

Continuous time process (but state space discrete)

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

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

in $\alpha t \rightarrow 0$ limit

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

OR

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

Master equation,

Ponti-M equation,

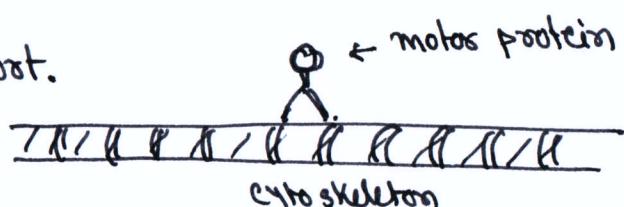
Rate equation.

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

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

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

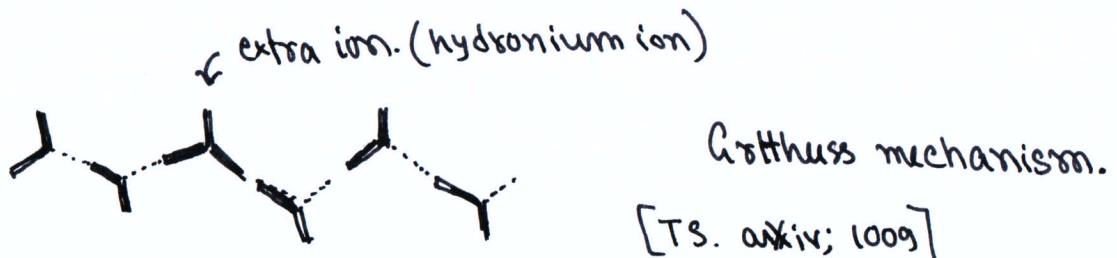
Example: Biological transport.



Example: chemical reactions



Example: transport of ion in hydrogen bonded network.



Example: Fermi's golden rule in QM.

time-dependent perturbation theory gives,

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

matrix element
in E_n basis.

↓ perturbation.
H_{system} + H_{external}.
↳ E_n are eigen states of H_{system}

↑ density of states.

Remark: ① Spot the difference of $\frac{d|p_t\rangle}{dt} = \omega|p_t\rangle$ 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

$$w(c, c) = - \sum_{c' \neq c} w(c', c) \quad \text{along the diagonal.}$$

$$W = \begin{pmatrix} & \cdots & w_1 & \cdots & \cdots \\ \cdots & w_2 & \ddots & & \\ & \vdots & \vdots & \ddots & \\ & & \vdots & \ddots & \\ & & & \ddots & \end{pmatrix}$$

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

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

↓

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

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

$$\cancel{+ \sum_{c'' \neq c'} w(c'', c') P_t(c')}$$

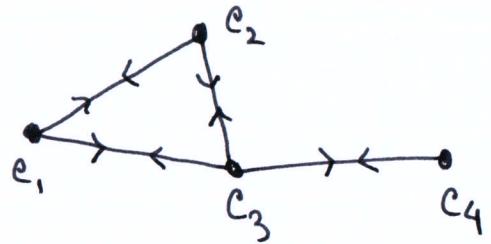
$$- \sum_{c \neq c'} w(c, c') P_t(c')$$

$$= \sum_{c \neq c'} \underbrace{[w(c', c) P_t(c) - w(c, c') P_t(c')]}_{\text{for } c=c', \text{ it is zero. so we can write}}$$

↑ for $c=c'$, it is zero. so we can write

$$\boxed{\frac{d}{dt} P_t(c') = \sum_c [w(c', c) P_t(c) - w(c, c') P_t(c)]}$$

stationary state, equilibrium, and out-of equilibrium.



Stationary state:

$$\frac{d}{dt} P_{st}(c') = 0 \quad \text{Prob does not change with time.}$$

$$\Rightarrow \sum_c [w(c', c) P_{st}(c) - w(c, c') P_{st}(c')] = 0$$

$$\Rightarrow \boxed{\sum_c w(c', c) P_{st}(c) = P_{st}(c') \sum_c w(c, c')} \quad \text{"global balance"}$$

incoming Prob flux = Out going Prob flux.



Equilibrium stationary state:

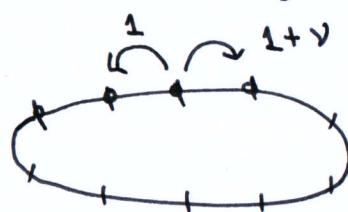
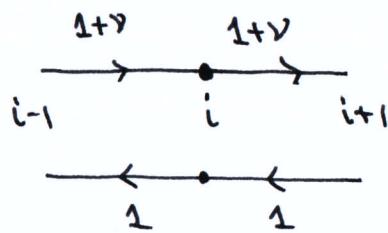
in flux = Out flux for each edge.

$$\boxed{w(c', c) P_{st}(c) = w(c, c') P_{st}(c')} \quad \text{Detailed balance.}$$

Out-of equilibrium:

Otherwise

Example: biased random walk on a ring.



