

Integrating the Gillespie Algorithm with Swarm

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Motivation (Systems Biology)

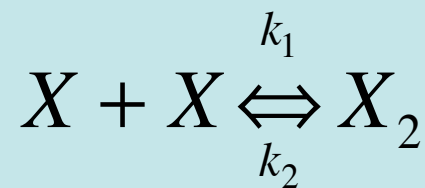
- ABM biological cell behaviors
 - Growth, movement, division, death
 - Interactions: cell-cell and cell-environment
 - Individual cell behaviors produce emergence of population level behavior (tissue function).
- Add more detail, selectively
 - Cell-cell interaction (Notch signaling pathway)
 - Insert gene network within each cell to “drive” one of the cell behaviors
 - Leave the other cell behaviors as ABM actions

Motivation (Systems Biology)

- Matching time scales (spatial scales)
 - Multi-scale, behaviors operating at different time scales
- Gene network simulation
 - Ordinary Differential Equations
 - Continuous, deterministic, population averages
 - Easier: numerically integrate over any time frame
 - Implicit separation of time scales (morpho-static limit)
 - Stochastic chemical kinetics
 - Exact for low molecule counts
 - Strong statistical correlations can be formed

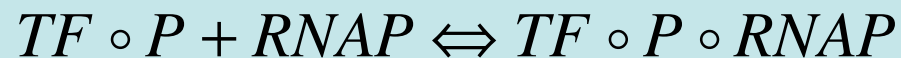
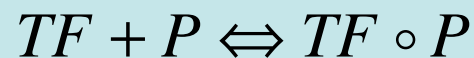
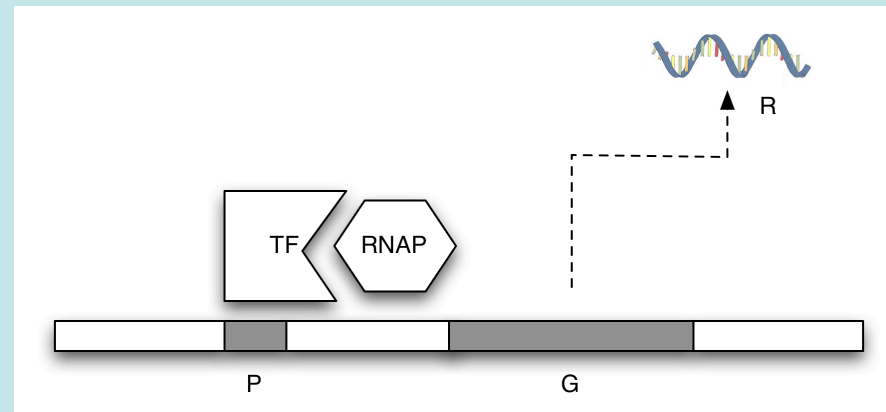
Stochastic Chemical Kinetics

- Rule-based description



Stochastic Chemical Kinetics

- Gene transcription



Gillespie Algorithm (GSA)

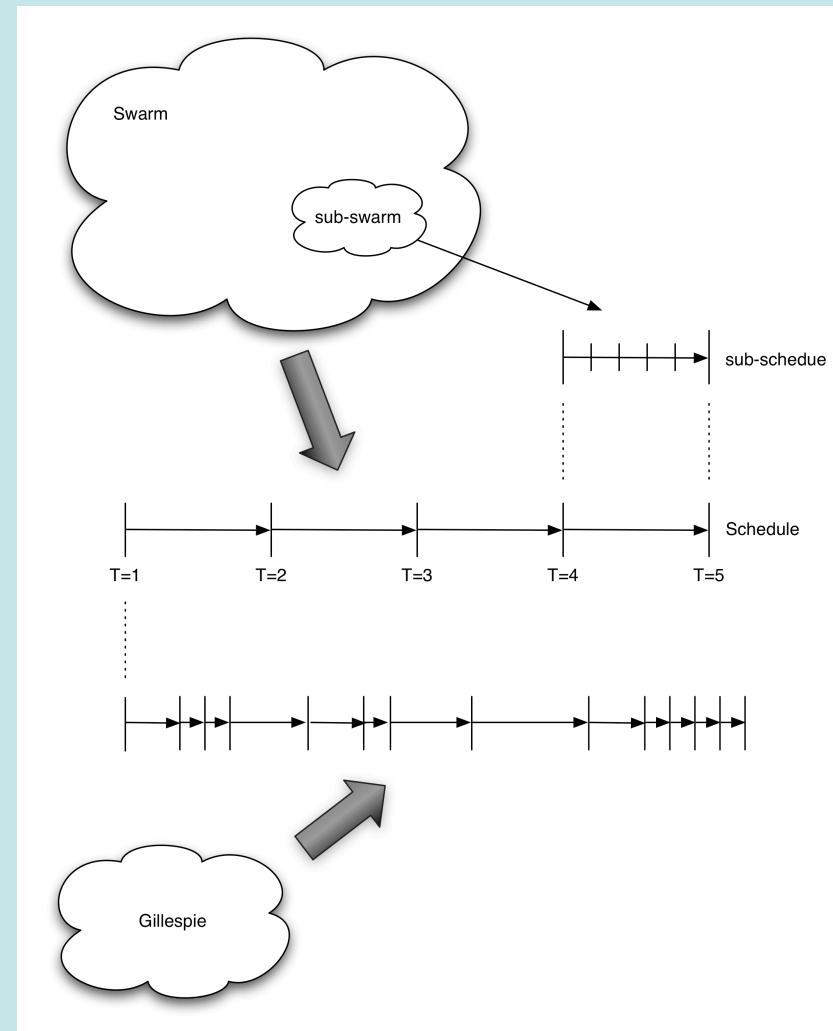
- Exact simulation of stochastic chemical kinetics (Gillespie 1977)
- Calculate propensity functions for each rule, which are normalized into probabilities
 - The larger the molecule count, the greater propensity for the rule.
- Rules occur at a rate per unit time according to Poisson distribution (basis in physical law).
 - However, non-homogenous Poisson process because rate changes with quantity of molecules
- Time of next reaction: Exponential distribution
- Randomly pick one reaction based upon rule probabilities
- Execute rule: update molecule counts
- Recalculate rule probabilities
- Repeat

