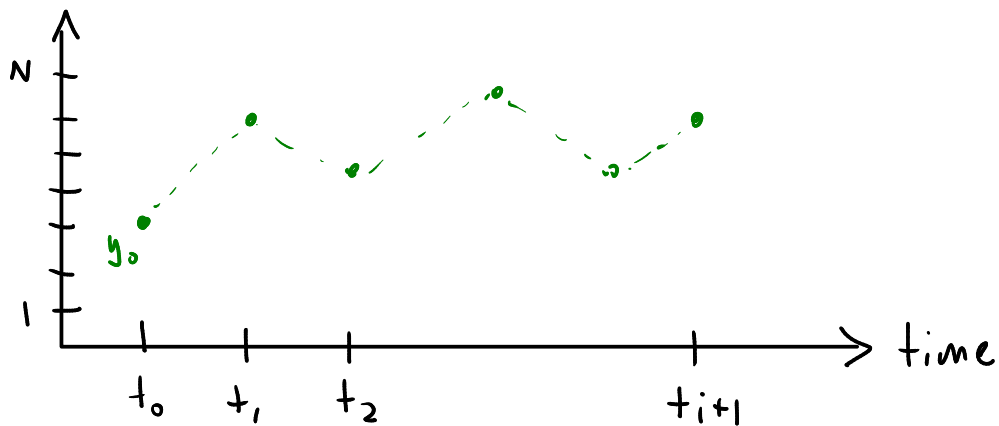


Recall: trajectory $v = (y_0, y_1, \dots)$



$\mathcal{P}(v)$ = prob. to observe v in ensemble

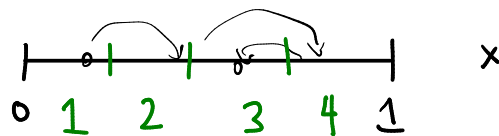
key assumption: Markovian dynamics

$$\mathcal{P}(y_{i+1} | y_0, y_1, \dots, y_i) \quad \text{prob. of } y_{i+1} \text{ given past history } y_0, \dots, y_i$$

$$= \mathcal{P}(y_{i+1} | y_i) \quad \text{only depends on immediate past state } y_i$$

example

from prob. 1
in PS #1



run simulations: traj.

$(1, 1, 2, 1, 3)$

$(1, 2, 3, 4, 2)$

...

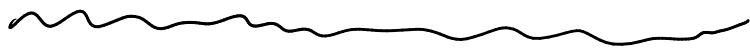
$(y_0, y_1, y_2, y_3, y_4)$

$$\mathcal{P}(y_4 = 3 | y_2 = 4, y_3 = 1) \stackrel{?}{=}_{\text{check}} \mathcal{P}(y_4 = 3 | y_3 = 1)$$

$$\frac{\# \text{ traj. of form } (*, *, 4, 1, 3)}{\# \text{ traj. of form } (*, *, 4, 1, *)} \stackrel{?}{=} \frac{(*, *, *, 1, 3)}{(*, *, *, 1, *)}$$

$$i\hbar \frac{\partial}{\partial t} |\psi(t)\rangle = \hat{H}(t) |\psi(t)\rangle \quad \left(\frac{\hat{H}(t)}{i\hbar} + \hat{I} \right) \Delta t |\psi(t)\rangle$$

$$|\psi(t+\Delta t)\rangle = \left(|\psi(t)\rangle + \frac{\hat{H}(t)}{i\hbar} |\psi(t)\rangle \right) \Delta t$$



Why is Markovianity useful?

1) makes calculating $\mathcal{P}(v)$ easier:

$$\mathcal{P}(A|B) = \frac{\mathcal{P}(A, B)}{\mathcal{P}(B)}$$

LHS:

$$\frac{\mathcal{P}(y_0, y_1, \dots, y_{i+1})}{\mathcal{P}(y_0, y_1, \dots, y_i)} \stackrel{\text{Markov}}{=} \mathcal{P}(y_{i+1} | y_i)$$

$$\mathcal{P}(y_0, y_1, \dots, y_{i+1}) = \mathcal{P}(y_{i+1} | y_i) \mathcal{P}(y_0, \dots, y_i)$$

$$\underbrace{\mathcal{P}(y_0, y_1, \dots, y_{i+1})}_{\mathcal{P}(v)}$$

= recursion

$$= \mathcal{P}(y_{i+1} | y_i) \mathcal{P}(y_i | y_{i-1}) \dots$$

$$= \left[\prod_{j=0}^i \mathcal{P}(y_{j+1} | y_j) \right] \mathcal{P}(y_0)$$

↓
initial dist.

$$y_{j+1} = 1, \dots, N \quad y_j = 1, \dots, N \quad N = \# \text{ states}$$

$$\mathcal{P}(y_{j+1} = n \mid y_j = m) \equiv W_{nm}(t_j)$$

"transition probability"

$W(t_j) = N \times N$ matrix
of probabilities

given m prob. of
going to n

(all real, b/t 0 and 1)

[not necessarily symmetric]

$$\Rightarrow \mathcal{P}(v) = W_{n_{i+1}n_i}(t_i) W_{n_i n_{i-1}}(t_{i-1}) \dots W_{n_1 n_0}(t_0) \mathcal{P}(n_0)$$

" "
($n_0, n_1, n_2, \dots, n_{i+1}$)

$$\mathcal{P}(A) = \sum_B \mathcal{P}(A, B)$$

2) we can find prob. of being in
some state at current time:

$$\mathcal{P}(y_{i+1}) = \sum_{y_0=1}^N \dots \sum_{y_i=1}^N \mathcal{P}(y_0, y_1, \dots, y_{i+1})$$

Markov

$$= \sum_{y_0} \dots \sum_{y_{i-1}} \sum_{y_i} \mathcal{P}(y_{i+1} \mid y_i) \mathcal{P}(y_0, \dots, y_i)$$

$$= \sum_{y_i} \mathcal{P}(y_{i+1} \mid y_i) \mathcal{P}(y_i)$$

rewrite in matrix-vector
form

$$\mathcal{P}(y_i = m) = p_m(t_i)$$

$$\mathcal{P}(y_{i+1} = n) = p_n(t_{i+1})$$

$$p_n(t_{i+1}) = \sum_{m=1}^N W_{nm}(t_i) p_m(t_i)$$

↓
elements of
 $\vec{p}(t_i)$ $\vec{p}(t_{i+1})$

⇒

$$\vec{p}(t_{i+1}) = W(t_i) \vec{p}(t_i)$$

discrete-state
discrete-time
master equation

$$\vec{p}(t_{i+1}) = W(t_i) \cdots W(t_1) W(t_0) \vec{p}(t_0)$$