

Abstract: Information geometry or the application of differential geometry to the information sciences has become a promising tool for the analysis of complex dynamical systems. From the information geometry tools, we have the concept of information length defined as the integral of the information rate and describing the total amount of statistical changes that a time-varying probability distribution takes through time.

Throughout this poster, we introduce the definition of information length and its application to stochastic thermodynamics, abrupt events detection, time series analysis and control engineering. All this is in the light of linear non-autonomous stochastic systems.

Keywords: Information geometry, Information length, Stochastic dynamics, Abrupt events, Time series, Control theory

Information Length (IL)

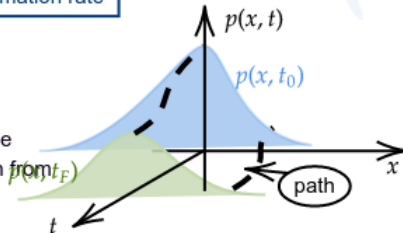
Definition

$$\mathcal{L}(t) = \int_{t_0}^{t_f} \Gamma(\tau) d\tau = \int_{t_0}^{t_f} \int_{\mathbb{R}^n} dx \sqrt{\frac{[\partial_\tau p(x; \tau)]^2}{p(x; \tau)}} d\tau$$

information rate

Interpretation

Information length is a metric of the total of statistical states that the system passes through from $p(x, t_0)$ to $p(x, t_f)$.



We focus on its application to:

Linear non-autonomous stochastic systems

Langevin equation

Assuming

$$\begin{cases} \zeta = [\zeta_1, \zeta_2, \dots, \zeta_n]^T \\ \mathbf{A} \in \mathbb{R}^{n \times n}, \mathbf{B} \in \mathbb{R}^{n \times 1}, u \in \mathbb{R} \\ \xi = [\xi_1, \xi_2, \dots, \xi_n]^T \\ \langle \xi_i(t) \rangle = 0 \\ \langle \xi_i(t) \xi_j(t') \rangle = 2D_{ij} \delta(t - t') \\ t, t' \in \mathbb{R}, i, j = 1, 2, \dots \end{cases}$$

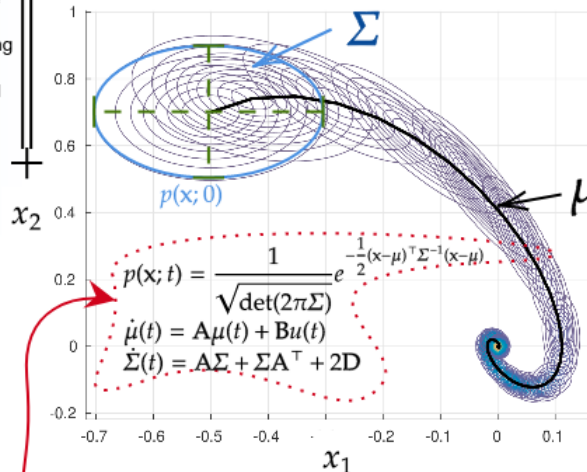
Corresponding Fokker-Planck equation

Solution

$$\frac{\partial p(x; t)}{\partial t} = - \sum_{i,j=1}^n \frac{\partial}{\partial x_i} [(A_{ij}x_j + B_i u(t))p(x; t)] + \sum_{i,j=1}^n D_{ij} \frac{\partial^2}{\partial x_i \partial x_j} p(x; t)$$

Example:

2nd order stochastic dynamical system [1]



IL in Linear non-autonomous stochastic systems [1]

Compute using $\Gamma(t) = \dot{\mu}^T \Sigma^{-1} \dot{\mu} + \frac{1}{2} \text{Tr}(\Sigma^{-1} \dot{\Sigma})^2$

Applications

IL in Stochastic Thermodynamics [5]

IL can be related to the entropy rate \dot{S}

$$\dot{S} = \int_{\mathbb{R}^n} \dot{p}(x; t) \ln(p(x; t)) dx = \Pi - \Phi$$

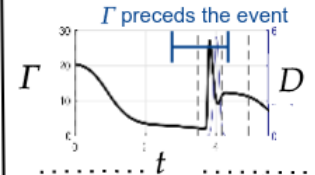
using $\Gamma^2 \leq \varepsilon_u := \text{Tr}(\Sigma^{-1}) \Pi \text{Tr}(\mathbf{D}) + \dot{S}^2 - 2g(\mathbf{H})$

where $g(\mathbf{H}) := \sum_{i < j}^n \lambda_i \lambda_j$

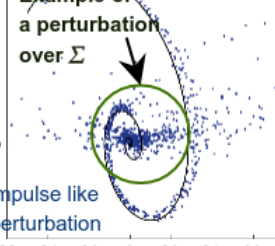
s.t. λ_i are the eigenvalues of the matrix $\mathbf{H} = \Sigma^{-1} \mathbf{D} + \mathbf{A}$

IL in abrupt events prediction [2]

The value of information rate Γ can be used to detect perturbations over μ and Σ .

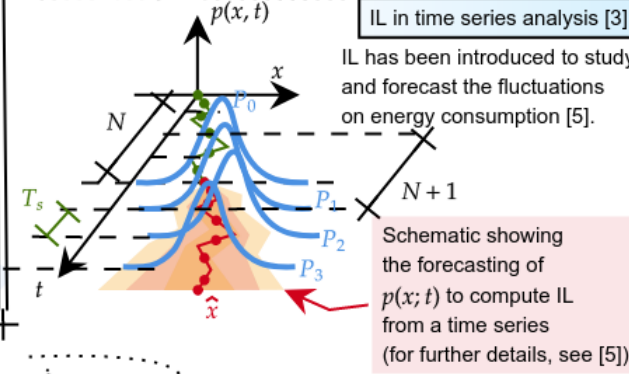


Example of a perturbation over Sigma

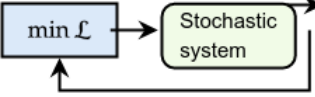


IL in time series analysis [3]

IL has been introduced to study and forecast the fluctuations on energy consumption [5].



IL in Control Engineering [6]



Design and implementation of minimum variability control through the minimisation of IL in real time [6].

References

- [1] Guel-Cortez, Adrian-Josue, and Eun-jin Kim. "Information length analysis of linear autonomous stochastic processes." Entropy 22.11 (2020): 1265.
- [2] Guel-Cortez, Adrian-Josue, and Eun-jin Kim. "Information geometric theory in the prediction of abrupt changes in system dynamics." Entropy 23.6 (2021): 694.
- [3] Chamorro, Harold, et al. "Information Length Quantification and Forecasting of Power Systems Kinetic Energy." IEEE Transactions on Power Systems (2022).
- [4] Kim, Eun-jin, and Adrian-Josue Guel-Cortez. "Causal Information Rate." Entropy 23.8 (2021): 1087.
- [5] Guel-Cortez, Adrian-Josue, and Eun-jin Kim. "Relations between entropy rate, entropy production and information geometry in linear stochastic systems." To be submitted. (2022).
- [6] Guel-Cortez, Adrian-Josue, Kim, Eun-jin, and Mohamed, W. Mehrez. "Minimum information variability in linear Langevin systems via model predictive control." To be submitted. (2022)