Gillespie Algorithm (GSA)

- Similar to agent-based simulation
 - Discrete entities (molecules)
 - Probabilistic rules of behavior
- Key differences
 - Entities (molecules) are not distinct agents, it is assumed that they indistinguishable from each other.
 - Thus only need to maintain counts (totals).
 - Time move forwards in random steps (exponential distribution)
 - Time scale is defined by the number of molecules and rule rates
 - Many molecules/fast rates: small time scale
 - Few molecules/slow rates: large time scale

Matching Time Scales

- ABM: discrete time
 - Consistent sub-intervals for sub-swarms
- GSA: random continuous time
- If this was the only issue, then integration can be handled:
 - Run GSA
 - If time for next reaction exceeds ABM next time
 - Then save next reaction time, perform ABM rules, go back to GSA



Other Problems

- ABM rules based upon molecule count
 - If ($X > 100$) then perform ABM rule
 - Molecule counts can fluctuate greatly within a unit time interval, events can be missed if only check at time boundary
- How long does an ABM rule “take”?
 - ABM typically assume each rule is performed instantaneously.
 - Biological processes like cell division, movement, etc. are complicated tasks that can take a long time compared individual chemical reactions.

Solution 1

- Approximate GSA
 - Convert GSA to unit time (time discretization)
 - Others: Tau-leap method, Langevin equation
- Time Discretization
 - Do not simulate each individual reaction.
 - Random draw from Poisson distribution for each reaction
 - Count of occurrence for each reaction
 - Apply all rules at each unit time

Time Discretization (Solution 1)

- Bring GSA “up” to ABM time scale
- Pros
 - Computationally efficient
- Cons
 - Potentially lose stochastic nature
 - Inaccuracy increases with longer time intervals
 - Invalid conditions: negative molecule counts

Solution 2

- Bring ABM “down” to GSA time scale
- Rewrite ABM rules just like GSA rules with appropriate rates.
- Just run Gillespie algorithm for everything
- No more discrete time, simulation is completely random continuous time

Solution 2

- Bring ABM “down” to GSA time scale
- Pros
 - It is “exact” (though ABM might not be Poisson)
- Cons
 - Computationally expensive
 - Can be hard to determine ABM rule rates
 - ABM rules can essentially become rare events
 - Somewhat defeats the purpose of detailed interactions “driving” an ABM rule

Solution 3

- Both solution 1 and 2 attempt to collapse the multi-scale problem into single time scale.
- Time Integral GSA
 - Normal GSA, simulate each reaction
 - Estimate the likelihood of a random variable based upon the reactions performed during time step.

$$L(r; X, C) = \frac{1}{Z} \int_{t_0}^{t_1} p_k(X, C) dT = \frac{1}{Z} \sum_{i=0}^{n_i} p_k(X, C) [t_i - t_{i-1}]$$

- Rewrite ABM rule
 - If (random[0,1] < r) then perform ABM rule

Solution 3

- Time Integral GSA
- Cons
 - Might not be obvious exactly what integral to calculate, or proper normalization.
- Pros
 - Many ABM rules have probabilistic rate anyways, so this ties that probability to a detailed underlying stochastic process.
 - Agent heterogeneity introduced by stochastic process that can change over time, versus random initial condition.

Summary

- Multi-scale requires careful consideration.
 - Interaction between scales
 - Just considered temporal, spatial introduces additional issues
- Techniques that attempt to collapse scales suffers from problems.
 - “up” is approximate, lose stochastic detail
 - “down” is computationally expensive
- Maintain the separation of scales. Define a functional relationship between the lower level process and the parameter (decision process) at the higher level.
 - Time Integral GSA
- Issue of non-instantaneous higher level behaviors is not resolved, not an issue if behavior is independent of lower level.
 - Cell movement, non-movement related intercellular process