

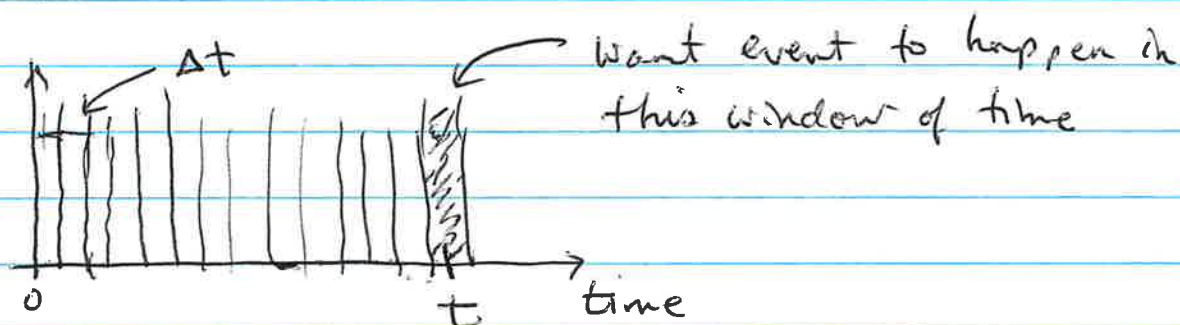
Continuous time Markov processes

Instead of discrete steps, consider continuous time with
No memory:

transitions from one state to the next (like $n \rightarrow n+1$) occur instantly at a specific time. Transitions are "events," which have a probability intensity, or prob per time (rate)

$$P(n \rightarrow n+1, t) dt \equiv p dt$$

when does event occur?



in each increment, there is a probability $p \Delta t$ that event happens

$t = N \Delta t$, $1 - p \Delta t \in$ prob. that event did not happen in Δt

if there are N steps where event did not happen, prob. is $(1 - p \Delta t)^N$

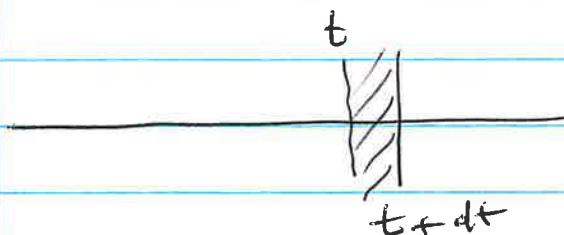
then, the event happens in the next Δt with probability $p \Delta t$

$\therefore P(t, t + \Delta t)$ of event happening is

Prob. event happens between t and $t+\Delta t$ is

$$(1 - p \frac{\Delta t}{N})^N p \Delta t \xrightarrow{N \rightarrow \infty} p e^{-p t} dt$$

"Waiting time" distribution $p e^{-p t} \equiv w(t)$



Suppose p depends on time $p = p(t)$

$$(1 - \frac{p_1 \Delta t}{N}) (1 - \frac{p_2 \Delta t}{N}) \dots \prod_{i=1}^N (1 - p_i \Delta t)$$

$$\ln(\pi) = \sum_i \ln(1 - p_i \Delta t) \rightarrow$$

generally,
waiting time distribution $w(t) = - \frac{dS}{dt} \Big|_t$

where $S(t)$ is the probability initial state survived (no transition) up to time t .

$$S(t) - S(t + \Delta t) = w(t) \Delta t$$

↓ $\rightarrow$ prob happened any time before t

$$[1 - P_{ext}(t)] - [1 - P_{ext}(t + \Delta t)] = w(t) \Delta t$$

$$\Rightarrow \underbrace{P_{ext}(t + \Delta t) - P_{ext}(t)}_{\text{prob of extinction up to time } t + \Delta t} = w(t) \Delta t$$

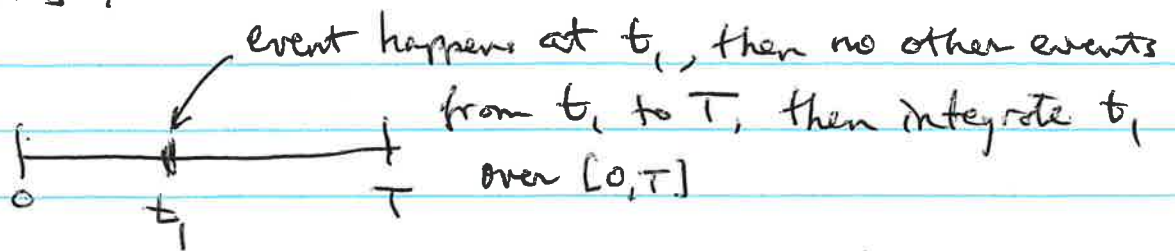
prob of extinction
up to time $t + \Delta t$

$$\therefore w(t) = - \frac{dS}{dt}, \quad S(t) = e^{-pt}$$

general,
not just
Markov

What is probability that only one event occurred in $[0, T]$?

