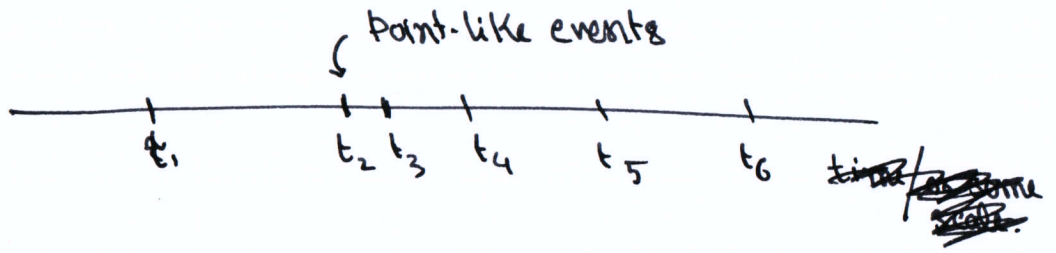


Stochastic processes: few definitions

① Point process:
(a sequence of events)



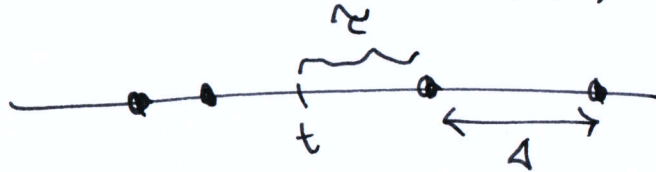
Ex: photon counters, earthquakes, neural spikes etc.

• A hot topic: determinantal point process: Poob as determinant of something.
Ex: Eigenvalues of Random matrix

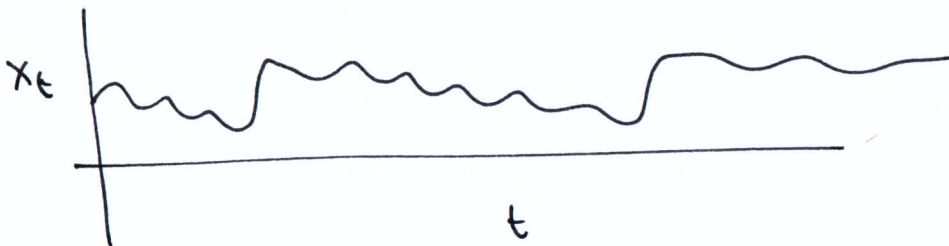
• Waiting time and renewal theory.

① what is the distribution of ~~time~~ before an event happens. (τ)

② what is the distribution of time-gap (Δ)



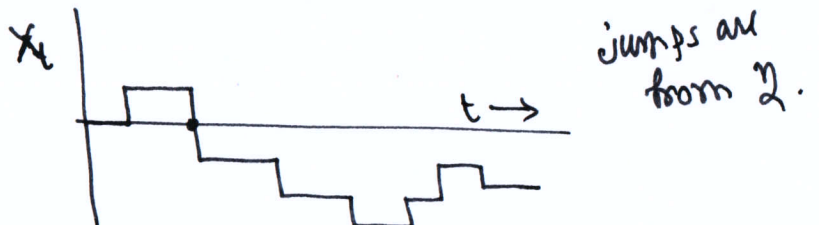
② Stochastic process:



Formal definition: let Ω is a random variable with prob $P(\Omega)$.

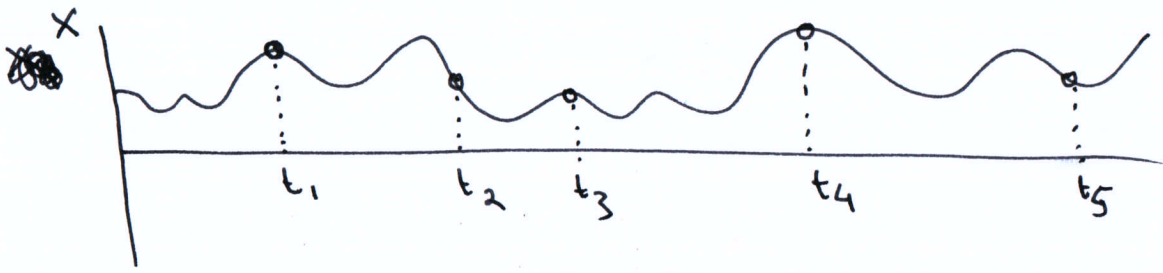
then, a mapping $\Omega \rightarrow X_t$ is called realization of a stochastic process.

