

Stochastic Processes

MATH5835, P. Del Moral

UNSW, School of Mathematics & Statistics

Lectures Notes 1

Consultations (RC 5112):

Wednesday 3.30 pm \rightsquigarrow 4.30 pm & Thursday 3.30 pm \rightsquigarrow 4.30 pm

Stochastic processes

- ▶ **Random** dynamical system.
- ▶ Sequential **simulation** of random variables.

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The generation of random numbers is too important to be left to chance.

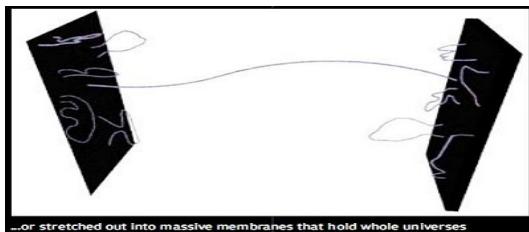
Robert Coveyou [Studies in Applied Mathematics, III (1970), 70-111.]

Lost in the Great Sloan Wall

The Tau Zero Foundation \rightsquigarrow interstellar travels in any dimension.

Tony Gonzales random travelling plans in the universe lattices at superluminal speeds

- ▶ dimension 1 and 2 from Reykjavik:

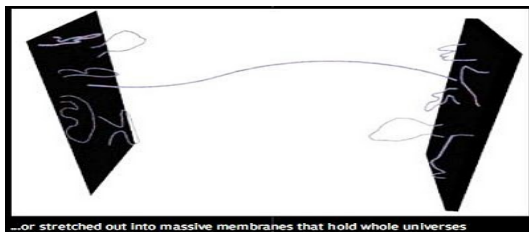


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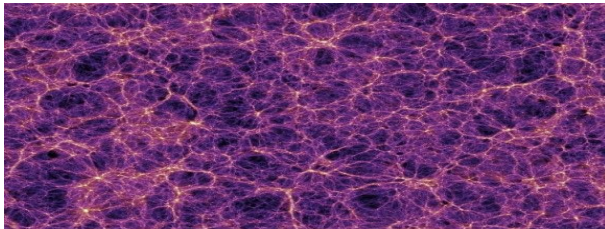
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Main drawbacks: infinite returns back home....

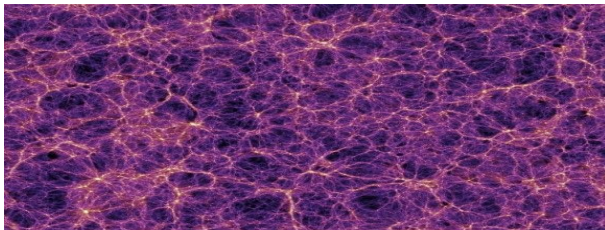
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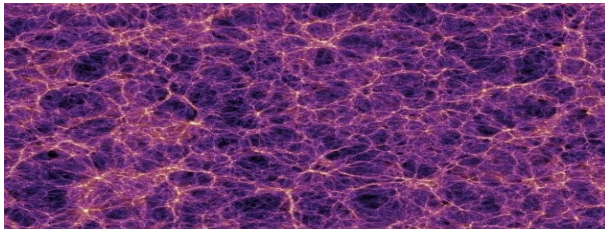
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Main advantage: finite mean returns back home.

Lost in the Great Sloan Wall

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Main advantage: finite mean returns back home.

Main drawbacks:

Wanders off in the infinite universe and never returns back home!

Lost in the Great Sloan Wall - Why??

