

Grammar Rules (Extended)

How do we write and interpret them?

- History, dynamic grammars¹:

$$\{\tau_{\alpha(p)}[x_p] \mid p \in L_r\}^* \longrightarrow \{\tau_{\beta(q)}[x_q] \mid q \in R_r\}^*$$

with $\rho_r([x_p], [y_q])$

$$\{\tau_{\alpha(p)}[x_p] \mid p \in L_r = R_r\}^* \longrightarrow \{\tau_{\beta(q)}[x_q] \mid q \in R_r = L_r\}^*$$

solving $\left\{ \frac{dx_{p,j}}{dt} = v_{p,j}([x_k]) \mid p, j \right\}$.

- Rate function factorization¹:

$$\rho_r([x_p]) \equiv \int \rho_r([x_p], [y_q]) \Delta[y_q]$$

$$P([y_q] \mid [x_p]) \equiv \frac{\rho_r([x_p], [y_q])}{\rho_r([x_p])}$$

$$\rho_r([x_p], [y_q]) \equiv \rho_r([x_p]) \times P([y_q] \mid [x_q])$$

- Simplified DGG graph notation², where λ is a label vector:

$$G\langle\langle\lambda\rangle\rangle \longrightarrow G'\langle\langle\lambda'\rangle\rangle \quad \text{with } \rho_r \text{ or solving } \dot{x} = v$$

Stochastic Growth Rule:

Left-hand Side (LHS)

$$(\bigcirc_1 \text{ --- } \bullet_2) \langle\langle(\mathbf{x}_1, \mathbf{u}_1), (\mathbf{x}_2, \mathbf{u}_2)\rangle\rangle$$

Right-hand Side (RHS)

$$\longrightarrow (\bigcirc_1 \text{ --- } \bigcirc_3 \text{ --- } \bullet_2) \langle\langle(\mathbf{x}_1, \mathbf{u}_1), (\mathbf{x}_2, \mathbf{u}_2), (\mathbf{x}_3, \mathbf{u}_3)\rangle\rangle$$

with $\hat{\rho}_{\text{grow}} \times H(\|\mathbf{x}_2 - \mathbf{x}_1\|; L_{\text{div}})$

where $\begin{cases} \mathbf{x}_3 = \mathbf{x}_2 - (\mathbf{x}_2 - \mathbf{x}_1)\gamma \\ \mathbf{u}_3 = \frac{\mathbf{x}_3 - \mathbf{x}_2}{\|\mathbf{x}_3 - \mathbf{x}_2\|} \end{cases}$

ρ_r in factored form,
where new labels are
sampled in the where
clause.

Improving the Exact Algorithm

Proposal for potential points of parallelization

- Parallelizable over propensity calculations.
- Parallelizable over propensity sums.
- Parallelizable over ODE solving when appropriate scaling and resources are available.
- Reactions must be still fired in order.
- Has potential for quick testing of medium-sized systems.
- Could be used to incorporate hierarchical parallelism in the approximate algorithms.

Parallel Exact Hybrid Parametrized SSA/ODE Algorithm

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factor  $\rho_r([x_p], [y_q]) = \rho_r([x_p]) * P([y_q] | [x_p]);$   
while  $t \leq t_{max}$  do  
  ParFor initialize SSA propensities as  $\rho_r([x_p]);$   
  ParReduce initialize  $\rho^{(total)} := \sum_r \rho_r([x_p]);$   
  initialize  $\tau := 0;$   
  draw effective waiting time  $\tau_{max}$  from  $\exp(-\tau_{max});$   
  while  $\tau < \tau_{max}$  do  
    ParFor solve ODE system, plus an extra ODE updating  $\tau;$   
     $\frac{d\tau}{dt} = \rho^{(total)}(t);$   
  draw reaction  $r$  from distribution  $\rho_r([x_p])/\rho^{(total)};$   
  draw  $[y_q]$  from  $P([y_q] | [x_p])$  and execute reaction  $r;$ 
```