Example: Random walk



Think like a dynamical system.

Markov-stochastic process : Book by von Kampen



If conditional probability

$$P(x_{n+1}, t_{n+1} | x_n, t_n; \dots; x_1, t_1) = P(x_{n+1}, t_{n+1} | x_n, t_n)$$

Then Markovian. \uparrow (Also called transition probability)

Remark: The dynamics is completely characterized by the Conditional probability $P(x_{n+1}, t_{n+1} | x_n, t_n)$. and initial ~~condi~~ probability.

Remark: Note, how time scale is important.

~~$P(x_{n+1}, t_{n+1}; x_n, t_n; \dots; x_1, t_1) = P(x_{n+1}, t_{n+1} | x_n, t_n) P(x_n, t_n; \dots; x_1, t_1)$~~

$$P(x_{n+1}, t_{n+1}; x_n, t_n) = M(x_{n+1}, t_{n+1} | x_n, t_n) P(x_n, t_n)$$

Repeating

$$P(x_{n+1}, t_{n+1}; x_n, t_n; \dots; x_1, t_1) = M(x_{n+1}, t_{n+1} | x_n, t_n) M(x_n, t_n | x_{n-1}, t_{n-1}) \dots M(x_2, t_2 | x_1, t_1) P(x_1, t_1)$$

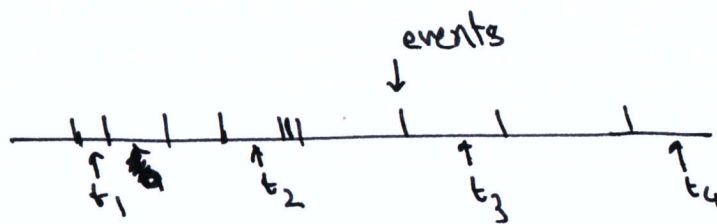
⊗ Show:

$$M(x_{n+1}, t_{n+1} | x_{n-1}, t_{n-1}) = \int dx_n M(x_{n+1}, t_{n+1} | x_n, t_n) M(x_n, t_n | x_{n-1}, t_{n-1})$$

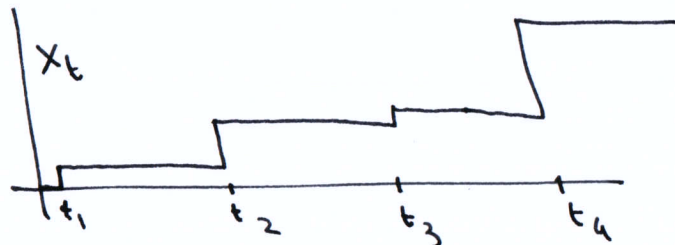
Chapman-Kolmogorov equation.

Examples of Markov process:

(a) Poisson process:



let $X_t =$ total number of events upto time t .



$$M(X_{t_{n+1}}, t_{n+1} | X_n, t_n) = \frac{[\lambda (t_{n+1} - t_n)]^{X_{n+1} - X_n}}{(X_{n+1} - X_n)!} e^{-\lambda (t_{n+1} - t_n)}$$

$\lambda :=$ rate of event.

Example: radioactive decay.

(b) Wiener-Process / Brownian motion

$$M(X_{t_{n+1}}, t_{n+1} | X_n, t_n) = \frac{1}{\sqrt{2\pi D(t_{n+1} - t_n)}} e^{-\frac{(X_{n+1} - X_n)^2}{2D(t_{n+1} - t_n)}}$$

Example of non-Markov process: fractional Brownian motion.

$$P[X(t)] \sim e^{-\int dt_1 dt_2 X(t_1) c^{-1}(t_1, t_2) X(t_2)}$$

with $c(t_1, t_2) = t_1^{2H} + t_2^{2H} - |t_1 - t_2|^{2H}$

Stationary Markov process:

$$M(x', t' | x, t) = M(x', t' - t | x, 0)$$

* There is ~~little~~ subtlety in exact definition. For details see van Kampen.

~~When~~

~~Markov process is~~

~~called a homogeneous Markov process.~~

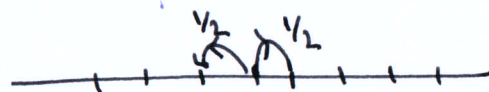
~~do~~

Markov chain:

~~homogeneous~~ A stationary Markov process, on discrete time $\{0, 1, 2, \dots\}$ and discrete space \mathcal{X} .