Pair wise cancellation

$$\left. \begin{aligned} (1+v) P_{st}(i-1) &= (1+v) P_{st}(i) \\ \text{and} \quad 1 \cdot P_{st}(i) &= 1 \cdot P_{st}(i+1) \end{aligned} \right\} P_{st} = \text{uniform.}$$

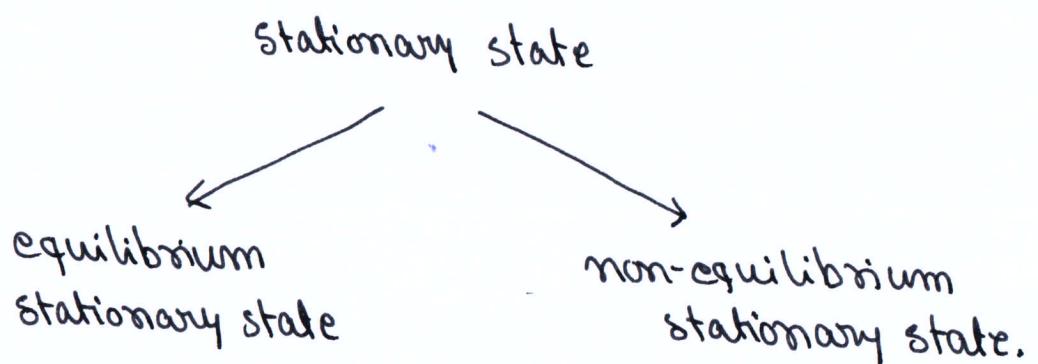
For $v=0$ %, detailed balance, P_{st} also uniform. But the state is in equilibrium.

Remark: Event ~~Markov~~ chain Monte Carlo. [Werner Krauth]

Remark: Stationary state does not mean that the system is frozen.

In fact system is still exploring its config space,
 but in a way that $P(c)$ ~~is~~ does not change with time.

Remark: System evolving towards an equilibrium state is also out-of-equilibrium.



Few useful facts about detailed balance.

① Detailed balance means \longleftrightarrow ~~separate states~~ zero probability current.

Also mean state is "statistically" time reversible.

(See soon, what it means)

② Often, given an equilibrium state, one can construct rates such that the $P_{st} \equiv P_{eq.}$. All that one needs is to satisfy detailed balance.

$$\frac{\omega(c', c)}{\omega(c, c')} = \frac{P_{eq}(c')}{P_{eq}(c)} = e^{-\beta [E(c') - E(c)]}$$

Central idea of Monte Carlo simulations!

A popular ~~choice~~ choice of such rates is

$$\omega(c', c) = \begin{cases} 1 & \text{if } E(c') \leq E(c) \\ e^{-\beta(E(c') - E(c))} & \text{if } E(c') > E(c). \end{cases}$$

Metropolis filter.

Example: Glauber spin flip dynamics of Ising model.
or
Kawasaki spin-exchange

Remark: In these, the dynamics need not be physical, ~~because~~ as long as we are interested only in the one-time (static) properties in equilibrium.

Ref: Event chain Monte Carlo for 2-step melting in two dimensions.

[Elior Bernaud, Werner Krauth]

Remark: $P_{st}(c) = e^{-\beta E(c)}$ does not mean equilibrium!

The prob current needs to vanish.

[Katz-Spohn-Lebowitz model: Ising spin under driving field]

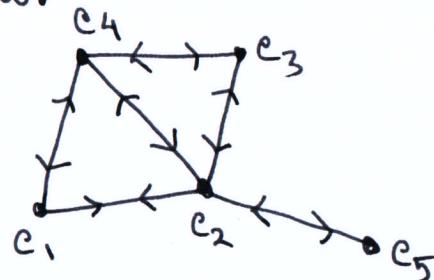
Exercise:

③ ~~for~~ Detailed balancee is stated in terms of stationary prob.

Can one test detailed balance without knowing the $P_{st} = P_{eq.}$?

Yes! by using Kolmogorov criteria.

You only need the rates w .



