LECTURE #4

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Problem

Given the master equations (MEs)

$$\frac{\partial p_{Z}(\mathbf{z};t)}{\partial t} = \sum_{m=1}^{M} \left\{ a_{m}(\mathbf{z} - \mathbf{e}_{m}) p_{Z}(\mathbf{z} - \mathbf{e}_{m};t) - a_{m}(\mathbf{z}) p_{Z}(\mathbf{z};t) \right\}, \quad t > 0$$

$$\frac{\partial p_{X}(\mathbf{x};t)}{\partial t} = \sum_{m=1}^{M} \left\{ \pi_{m}(\mathbf{x} - \mathbf{s}_{m}) p_{X}(\mathbf{x} - \mathbf{s}_{m};t) - \pi_{m}(\mathbf{x}) p_{X}(\mathbf{x};t) \right\}, \quad t > 0$$

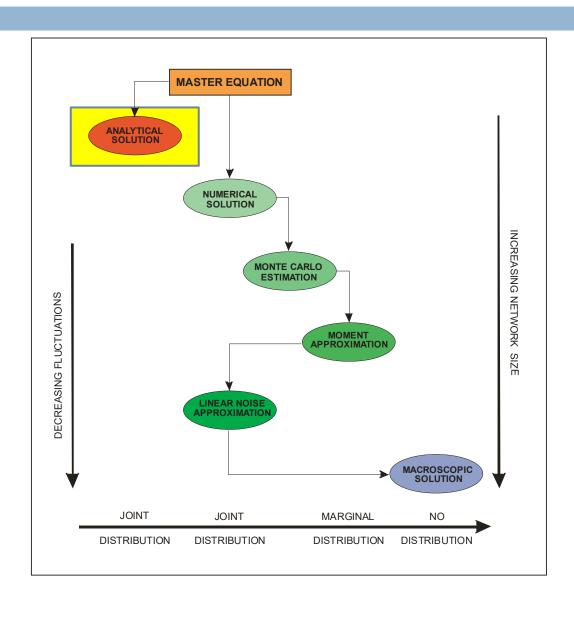
we want to calculate the probabilities $p_X(\mathbf{x};t)$ and $p_Z(\mathbf{z};t)$, for every t>0 .

This is a <u>difficult</u> task in general !!

Important Theoretical Result

- Markovian reaction networks with mass action propensities have been theoretically shown to perform <u>Turing universal computations</u> with an error that becomes zero only at the thermodynamic limit (at which <u>no randomness</u> is present).
- This amazing result turns out to be one of the most profound weaknesses of Markovian reaction networks: there will be no single analytical or even computational method capable of calculating the <u>exact</u> solution of the underlying master equation in <u>complete generality</u> away from the thermodynamic limit using finite resources.
- This is a consequence of the fact that any problem that is not computable by the universal machine is considered <u>not to be computable</u> by <u>any</u> machine.
- Consequently, developing accurate and computationally feasible techniques for studying the dynamic behavior of large nonlinear Markovian reaction networks is still the <u>most important</u> and <u>challenging</u> problem.

Available Methods



- Possible only in some simple cases.
- An analytical solution can also be derived in the case of a <u>linear</u> reaction network (a network with <u>linear</u> propensity functions).
- The probability distribution $p_X(\mathbf{x};t)$ of the population process in a <u>linear</u> reaction network can be expressed as the <u>convolution</u> of <u>multinomial</u> and <u>Poisson</u> distributions with time-dependent parameters that evolve according to well-defined systems of first-order <u>linear</u> differential equations.

- Example (birth and death process closed system)
 - Consider the following <u>linear</u> reaction network:

$$X_1 \to X_2$$
 (birth reaction)
 $X_2 \to X_1$ (death reaction) $S = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$

The (mass-action) propensity functions are given by:

Kronecker delta
$$\pi_1(x_1,x_2)=\kappa_1x_1\delta(N-x_1-x_2)$$

$$\pi_2(x_1,x_2)=\kappa_2x_2\delta(N-x_1-x_2)$$
 Kronecker delta

■ Mass conservation equation: $x_1 + x_2 = N$ (system is closed).

- Example (birth and death process closed system)
 - The network is governed by the following population ME:

$$\frac{\partial p_X(x_1, x_2; t)}{\partial t} = \left[\kappa_1(x_1 + 1) p_X(x_1 + 1, x_2 - 1; t) + \kappa_2(x_2 + 1) p_X(x_1 - 1, x_2 + 1; t) - (\kappa_1 x_1 + \kappa_2 x_2) p_X(x_1, x_2; t) \right] \delta(N - x_1 - x_2)$$
Kronecker delta

A solution can be found in certain circumstances by transforming the ME using the <u>probability generating function</u> and by solving the resulting equation.