Not a strict definition

Example random walk



Few basic properties of Markov chains

~~When~~

* Time-homogeneous Markov chain:



$$M(e', t | e, 0) = M(e', t | e)$$

does not depend on time.

$$P_{t+dt}(e') = \sum_{e''} M(e', dt | e'') P_t(e'')$$

Master-equation.

~~Properties of~~ ~~in vector notation~~

~~1)~~

~~P_t = M^t~~

in operator notation: $|P_{t+dt}\rangle = M |P_t\rangle$

$$|P_t\rangle = \begin{pmatrix} P_t(e_1) \\ \vdots \\ P_t(e_n) \end{pmatrix} \quad M = \begin{pmatrix} & \\ & M(e'_j | e) \end{pmatrix}$$

Basic properties of M :

① $M(c'|c) \geq 0$ ~~is not zero~~. (because they are probability)

② Conservation of probability gives

$$1 \Leftarrow \sum_{c'} P_{t+\Delta t}(c') = \sum_c \left(\underbrace{\sum_{c'} M(c',c)}_1 \right) P_t(c)$$

$$= \sum_c P_t(c) = 1$$

$\Rightarrow \sum_{c'} M(c',c) = 1$ unit column sums ~~is~~
(Markov-matrix)

~~This is usually incorporated by writing~~

~~$M(c',c) \geq 1 - \sum_{c'' \neq c} M(c',c'')$ in master equation, which~~

~~is~~

~~$\sum_{c'} M(c',c) \geq 1$~~

③ M is NOT symmetric! (non-Hermitian)

Means, left and right eigen vectors are not same.

For each eigenvalue λ :

$$M|r_\lambda\rangle = \lambda|r_\lambda\rangle \quad \text{and} \quad \langle l_\lambda|M = \lambda\langle l_\lambda|$$

peculiarities:

⊗ λ could be complex numbers.

⊗ May not be fully diagonalizable (Jordan block decomposition)

⊗ Eigenstates may not form a ~~complete~~ ^{complete} set

Example: TS & Deepak Dhar, J Stat Phys.

⊗ Interesting reading: Hermitian is not self-adjoint.

For QM example, see a phys today article.

Stationary state at large time.

$$|P_{t+4t}\rangle = |P_t\rangle = |P_{st}\rangle$$

Expressing in terms of Eigenbasis:

$$\sum_{\lambda} |q_{\lambda}\rangle \langle q_{\lambda}| = \mathbb{1}_{N \times N} \quad \left. \begin{array}{l} \text{Completeness} \\ \langle q_{\lambda} | q_{\lambda'} \rangle = \delta_{\lambda, \lambda'} \end{array} \right\}$$

Orthonormality.

We write

$$|P_t\rangle = \sum_{\lambda} \langle q_{\lambda} | P_t \rangle \cdot |q_{\lambda}\rangle$$

Then, Time evolution gives,

$$\begin{aligned} |P_t\rangle &= \mathcal{M} |P_{t-4t}\rangle = \mathcal{M}^2 |P_{t-24t}\rangle \\ &= \mathcal{M}^n |P_0\rangle \\ &= \mathcal{M}^n \sum_{\lambda} \langle q_{\lambda} | P_0 \rangle |q_{\lambda}\rangle \\ &= \sum_{\lambda} \langle q_{\lambda} | P_0 \rangle \lambda^n |q_{\lambda}\rangle \end{aligned}$$

Stationarity: If there is a time independent state at $t \rightarrow \infty$, then ^① that state corresponds to $\lambda = 1$.

② And just $|\lambda| < 1$.

③ and $\langle q_1 | = \langle 1 |$ because $\sum_{c'} \mathcal{M}(c', c) = 1$
 $\Rightarrow \langle 1 | \mathcal{M} = \langle 1 |$

$$\text{④ } |P_t\rangle \xrightarrow{t \rightarrow \infty} |P_{st}\rangle = |q_1\rangle$$

The right eigenvector associated to largest eigenvalue gives stationary state.

Q. What ensures stationarity?

Perron-Frobenius theorem: Let M be an irreducible, non-negative square matrix, (with period 1), then

- ① Largest eigenvalue λ_0 is real and positive. $\lambda_0 \geq 0$.
- ② Rest eigenvalues $|\lambda| < \lambda_0$.
- ③ Largest eigenstate is non-degenerate, meaning there is exactly one right eigenvector and one left eigenvector.
 (r_0) (l_0)
- ④ All elements of r_0 and l_0 are ~~non~~ positive (> 0).
And they are the only such eigenvectors.

* Irreducible essentially means: if M has some zero entries, then M^n has all positive elements for finite n .