If for all loops on config space product of rates in

clockwise = anti clockwise

then, it is a necessary and sufficient condition for detailed balance.

necessary condition:

using

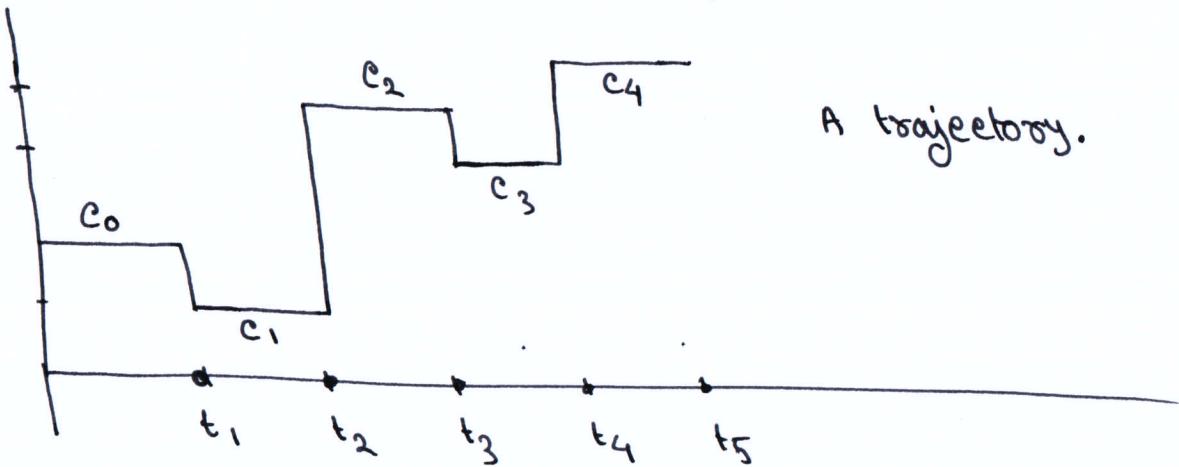
detailed balance.

$$\frac{w(c_1, c_4) w(c_4, c_3) w(c_3, c_2) w(c_2, c_1)}{w(c_4, c_1) w(c_3, c_4) w(c_2, c_3) w(c_1, c_2)} \stackrel{\text{using detailed balance}}{=} \frac{P(c'_1)}{P(c_1)} \cdot \frac{P(c'_4)}{P(c_4)} \cdot \frac{P(c'_3)}{P(c_3)} \cdot \frac{P(c'_2)}{P(c_2)} = 1$$

sufficient condition: See David Mukamel, ...

④ Time reversibility (stochastic) : Onsager

(17)



Prob to stay in a config(c) for time length t ,

$$= \lim_{\Delta t \rightarrow 0} \left[1 - \Delta t \sum_{c' \neq c} \omega(c', c) \right]^{t/\Delta t} = e^{-t \sum_{c' \neq c} \omega(c', c)} = e^{-t \omega(c, c)}$$

Prob for jump from $c \rightarrow c'$ is $\Delta t \cdot \omega(c', c)$

~~$$\Rightarrow \text{Prob}[c(t)] = P(c_0) \cdot e^{-t_1 \omega(c_0, c_0)} \cdot \Delta t \omega(c_0, c_1) \cdot e^{-t_2 \omega(c_1, c_1)} \cdot \Delta t \omega(c_1, c_2) \cdots \Delta t \omega(c_4, c_4)$$~~

$$\Rightarrow \text{Prob}[c(t)] = P(c_0) e^{-t_1 \omega(c_0, c_0)} \cdot \Delta t \omega(c_0, c_1) \cdot e^{-(t_2 - t_1) \omega(c_1, c_1)} \cdots \cdots e^{-(t_5 - t_4) \omega(c_4, c_4)}$$

Prob of time reversed trajectory

$$\text{Prob}[c^*(t)] = P(c_4) e^{-(t_5 - t_4) \omega(c_4, c_4)} \cdot \Delta t \omega(c_3, c_4) \cdots \cdots e^{-t_1 \omega(c_0, c_0)}$$

$$\Rightarrow \frac{\text{Prob}[c(t)]}{\text{Prob}[c^*(t)]} = \frac{P(e_0) \cdot w(e_1, e_0) \cdot w(e_2, e_1) \cdot w(e_3, e_2) \cdot w(e_4, e_3)}{P(e_4) \cdot w(e_3, e_4) \cdot w(e_2, e_3) \cdot w(e_1, e_2) \cdot w(e_0, e_1)}$$

↓ detailed balance

$$= \frac{P(e_0)}{P(e_4)} \cdot \frac{P(e_1)}{P(e_0)} \cdot \frac{P(e_2)}{P(e_1)} \cdot \frac{P(e_3)}{P(e_2)} \cdot \frac{P(e_4)}{P(e_3)} \\ = 1$$

$$\Rightarrow \boxed{\text{Prob of a forward trajectory} = \text{Prob of a time reversed trajectory}}$$

A consequence: In equilibrium, "the way" a fluctuation is spontaneously created, is "the same way" that fluctuation relaxes.

