



Applied Stochastic Processes

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Plan for Today

1. **What is a stochastic process?**

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2. **Goals/content of the lectures**

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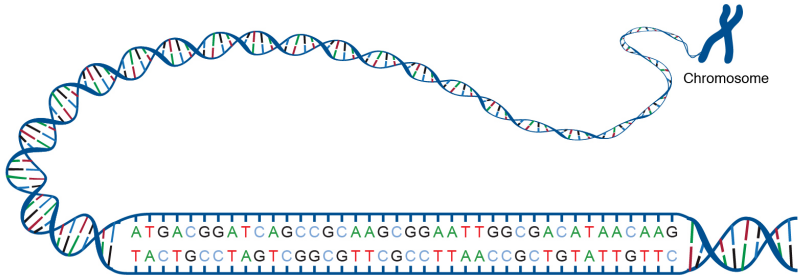
- 1. What is a stochastic process?**
- 2. Goals/content of the lectures**
- 3. Administrative and practical information**

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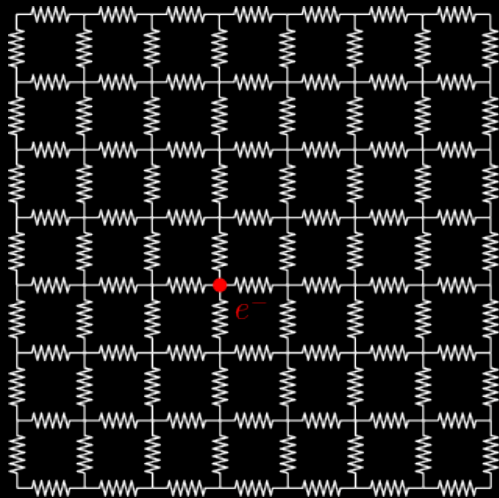
1. **What is a stochastic process?**
2. **Goals/content of the lectures**
3. **Administrative and practical information**
4. **Chapter 1: Markov Chains**

1. What is a stochastic process?

A stochastic process in Genetics

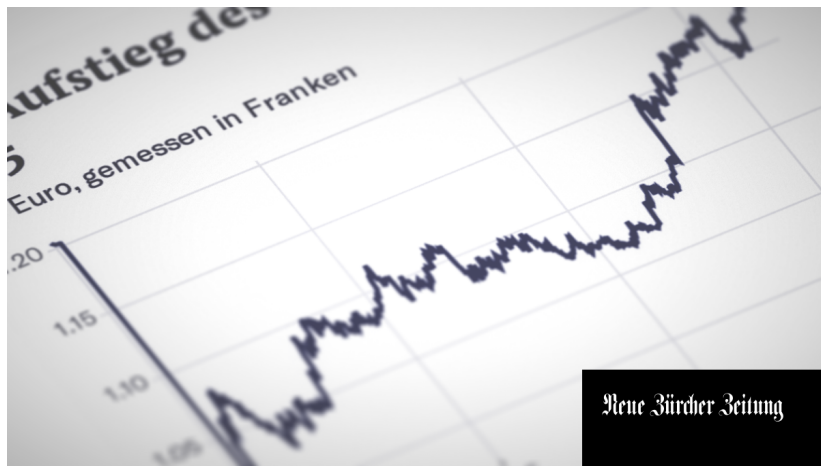


In Electronics



On the road





In Geneva



Mysterious



Mathematical definition

Setup:

- $(\Omega, \mathcal{F}, \mathbb{P})$ probability space.
- (S, \mathcal{S}) measured space. "state space"

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A random variable in S is a measurable map

$$X : \Omega \rightarrow S.$$

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Examples:

- $S = \{-1, 1\}$,

$$\mathbb{P}[X = -1] = \mathbb{P}[X = 1] = \frac{1}{2}. \quad \text{"Coin Flip"}$$

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- $S = \mathbb{R}$, $X \sim \mathcal{N}(0, 1)$.

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A discrete-time stochastic process in S is a sequence $X = (X_n)_{n \in \mathbb{N}}$ of random variables in S .

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Examples:

- $(X_n)_{n \in \mathbb{N}}$ iid coinflips.
- $(S_n)_{n \in \mathbb{N}}$ where $S_n = X_1 + \dots + X_n$. "random walk"
- $(M_n)_{n \in \mathbb{N}}$ martingale.

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Definition

A continuous-time stochastic process in S is a collection $X = (X_t)_{t \in \mathbb{R}_+}$ of random variables in S .

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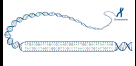
Definition

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Continuous Stochastic Process = "random function".

Applications



$X_n \in \{A, C, T, G\}$
 $(X_n)_{n \in \mathbb{N}} \in E$
 "MARKOV CHAIN"



$X_n \in \mathbb{Z}^2$
 $(X_n)_{n \in \mathbb{N}}$
 "RANDOM WALK"



POISSON PROCESS
 $(N_t)_{t \in \mathbb{R}_+}$
 $N_t =$ "number of cars passing a fixed place during $[0, t)$ "



$B_t \in \mathbb{R}$
 $(B_t)_{t \in \mathbb{R}_+}$
 Brownian Motion



$\Gamma_t \in \mathbb{R}^2$
 $(\Gamma_t)_{t \in \mathbb{R}_+}$
 SLE process



$(K_t) \in \mathcal{K}_0(\mathbb{R}^2)$
 $(K_t)_{t \in \mathbb{R}_+}$
 DLA

2. Goals/content of the lectures

Outlook

1. Markov Chains

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2. Renewal Processes

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- Prepare to Brownian Motion.