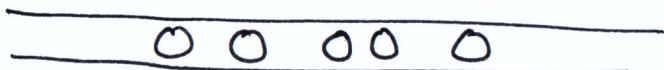
Example: $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ is reducible, \Rightarrow No Perron-Fro. Th.

* Irreducible M means, ergodic. \Rightarrow it is possible to go from one state to another by finite steps.

Example of a reducible (non-ergodic) matrix M :

$$M = \begin{pmatrix} [A] & 0 \\ 0 & [B] \end{pmatrix} \text{ block diagonal.}$$

Example of such a dynamics \rightarrow energy conserving moves.



Elastic collision. Only two energies E_1, E_2

* Recall the standard definition of irreducible matrix.

that M can not be transformed into a upper block triangular form by PMP^{-1} , where P is a permutation matrix.

How is this related to the definition we used?

Ans: Hint: permutation means relabelling of states.

Periodicity: $M = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \Rightarrow \lambda = 1$ degenerate.

check: $M^2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$
 $M^3 = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = M$ } \Rightarrow ergodic/irreducible.

what it means? the ~~conf~~ system jumps from one config to another and back periodically.

$$(10) \rightarrow (01) \rightarrow (10) \rightarrow (01)$$

\Rightarrow Period = 2.

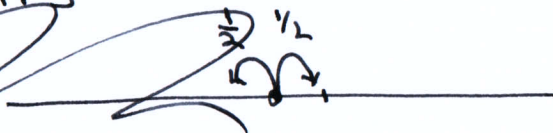
Means, ergodic, but still no unique stationary state.

~~Remark: An example where PF does not apply.~~

~~Random walk~~

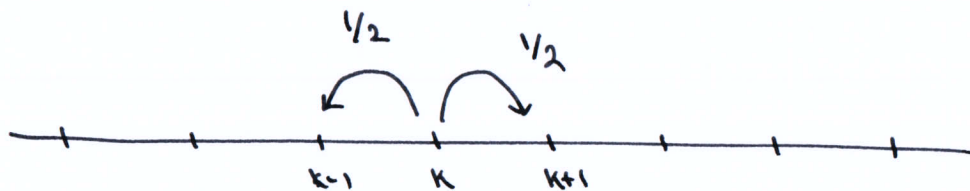
Remark: An example where PF does not apply.

Random walk on an infinite line.



~~100~~

An example where PF does not apply.



$$P_{t+\Delta t}(k) = \frac{1}{2} [P_t(k-1) + P_t(k+1)]$$

$$\Rightarrow |P_{t+\Delta t}\rangle = M |P_t\rangle \quad \text{with} \quad M = \begin{pmatrix} \ddots & & & & & & \\ & \ddots & & & & & \\ & & \frac{1}{2} & & 0 & & \\ & & & \ddots & & & \\ & & & & 0 & & \\ & & & & & \ddots & \\ & & & & & & \frac{1}{2} & \\ & & & & & & & \ddots \end{pmatrix}$$

because it is infinite.

(we solve this problem soon). (There is no stationary state!)

But, PF applies if it is on a periodic boundary.

$$M = \begin{pmatrix} 0 & \frac{1}{2} & 0 & \dots & 0 & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & \dots & \\ 0 & \dots & \dots & \dots & \dots & \\ \vdots & & & & & \frac{1}{2} \\ \frac{1}{2} & 0 & 0 & \frac{1}{2} & 0 & \dots \end{pmatrix} \quad \text{finite matrix.}$$

There is a unique stationary state, where all sites are equally probable.

Easy to see: M is symmetric. $\Rightarrow |l_0\rangle \equiv |r_0\rangle$

we know $|l_0\rangle = |1\rangle \Rightarrow |r_0\rangle = |2\rangle$.

Remark: PF is related to phase transition in equilibrium.

Phase transitions happen when PF does not apply for transfer matrix.

Ref: Cuesta & Sánchez, J Stat Phys (2004).

Sanyal, Klamys, Sadhu, Dhar, PRL (2018).