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Stoch. model = Simple random walks on \mathbb{Z}^d :

$$d = 1 \quad \Rightarrow \quad X_n = X_{n-1} + U_n \quad \text{with} \quad U_n = +1 \text{ or } -1 \quad \text{proba } 1/2$$

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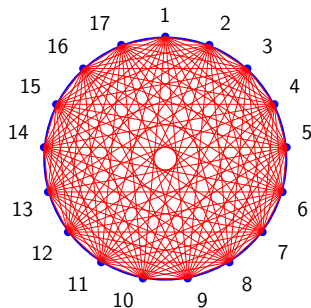
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⊕ More rigorous analysis for $d = 2, 3$ = Project N° 1 

Meeting Alice in Wonderland

Alice + White rabbit \in Polygonal labyrinth (no communicating edges!):



- ▶ Chances to meet after 5 moves : $> 25\%$
- ▶ Chances to meet after 12 moves : $> 51\%$
- ▶ Chances to meet after 40 moves : $> 99\%$

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Neighborhood systems:

On the set of vertices $x \sim y \implies (x, y) \in \mathcal{E} = \text{set of edges}$

\Downarrow

Set of neighbors of $x \in \mathcal{V} := \mathcal{N}(x) = \{y \in \mathcal{V} : y \sim x\}$

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Stochastic model:

X_n uniform r.v. on the set of neighbors $\mathcal{N}(X_{n-1})$

\Updownarrow

$$\mathbb{P}(X_n = y \mid X_{n-1} = x) = \frac{1}{\#\mathcal{N}(X_{n-1})} \mathbf{1}_{\mathcal{N}(X_{n-1})}(y)$$

Meeting Alice in Wonderland - Simulation \oplus Analysis

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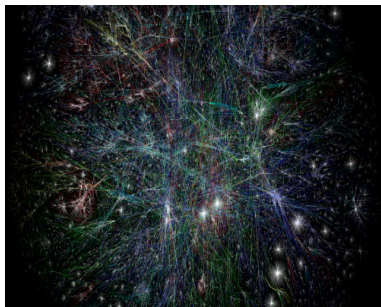
Other application domains?

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Other application domains?

Ex.: web-graph analysis (ranking \Rightarrow recommendations)



The MIT Blackjack team

Mr M. card shuffle tracking and P. Diaconis magic number

≤ 5 shuffles \implies possible predictions $\geq 90\%$,

6 shuffles \implies possible predictions $\geq 40\%$!

Some questions?

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$$52! \simeq 8.053 \cdot 10^{67}$$

$$78! \simeq 1.13 \cdot 10^{115} \gg \text{number } 10^{80} \text{ of particles in the universe}$$

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Change of order σ_{n+1} or change of values

$i \rightsquigarrow Y_n(i) = \text{position} \rightsquigarrow \sigma_{n+1}(Y_n(i)) = Y_{n+1}(i) = \text{new position}$

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\Downarrow

$$X_n = X_{n-1} \circ \sigma_n^{-1} \quad \text{or} \quad Y_n = \sigma_n \circ Y_{n-1}$$

with some i.i.d. r.v. σ_n in some class : transpositions, top-in, riffles,...

Shuffling cards – Simulation+Analysis

- ▶ Simulation ? \rightsquigarrow

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- ▶ Math analysis ? \rightsquigarrow convergence to stationarity.

X_n and $Y_n \xrightarrow{n \rightarrow \infty}$ limiting r.v. X_∞ and Y_∞

with for any uniform r.v. σ on \mathcal{G}_d

$$X_\infty \stackrel{\text{in law}}{=} X_\infty \circ \sigma^{-1} \quad \text{and} \quad Y_\infty \stackrel{\text{in law}}{=} \sigma \circ Y_\infty$$

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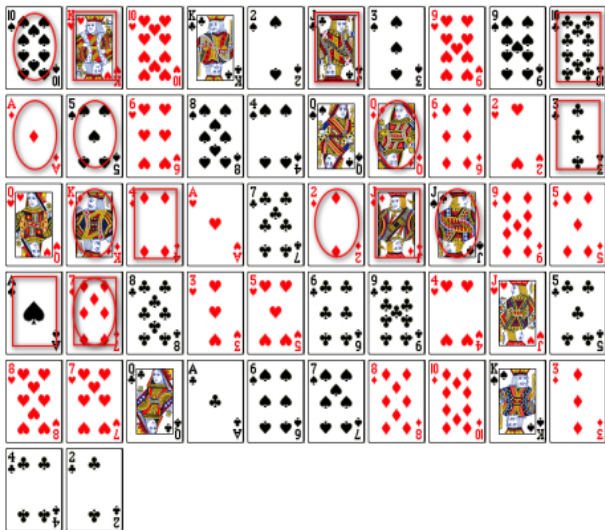
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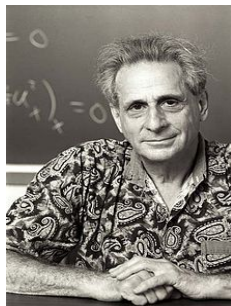
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⊕ Top-in & Transpositions = Project $N^o 2$ 

Kruskal count (with the classroom!)