(3)



prob no event up to t_1 : $S(t_1) = e^{-\lambda t_1}$

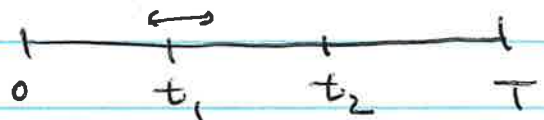
prob event at $t_1, t_1 + dt_1$: λdt_1

prob of no event from t_1 to T : $e^{-\lambda(T-t_1)}$

$$\therefore \int_0^T e^{-\lambda t_1} e^{-\lambda(T-t_1)} \lambda dt_1$$

$= \lambda e^{-\lambda T} T$ is prob. that only one event occurred in $[0, T]$

now, extend to two events



prob that only one event occurred before t_2

$$\lambda e^{-\lambda t_2} t_2$$

$\times$ prob that event occurs λdt_2
 $\quad \quad \quad \wedge$
 $\quad \quad \quad \text{second}$

$\times$ prob that no other event occurs between t_2 and T :

$$e^{-\lambda(T-t_2)}$$

$\therefore$ prob that exactly two events occurred in $[0, T]$:

$$\int_0^T \lambda t_2 e^{-\lambda t_2} e^{-\lambda(T-t_2)} \lambda dt_2 =$$

$$= p^2 \int_0^T e^{-pT} t_2 dt_2 = \frac{(pT)^2}{2} e^{-pT}$$

for n events: $\text{Prob}\{n \text{ events in } (0, T)\} = \frac{(pT)^n}{n!} e^{-pT} \quad n=0, 1, 2, \dots$

Poisson distribution

Waiting time distribution for completing n steps



total time $T = t_1 + t_2 + t_3 + \dots + t_n$ where each time is exponentially distributed.

consider two events $\int_0^{t_2} w_1(t_1) w_2(t_2 - t_1) dt_1 = \text{prob density that time of second event is between } t_2 \text{ and } t_2 + dt_2$

convolution: therefore we define $w_{12}(t_2) dt_2$ as prob second event occurs in $(t_2, t_2 + dt_2]$

Now, add a third event;

$$\int_0^{t_3} w_{12}(t_2) w_3(t_3 - t_2) dt_2 = \text{prob density of 3rd event at time } t_3$$

Another convolution

use Laplace transforms $\mathcal{L}\left\{\int_0^{t_2} w_1(t_1) w_2(t_2 - t_1) dt_1\right\}$
 $= \mathcal{L}\{w_1\} \mathcal{L}\{w_2\}$

$$\tilde{w}_1(s) \tilde{w}_2(s) = \tilde{w}_{12}(s)$$

$$\downarrow \int_0^\infty \underbrace{(p_1 e^{-p_1 t})}_{w_1(t)} e^{-st} dt \equiv \tilde{w}_1(s)$$

$$= \frac{p_1}{p_1 + s} \quad ; \quad w_{12}(s) = \frac{p_1 p_2}{(p_1 + s)(p_2 + s)}$$

$$w_{12}(t) = \frac{p_1 p_2}{p_2 - p_1} e^{-p_1 t} + \frac{p_1 p_2}{p_1 - p_2} e^{-p_2 t}$$

$$= \frac{p_1 p_2}{p_1 - p_2} \underbrace{(e^{-p_2 t} - e^{-p_1 t})}_{p_2 \rightarrow p_1} \longrightarrow e^{-p_2 t} (1 - e^{-(p_1 - p_2)t})$$

$$e^{-p_2 t} [1 - (1 - (p_1 - p_2)t + \dots)]$$

$$\Rightarrow \frac{p_1 p_2}{p_1 - p_2} (p_1 - p_2)t e^{-p_2 t} \rightarrow p_1^2 t e^{-p_1 t}$$

For n steps $w_{123\dots n}(t) = \sum_{i=1}^n \left(\prod_{j \neq i} \frac{p_j}{p_j - p_i} \right) p_i e^{-p_i t}$

intermediate "hidden" states can be used to generate arbitrary waiting times

$$E[T] = \sum_{i=1}^n \frac{1}{p_i}$$

$$\text{Var}[T] = \sum_{i=1}^n \frac{1}{p_i^2}$$

For all $p_i = p$,

$$W_{12...n}(t) = \frac{p^n t^{n-1} e^{-pt}}{(n-1)!}$$

Connect w/ Poisson $P(n \text{ events before } T)$

$$= P(S_n \leq T < S_{n+1})$$

where $S_n = t_1 + t_2 + \dots + t_n$

and $S_{n+1} = \cancel{t_1} + t_2 + \dots + t_n + t_{n+1} = S_n + t_{n+1}$ ↙ exponentially distributed

$$P(n \text{ event before } T) = P(S_n \leq T < S_n + t_{n+1})$$

for some fixed $S_n = s$, we want $\underbrace{t_{n+1}}_{e^{-p \cdot (T-s)}} > T-s$

