 x_{\min}/2$, $\eta = 3/2$, $\nu = 3$ and $x_{\max} = \infty$ Transition probability is

$$P_{x}(x',t|x,0) = \frac{x_{\min}}{(1-e^{-\mu t})} \sqrt{\frac{x}{x'^{5}}} \exp\left(\frac{1}{2}\mu t - \frac{x_{\min}}{(1-e^{-\mu t})} \left(\frac{1}{x'} + \frac{1}{x}e^{-\mu t}\right)\right) \\ \times I_{1}\left(\frac{x_{\min}}{\sinh\left(\frac{1}{2}\mu t\right)} \frac{1}{\sqrt{xx'}}\right)$$

The steady-state probability distribution has the form

$$P_0(x) = x_{\min}^2 x^{-3} \exp(-x_{\min}/x)$$

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The autocorrelation function

$$C(t) = -x_{\min}^2 e^{\mu t} \ln\left(1 - e^{-\mu t}\right)$$

When $\mu t \ll 1$ we get $C(t) \approx -x_{\min}^2 \ln(\mu t)$ The power spectral density

$$S(f) = -4x_{\min}^2 \operatorname{Re}\left[rac{\gamma + \psi(\mathrm{i}\omega/\mu)}{\mu - \mathrm{i}\omega}
ight]$$

where γ is the Euler's constant and $\psi(\cdot)$ is the digamma function. When $\omega \gg \mu$ then the power spectral density is $S(f) \approx x_{\min}^2/f$

Analytically solvable case



Probability distribution function $P_0(x)$ and power spectral density S(f)

1/f noise and eigenvalues of the F-P equation

- Solutions of the Fokker-Planck equation having the form $P(x, t) = P_{\lambda}(x)e^{-\lambda t}$ determine eigenfunctions $P_{\lambda}(x)$ and eigenvalues λ
- The power spectral density

$$\mathcal{S}(f) = 4 \sum_{\lambda} \frac{\lambda}{\lambda^2 + \omega^2} X_{\lambda}^2, \qquad X_{\lambda} = \int_{x_{\min}}^{x_{\max}} x \mathcal{P}_{\lambda}(x) \, \mathrm{d}x$$

- The shape of the power spectral density depends on the behavior of the eigenfunctions and the eigenvalues
- Expression for the power spectral density resembles the models of 1/f noise using the sum of the Lorentzian spectra. The Lorentzians can arise from the single nonlinear stochastic differential equation

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1/f noise and eigenvalues of the F-P equation

$$\mathcal{S}(f) pprox 4 \int rac{\lambda}{\lambda^2 + \omega^2} X_{\lambda}^2 \mathcal{D}(\lambda) \, \mathrm{d}\lambda \sim \int_{\lambda_{\min}}^{\lambda_{\max}} rac{1}{\lambda^{eta - 1}} rac{1}{\lambda^2 + \omega^2} \, \mathrm{d}\lambda$$

The largest contribution make the terms corresponding to the eigenvalues λ obeying the condition $\lambda_{\min} \ll \lambda \ll \lambda_{\max}$, where

$$\begin{split} \lambda_{\min} &= \sigma^2 \boldsymbol{x}_{\min}^{2(\eta-1)}, \qquad \lambda_{\max} = \sigma^2 \boldsymbol{x}_{\max}^{2(\eta-1)}, \qquad \eta > 1\\ \lambda_{\min} &= \sigma^2 \boldsymbol{x}_{\max}^{2(\eta-1)}, \qquad \lambda_{\max} = \sigma^2 \boldsymbol{x}_{\min}^{2(\eta-1)}, \qquad \eta < 1 \end{split}$$

When $\lambda_{\min} \ll \omega \ll \lambda_{\max}$ then the leading term in the expansion in the power series of ω is

$$S(f) \sim \omega^{-\beta}, \qquad \beta < 2$$

- The shape of the power spectral density depends on the behavior of the eigenfunctions and the eigenvalues in terms of the function X²_λD(λ).
- One obtains 1/f^β behavior of the power spectral density when function X²_λD(λ) is proportional to λ^{-β} for a wide range of eigenvalues λ

J. Ruseckas and B. Kaulakys, Phys. Rev. E 81, 031105 (2010).

- We obtain a class of nonlinear SDEs, giving the power-law behavior of the power spectral density in any desirably wide range of frequencies
- and power-law steady state distribution of the signal intensity.
- The equations, as special cases, contain the well-known SDEs in economics and finance.
- One of the reasons for the appearance of the 1/f spectrum is the scaling property of the SDE.
- The power spectral density may be represented as a sum of the Lorentzian spectra.

Thank you for your attention!