- Example (birth and death process closed system)
 - If $X_1(0) = 0$ and $X_2(0) = N$, then the solution is given by:

$$p_{x}(x_{1}, x_{2}; t) = \binom{N}{x_{1}} [p(t)]^{x_{1}} [1 - p(t)]^{x_{2}} \delta(N - x_{1} - x_{2})$$

$$p(t) = \frac{\kappa_{2}}{\kappa_{1} + \kappa_{2}} [1 - e^{-(\kappa_{1} + \kappa_{2})t}]$$
Kronecker delta

binomial distribution

- Example (birth and death process closed system)
 - If $X_1(0) = N$ and $X_2(0) = 0$, then the solution is given by:

$$p_{x}(x_{1}, x_{2}; t) = \binom{N}{x_{1}} [q(t)]^{x_{1}} [1 - q(t)]^{x_{2}} \delta(N - x_{1} - x_{2})$$

$$q(t) = \frac{\kappa_{2} + \kappa_{1} e^{-(\kappa_{1} + \kappa_{2})t}}{\kappa_{1} + \kappa_{2}}$$
Kronecker delta

binomial distribution

Example (birth and death process – closed system)

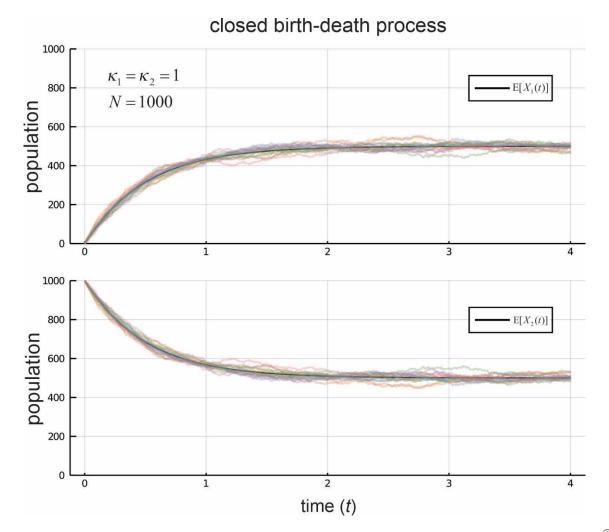
- If none of the previous initial conditions is true, then the solution cannot be found analytically.
- However, it can be shown that

$$\lim_{t\to\infty} p_x(x_1,x_2;t) = \binom{N}{x_1} \overline{p}^{x_1} (1-\overline{p})^{x_2} \, \delta(N-x_1-x_2)$$
 binomial distribution
$$\overline{p} = \lim_{t\to\infty} p(t) = \lim_{t\to\infty} q(t) = \frac{\kappa_2}{\kappa_1 + \kappa_2}$$
 Kronecker delta distribution

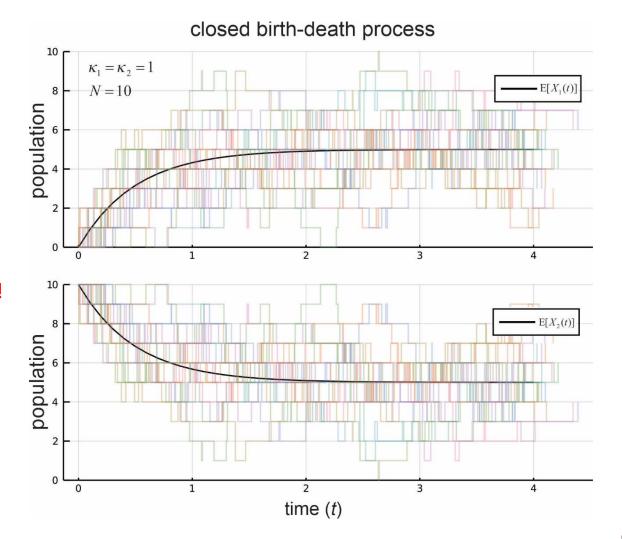
regardless of the initial condition.

It is easier to derive the solution of the ME at steady-state !!









Example (birth and death process – open system)

Consider now the following <u>linear</u> reaction network:

$$\varnothing \to X$$
 (birth reaction)
 $X \to \varnothing$ (death reaction) $S = \begin{bmatrix} 1 & -1 \end{bmatrix}$

The (mass-action) propensity functions are given by:

$$\pi_1(x) = \kappa_1$$

$$\pi_2(x) = \kappa_2 x$$

It is assumed here that reactants enter the system from its surroundings, and that there is an <u>unlimited</u> number of such reactants.

- Example (birth and death process open system)
 - The network is governed by the following population ME:

$$\frac{\partial p_X(x;t)}{\partial t} = \left[\kappa_1 p_X(x-1;t) + \kappa_2(x+1)p_X(x+1;t) - (\kappa_1 + \kappa_2 x)p_X(x;t)\right]$$

- Example (birth and death process open system)
 - If X(0) = 0, then the solution is given by:

$$p_{x}(x;t) = e^{-\lambda(t)} \frac{\left[\lambda(t)\right]^{x}}{x!}$$
$$\lambda(t) = \frac{\kappa_{1}}{\kappa_{2}} (1 - e^{-\kappa_{2}t})$$

Poisson distribution

- Example (birth and death process open system)
 - If $X(0) \neq 0$, then the solution <u>cannot be found</u> analytically.
 - However, it can be shown that

$$\lim_{t \to \infty} p_x(x;t) = e^{-\overline{\lambda}} \frac{\overline{\lambda}^x}{x!}$$

$$\overline{\lambda} = \lim_{t \to \infty} \lambda(t) = \frac{\kappa_1}{\kappa_2}$$
Poisson distribution

regardless of the initial condition.