Integrate over all possible $s \leq T$

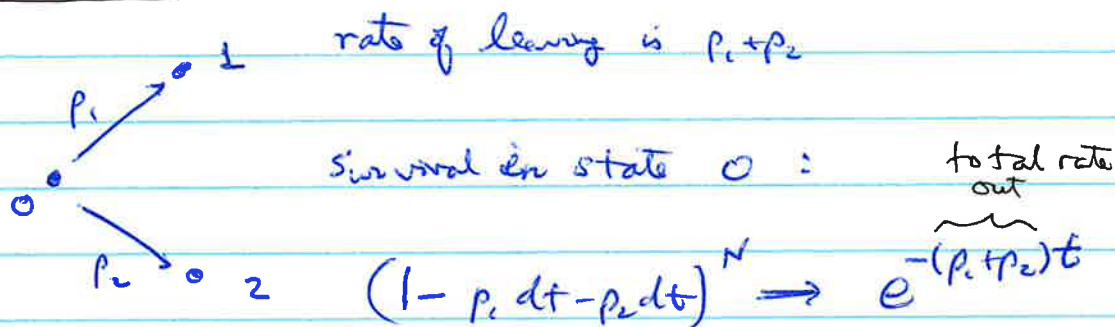
$$P(S_n \leq T < S_{n+1}) = \int_0^T W_{12...n}(s) e^{-p(T-s)} ds$$

$$P(n \text{ in time } T) = \int_0^T \frac{p^n s^{n-1}}{(n-1)!} e^{-ps} e^{-p(T-s)} ds$$

$$= e^{-pT} \int_0^T \frac{p^n s^{n-1}}{(n-1)!} ds$$

$$= e^{-pT} \frac{(pT)^n}{n!} \checkmark$$

Branching transitions



$$\text{total waiting time distribution} = -\frac{dS}{dt} = (p_1 + p_2) e^{-(p_1 + p_2)t}$$

What about conditional waiting time distributions?

$$W(t; 1) = W(t|1) P(1)$$

↓

$$(1 - p_1 dt - p_2 dt)^N p_1 = W(t|1) P(1), \quad P(1) = \int_0^\infty p_1 e^{-(p_1 + p_2)t} dt$$

$$p_1 e^{-(p_1 + p_2)t} = W(t|1) \left(\frac{p_1}{p_1 + p_2} \right) = \frac{p_1}{p_1 + p_2}$$

$$\Rightarrow W(t|1) = (p_1 + p_2) e^{-(p_1 + p_2)t} \quad \text{Same as } \underline{\text{unconditioned}}$$

Master equation discrete state n

$P(n, t | n_0, 0)$ prob state is n at time t given start at state n_0 at time $t=0$

$$P(n, t+dt | n_0, 0) = P(n, t | n_0, 0) \left(1 - \sum_m W_{mn} dt\right) + \sum_m P(m \neq n, t | n_0, 0) W_{nm} dt$$

$$\frac{P(n, t+dt | n_0, 0) - P(n, t | n_0, 0)}{dt}$$

$$= \sum_m P(m, t | n_0, 0) W_{nm} dt$$

$$- P(n, t | n_0, 0) \left(\sum_m W_{mn}\right) dt$$

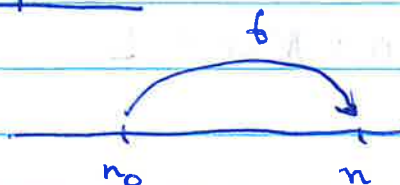
$$\Rightarrow \frac{\partial P(n, t | n_0, 0)}{\partial t} = \sum_m P(m, t | n_0, 0) W_{nm} - P(n, t | n_0, 0) \sum_m W_{mn}$$

in rate from all states m to state n

$$- P(n, t | n_0, 0) \sum_m W_{mn}$$

out rate to all states $m \neq n$

Backward picture



$$P(n, t | n_0, 0) = P(n, t-\Delta t | n_0, 0) \times \left(1 - \sum_m W_{mn_0} \Delta t\right)$$

$$+ \sum_k P(n, t-\Delta t | k, 0) W_{kn_0} \Delta t$$

$$P(n, t | n_0, 0) - \underbrace{P(n, t - \Delta t | n_0, 0)}_{\text{"backward"}} = - \left(\sum_m W_{mn_0} \right) \Delta t P(n, t - \Delta t | n_0, 0) + \sum_k P(n, t - \Delta t | k, 0) W_{kn_0} \Delta t$$

Shift time $t \rightarrow t + \Delta t$

$$\frac{P(n, t + \Delta t | n_0, 0) - P(n, t | n_0, 0)}{\Delta t} = \frac{\partial P(n, t | n_0, 0)}{\partial t}$$

$$= \sum_k W_{kn_0} P(n, t | k, 0) - P(n, t | n_0, 0) \sum_m W_{mn_0}$$

RHS operates on initial position

Useful for considering residence in subsets of states:

we can sum over a set of final states of interest:

$$\sum_{n \in N} P(n, t | n_0, 0) \equiv S(t | n_0, 0) \text{ survived within } N$$

the sum can be taken over final states over the Backward equation

$$\frac{\partial S(t | n_0, 0)}{\partial t} = \sum_k W_{kn_0} S(t | k, 0) - S(t | n_0, 0) \sum_m W_{mn_0}$$

Initial conditions $S(0 | n_0 \in N, 0) = 1$