

University of Washington
Department of Chemistry
Chemistry 553
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Lecture 8: Introduction to Conditional Probabilities

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A. Markov Processes and Conditional Probabilities

- For a completely random process, the condition of a system at a time t is independent of its history so

$$W_n(x_n, t_n; x_{n-1}, t_{n-1}; \dots x_0, t_0) = W_1(x_0, t_0) W_1(x_1, t_1) \dots W_1(x_n, t_n) \quad (8.1)$$

- The next level of complexity is called a Markov Process. To define such a process we first define a conditional probability

$P_n(x_n, t_n | x_{n-1}, t_{n-1}; \dots x_1, t_1; x_0, t_0)$ where $t_0 < t_1 < \dots < t_n$ which is the probability that a system is at x_n at t_n granted that it was at x_0 at t_0 , x_1 at t_1 , etc. Some useful properties of joint and conditional probabilities follow...

$$\begin{aligned} \circ W_n(x_n, t_n; x_{n-1}, t_{n-1}; \dots x_1, t_1; x_0, t_0) &= P_n(x_n, t_n | x_{n-1}, t_{n-1}; \dots x_1, t_1; x_0, t_0) \\ \circ &\times W_{n-1}(x_{n-1}, t_{n-1}; \dots x_1, t_1; x_0, t_0) \end{aligned} \quad (8.2)$$

○ Examples

$$W_2(x_1, t_1; x_0, t_0) = W_1(x_0, t_0) P_2(x_1, t_1 | x_0, t_0)$$

$$W_3(x_2, t_2; x_1, t_1; x_0, t_0) = W_2(x_1, t_1; x_0, t_0) P_3(x_2, t_2 | x_1, t_1; x_0, t_0)$$

(8.3)

- For a Markov process the probability that a system is at x_n at time t_n is only dependent upon its immediate and most recent history.

$$P_n(x_n, t_n | x_{n-1}, t_{n-1}; \dots x_1, t_1; x_0, t_0) = P_2(x_n, t_n | x_{n-1}, t_{n-1}). \quad (8.4)$$

- For a Markov Process

$$W_n(x_n, t_n; x_{n-1}, t_{n-1}; \dots x_1, t_1; x_0, t_0) = P_2(x_n, t_n | x_{n-1}, t_{n-1}) \times W_{n-1}(x_{n-1}, t_{n-1}; \dots x_1, t_1; x_0, t_0)$$

$$\therefore W_2(x_1, t_1; x_0, t_0) = P_2(x_1, t_1 | x_0, t_0) \times W_1(x_0, t_0)$$

(8.5)

- For stationary Markov Processes

$$P_2(x_n, t_n | x_{n-1}, t_{n-1}) = P_2(x_n | x_{n-1}, t_n - t_{n-1}) = P_2(x_n | x_{n-1}, \tau) \quad (8.6)$$

$$\therefore W_2(x_1, t_1; x_0, t_0) = W_2(x_1; x_0, \tau) = W_1(x_0) P_2(x_1 | x_0, \tau)$$

- The conditional probability plays an important role in non-equilibrium statistical mechanics. It is used in the expression for the correlation function for property A whose change in time is described as a stationary Markov process:

$$\begin{aligned}
 K(\tau = \Delta t) &= \int \cdots \int dp_0 dq_0 dp_1 dq_1 W_2(p_1, q_1; p_0, q_0, \tau) A(p_0, q_0) A(p_1, q_1) \\
 &= \int \cdots \int dp_0 dq_0 dp_1 dq_1 W_1(p_0, q_0) P_2(p_1, q_1 | p_0, q_0, \tau) A(p_0, q_0) A(p_1, q_1)
 \end{aligned}
 \tag{8.7}$$

D. How to Obtain Conditional Probabilities: Master Equations

- Conditional probabilities are calculated using differential equations called master equations. Master equations may be thought of as rate equations and they utilize transition rate probabilities. Example: $w_{j \rightarrow l} d\tau dl = w_{jl} d\tau dl$ is the probability that in a short period of length $d\tau$, a transition is made j to $l+dl$. As such w_{jl} is a transition rate probability. A pleasant feature of master equations is that their derivation and use is pretty clear to a chemist.
- To obtain the master equation we need a mathematical relationship called Markov's Integral Equation, also called the Chapman-Kolmogorov (CK) Equation:

$$\begin{aligned}
 P_2(x_2, t_2 | x_0, t_0) &= \int P_2(x_1, t_1 | x_0, t_0) P_2(x_2, t_2 | x_1, t_1) dx_1 \\
 &= \int P_2(x_1 | x_0, t_1 - t_0) P_2(x_2 | x_1, t_2 - t_1) dx_1 \\
 &= \int P_2(x_1 | x_0, t) P_2(x_2 | x_1, \tau) dx_1 = P_2(x_2 | x_0, t + \tau)
 \end{aligned}
 \tag{8.8}$$

where the second and third steps assume the process is Markovian and stationary. We therefore define $\tau = t_2 - t_1$ and $t = t_1 - t_0$

- Equation 8.8 has the discrete form:

$$P_2(l | l_0; t + \tau) = \sum_j P_2(j | l_0; t) P_2(l | j; \tau)
 \tag{8.9}$$

- We also need to take note of two important properties possessed by conditional probabilities:
 - The system or property A has to make a transition to some other state in a time $t_2 - t_1 = \tau$: $\int P(l | j, \tau) dl = 1$. For discrete states this can be written as

$$\sum_l P(l | j, \tau) = 1$$
 - The system or property cannot change instantaneously: $P(l | j, \tau = 0) = \delta_{jl}$

Note for convenience we dropped the subscript 2, because it is understood that from now on we deal with stationary Markov processes.

- Three steps to setting up a master equation...

○ Because a system cannot change instantaneously

$$P(l|j, \tau) \rightarrow 0 \text{ as } \tau \rightarrow 0. \quad \therefore P(l|j, \tau) \approx \tau w_{j \rightarrow l} \quad (8.10a)$$

$$\text{and } P(l|l, \tau) \rightarrow 1 \text{ as } \tau \rightarrow 0. \quad \therefore P(l|l, \tau) \approx 1 - \tau \sum_{k \neq l} w_{l \rightarrow k} \quad (8.10b)$$

$$\text{○ Combine these two equations: } P(l|j, \tau) = \tau w_{j \rightarrow l} + \delta_{lj} \left(1 - \tau \sum_k w_{l \rightarrow k} \right) \quad (8.10c)$$

• We take (8.10c) and put it into the discrete form of the C-K equation to obtain

$$\frac{P(l|l_0; t + \tau) - P(l|l_0; t)}{\tau} = \sum_j \left[w_{j \rightarrow l} P(j|l_0; t) - w_{l \rightarrow j} P(l|l_0; t) \right] \quad (8.11)$$

• (8.11) can be written as a differential equation in the limit $\tau \rightarrow 0$

$$\lim_{\tau \rightarrow 0} \frac{P(l|l_0; t + \tau) - P(l|l_0; t)}{\tau} = \frac{\partial P(l|l_0; t)}{\partial t} = \sum_j \left[w_{j \rightarrow l} P(j|l_0; t) - w_{l \rightarrow j} P(l|l_0; t) \right] \quad (8.12)$$

which in continuous form is: $\frac{\partial P(l|l_0; t)}{\partial t} = \int_{-\infty}^{+\infty} \left[w_{j \rightarrow l} P(j|l_0; t) - w_{l \rightarrow j} P(l|l_0; t) \right] dj$