[Einstein, Onsager - Machlup]

Remark: Outside equilibrium, this symmetry breaks down.

But there is a more generalized symmetry.

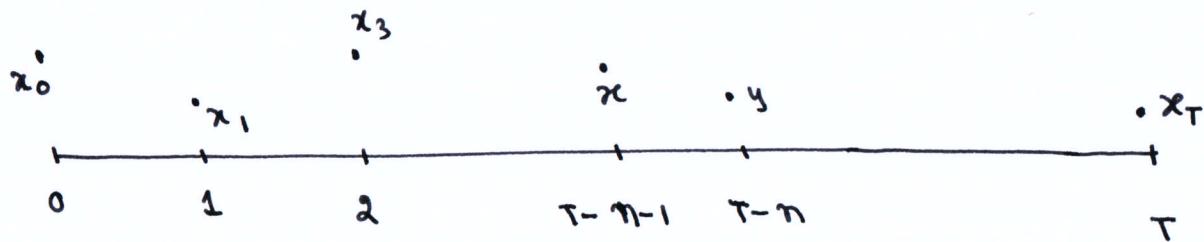
[Callanotti-Cohen fluctuation relations]

Remark: Fluctuation-dissipation relation in linear response is a consequence of this symmetry.

Time-reversed process

[Book by Stroock
An intro to Markov processes]

(19a)



Forward evolution:

$$\mathcal{C}(t) \in \left\{ c_0 = x_0, c_1 = x_1, \dots, c_T = x_T \right\}$$

Prob. evolves by $M(c', c)$

Time-reversed evolution:

$$c^*(t) = c(\tau-t)$$

$$c^*(t) = \left\{ c_0^* = x_\tau, c_1^* = x_{\tau-1}, \dots, c_T^* = x_0 \right\}$$

Q: What dynamics describe $c^*(t)$?

Is it Markovian?

Answer:

$$P^*\left(c_{n+1}^* = x, c_n^* = y, \dots, c_0^* = x_\tau\right) = P\left(c_{\tau-n-1} = x, c_{\tau-n} = y, \dots, c_\tau = x_\tau\right)$$

$$\Rightarrow M^*\left(c_{n+1}^* = x \mid c_n^* = y, \dots, c_0^* = x_\tau\right) P^*\left(c_n^* = y, \dots, c_0^* = x_\tau\right) = \dots$$

Also

$$P^*\left(c_n^* = y, \dots, c_0^* = x_\tau\right) = P\left(c_{\tau-n} = y, \dots, c_\tau = x_\tau\right)$$

• Taking ratio

$$\begin{aligned}
 M^*(c_{n+1}^* = x \mid c_n^* = y, \dots, c_0^* = x_T) \\
 &= \frac{P(c_{T-n-1} = x, c_{T-n} = y, \dots, c_T = x_T)}{P(c_{T-n} = y, \dots, c_T = x_T)} \\
 &= \frac{P(c_{T-n-1} = x) M(x \rightarrow y) M(y \rightarrow x_{T-n}) \dots M(x_{T-1} \rightarrow x_T)}{P(c_{T-n} = y) M(y \rightarrow x_{T-n}) \dots M(x_{T-1} \rightarrow x_T)} \\
 &= \frac{P(c_{T-n-1} = x)}{P(c_{T-n} = y)} M(y, x) \quad [M(x \rightarrow y) \equiv M(y, x)]
 \end{aligned}$$

This means that the time reversed dynamics is also Markovian (because it only depends on y , and not on $c_{n-1}^*, c_{n-2}^*, \dots, c_0^*$).

But $M^*(x, y)$ depends on time (n) \Rightarrow a time inhomogeneous process.

Remark: If the forward process has reached a stationary state, i.e. $P(c_n = x) = P_{st}(x)$, then

$$M^*(x, y) = \frac{P_{st}(x)}{P_{st}(y)} M(y, x)$$

time-homogeneous.
(check: $\sum_x M^*(x, y) = 1$)

Remark: note, generally $M^*(x, y) \neq M(x, y)$. This equality happens only for equilibrium by detailed balance condition.

Remark: continuous time case can be obtained by

$$M^*(x, y) = \sum_{z,y} + dt W^*(z, y) + \dots = \frac{P(x) + dt h(x)}{P(y) + dt h(y)} \cdot (S_{y,x} + dt W(y, x))$$