Kruskal count - Stochastic model \oplus Simulation \oplus Analysis

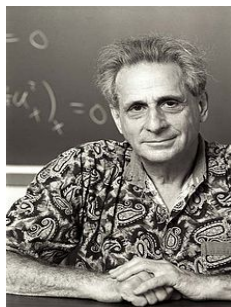


Martin David Kruskal (1925-2006)

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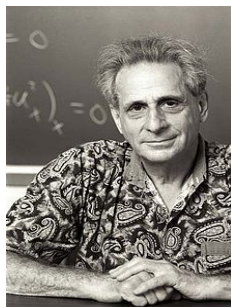
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 \rightsquigarrow **Deterministic** walker on a **random environment**.
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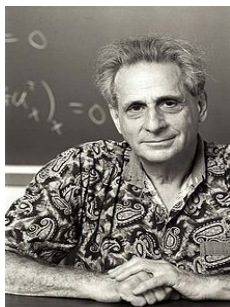
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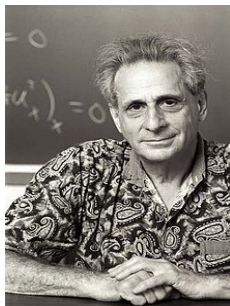
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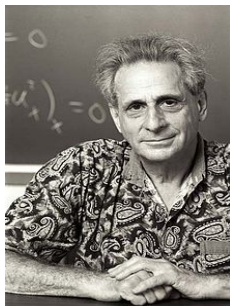
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The magic fern from Daisetsuzan



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The magic fern - The stochastic model

Stoch. model = Iterated random functions (IRF):

- ▶ Some functions $x \in \mathbb{R}^2 \mapsto f_i(x) = A_i x + b_i \in \mathbb{R}^2$, with $i \in I = \{1, \dots, d\}$.
- ▶ Some i.i.d. r.v. ϵ_n with some law μ on $\{1, \dots, d\}$

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$$X_n = f_{\epsilon_n}(X_{n-1}) = (f_{\epsilon_n} \circ f_{\epsilon_{n-1}} \circ \dots \circ f_{\epsilon_1})(x_0)$$

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$$X_n = f_{\epsilon_n}(X_{n-1}) = (f_{\epsilon_n} \circ f_{\epsilon_{n-1}} \circ \dots \circ f_{\epsilon_1})(x_0)$$

Do you believe this?

⊕ Fractal & IRF = Project N° 3 🌶

The magic fern from Daisetsuzan - An illustration

Scilab program: *fractal.tree.sce*

$$A_1 = \begin{pmatrix} 0 & 0 \\ 0 & c \end{pmatrix} \quad b_1 = \begin{pmatrix} 1/2 \\ 0 \end{pmatrix}$$

$$A_2 = \begin{pmatrix} r \cos(\varphi) & -r \sin(\varphi) \\ r \sin(\varphi) & r \cos(\varphi) \end{pmatrix} \quad b_2 = \begin{pmatrix} \frac{1}{2} - \frac{r}{2} \cos(\varphi) \\ c - \frac{r}{2} \sin(\varphi) \end{pmatrix}$$

et

$$A_3 = \begin{pmatrix} q \cos(\psi) & -r \sin(\psi) \\ q \sin(\psi) & r \cos(\psi) \end{pmatrix} \quad b_3 = \begin{pmatrix} \frac{1}{2} - \frac{q}{2} \cos(\psi) \\ \frac{3c}{5} - \frac{q}{2} \sin(\psi) \end{pmatrix}$$

with ϵ_n i.i.d. uniform on $\{1, 2, 3\}$, and with the parameters

$$\begin{aligned} c &= 0.255, & r &= 0.75, & q &= 0.625 \\ \varphi &= -\frac{\pi}{8}, & \psi &= \frac{\pi}{5}, & |X_0| &\leq 1. \end{aligned}$$