Continuous time process (but state space discrete)

$$M(c', c) = \delta_{c', c} + \Delta t \omega(c', c) + \mathcal{O}(\Delta t^2)$$

$$\begin{aligned} \Rightarrow P_{t+\Delta t}(c') &= \sum_c M(c', c) P_t(c) \\ &= \sum_c \left[\delta_{c', c} + \Delta t \omega(c', c) + \dots \right] P_t(c) \\ &= P_t(c') + \Delta t \sum_c \omega(c', c) P_t(c) + \mathcal{O}(\Delta t^2) \end{aligned}$$

$$\Rightarrow \frac{P_{t+\Delta t}(c') - P_t(c')}{\Delta t} = \sum_c \omega(c', c) P_t(c) + \mathcal{O}(\Delta t)$$

in $\Delta t \rightarrow 0$ limit

$$\boxed{\frac{dP_t(c')}{dt} = \sum_c \omega(c', c) P_t(c)}$$

OR

$$\boxed{\frac{d|P_t\rangle}{dt} = W|P_t\rangle}$$

Master equation,

Pauli-M equation,

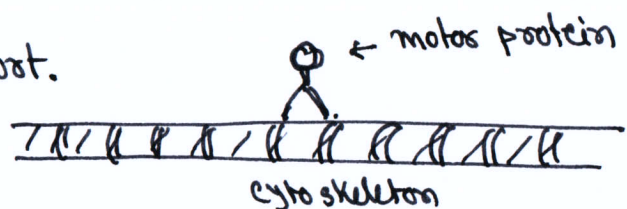
Rate equation.

Remark: $\omega(c', c)$ is transition rate, therefore can be larger than 1.

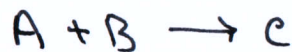
$\Delta t \cdot \omega(c', c)$ gives the probability of a transition for small Δt .

Remark: $\omega(c', c)$ can be computed ~~or~~ or measured ~~for a given~~ from the dynamics of a given system.

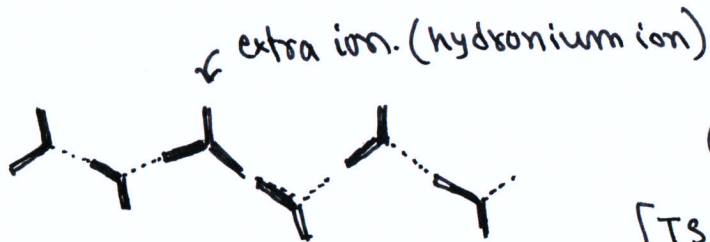
Example. Biological transport.



Example: Chemical reactions



Example: transport of ion in hydrogen bonded network.



Grotthuss mechanism.

[Ts. arxiv; 1009]

Example: Fermi's golden rule in QM.

time-dependent perturbation theory gives,

↙ perturbation.

$H_{\text{system}} + H_{\text{exterm.}}$

↳ E_n are eigen states of H_{system}

$$W(n', n) = \frac{2\pi}{\hbar} |H_{\text{Ext}}(n', n)|^2 \rho(E_n)$$

↖ matrix element in E_n basis.

↖ density of states.

Remark: ① spot the difference of $\frac{d|\rho_t\rangle}{dt} = W|\rho_t\rangle$ with Sch equation

$$i\hbar \frac{d|\psi_t\rangle}{dt} = H|\psi_t\rangle$$

② Compare with Boltzmann equation. See that B-equation is non-linear in phase space density, but M-equation is linear in p.

* A conventional form of the M-equation.

$$\sum_{c'} M(c', c) = 1$$

$$\Rightarrow 1 + \sum_{c'} w(c', c) = 1$$

$$\Rightarrow \boxed{\sum_{c'} w(c', c) = 0}$$

Column sum of W-matrix vanish.

A natural choice

$$w(c, c) = - \sum_{c' \neq c} w(c', c)$$

along the diagonal.

$$W = \begin{pmatrix} \dots & \dots & \dots & \dots & \dots \\ \dots & w_1 & \dots & \dots & \dots \\ \dots & \dots & w_2 & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \end{pmatrix}$$

$\rightarrow \sum w_i$
 w_4

Typically, this is incorporated in the M-equation by writing

$$\frac{d}{dt} P_+(e') = \sum_{c \neq c'} w(c', c) P_+(c) + w(c', c') P_+(e')$$

↓

$$- \sum_{c'' \neq c'} w(c'', c') P_+(e')$$

↓ c'' is dummy variable
then, call it c.

~~$$- \sum_{c \neq c'} w(c, c') P_+(e')$$~~

$$- \sum_{c \neq c'} w(c, c') P_+(e')$$

$$= \sum_{c \neq c'} \left[w(c', c) P_+(c) - w(c, c') P_+(e') \right]$$

↑ for c=c', is zero. so we can write

$$\boxed{\frac{d}{dt} P_+(e') = \sum_c \left[w(c', c) P_+(c) - w(c, c') P_+(e') \right]}$$