The Kepler-22b Eve

Migration in 2457 of selected 1000 humans:



Reproduction rate (20 years)

- ▶ 5.5 thousands years \Rightarrow 25% population \in same family.
- ▶ 6.5 thousands years \Rightarrow 68% population \in same family.
- ▶ 10 thousands years \Rightarrow more than 99% population \in same family.

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The Seven Daughters of Eve of Bryan Sykes:

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The Seven Daughters of Eve of Bryan Sykes: **Mitochondrial Eve**
 \simeq 140 – 200 thousands years (missprint in manuscript)

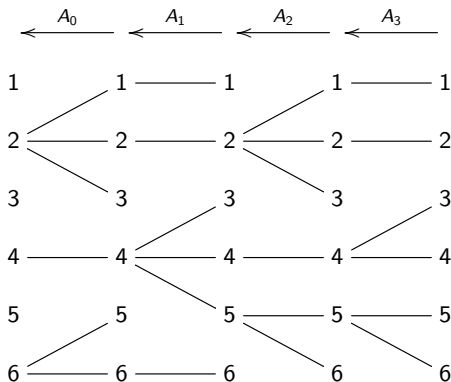
The Kepler-22b Eve - Stochastic model

Random walks on functions $a : i \in \{1, \dots, d\} \mapsto \{1, \dots, d\}$

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Random walks on functions $a : i \in \{1, \dots, d\} \mapsto \{1, \dots, d\}$

Birth and death = selection of the parents/ancestors/types



The Kepler-22b Eve - Simulation \oplus Analysis

- ▶ **Simulation?**

The Kepler-22b Eve - Simulation \oplus Analysis

► Simulation?

Neutral genetic model \rightsquigarrow Each $A(i)$ uniform on $\{1, \dots, d\}$

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The label of the "surviving" ancestors of the population at time n :

$$X_n = X_{n-1} \circ A_n = A_0 \circ \dots \circ A_n : i \in \{1, \dots, d\} \mapsto \{1, \dots, d\}$$

Intuitively:

$$X_n \xrightarrow{n \rightarrow \infty} X_\infty = \text{Constant random function}$$

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\oplus More rigorous analysis = Project N^o 4 

Poisson typos (reminder of stats course)

- ▶ Arrival times T_n of events :
misprints in a text, trades counts, bus arrivals, machine failures, catastrophies, number of tries in rugby games, ...
- ▶ Stat. model on the time axis = i.i.d. exponential inter-arrivals

$$\mathbb{P}((T_{n+1} - T_n) \in dt \mid T_n) = e^{-t} \mathbf{1}_{[0, \infty[}(t) dt$$

► **Consequences (admitted, cf. pb.1 p. 71)**

$$\forall n \geq 0 \quad \mathbb{P}(\#\{T_i \in [0, \lambda]\} = n) = e^{-\lambda} \frac{\lambda^n}{n!} = \text{Poisson r.v.}$$

and

$$\left(\frac{T_1}{T_{n+1}}, \frac{T_2}{T_{n+1}}, \dots, \frac{T_n}{T_{n+1}} \right) = \text{ordered unif. stats on } [0, 1] \perp T_{n+1}$$

► **Extension to any state space (\supset the book of D. Poisson)?**

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► **Extension to any state space (\supset the book of D. Poisson)?**

\rightsquigarrow Counting processes with intensity $x \in S \mapsto f(x) \in [0, \infty[$

$$\mathcal{X} = \sum_{1 \leq i \leq N} \delta_{X_i}$$

with

- N Poisson r.v. with $\lambda = \int f(x) dx$.
- Given $N = n$, X^1, \dots, X^n i.i.d. with density $g(x) = f(x)/\lambda$.