

# Parallelism and Concurrency of Stochastic Graph Transformations

Reiko Heckel, University of Leicester  
Joint work with Hartmut Ehrig,  
Ulrike Golas, Frank Hermann

IFIP WG 1.3 Meeting, Aussois 06/01/2011

# Motivation

- ✘ Quantitative analysis of processes with dynamic reconfigurations, modelled as stochastic graph transformations
- ✘ Analysis through
  - Model checking: limited in scale
  - Stochastic simulation: statistical results only
  - *Symbolic calculation*

**Question:** What is the *distribution of completion times of a stochastic process* (deterministic, concurrent), given the *distribution of delays of its constituent steps*.

# Basic Notions of Probability

Delays and durations are real-valued random variables, known to fall into certain intervals with probabilities described by distributions.

- $F : \mathbb{R}_+ \rightarrow [0, 1]$  is distribution of real-valued random variable  $v$  if  $F(x)$  is the probability for  $v \leq x$ .
- A value chosen randomly based on  $F$  is denoted  $v = RN_F$ .
- Sum and maximum of independent random variables correspond to operations on distributions, such as convolution

$$F_1 * F_2(u) = \int_{-\infty}^{\infty} F_1(x)F_2(u - x)dx$$

and product

$$F_1 \cdot F_2(x) = F_1(x) \cdot F_2(x)$$

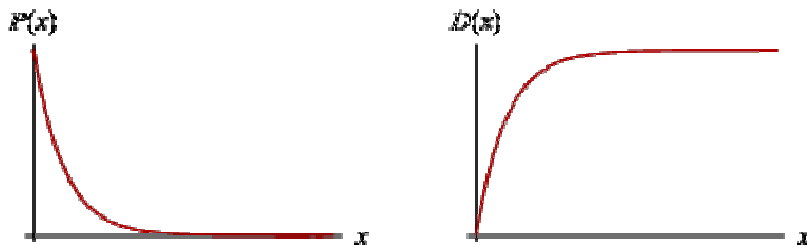
# Stochastic Graph Transformation Systems

$$\mathbf{SG} = (\mathbf{TG}, \mathbf{P}, \pi, \mathbf{F})$$

- **TG** – type graph
- **P** – set of rule names
- $\pi(\mathbf{p}): \mathbf{L} \rightarrow \mathbf{R}$  – rules typed over **TG**
- $\mathbf{F}: \mathbf{P} \rightarrow (\mathbb{R} \rightarrow [0,1])$  – distribution functions for delay

Exponential distribution

- given by rate  $\lambda$   
→  $1/\lambda$  avg. delay



Normal distribution

- given by mean and deviation



# Generalised Semi-Markov Processes

- ✘ Transitions do not depend on *history prior to the current state*

$$\mathcal{P} = \langle \begin{array}{l} S : \text{State set} \\ E : \text{Event set} \\ \Gamma : \text{State} \rightarrow \wp(\text{Event set}) \\ \Sigma : \text{State} \times \text{Event} \rightarrow \text{State} \\ \Delta : \text{Event} \rightarrow (\mathbb{R} \rightarrow [1, 0]) \\ s_0 : \text{State} \end{array} \rangle$$

# Timed Runs and Simulation

- ✗ **SG\*** - sequences of transformations labelled by time stamps

$$G_0 \rightarrow_{p_1, t_1} G_1 \rightarrow_{p_2, t_2} \dots \rightarrow_{p_n, t_n} G_n$$

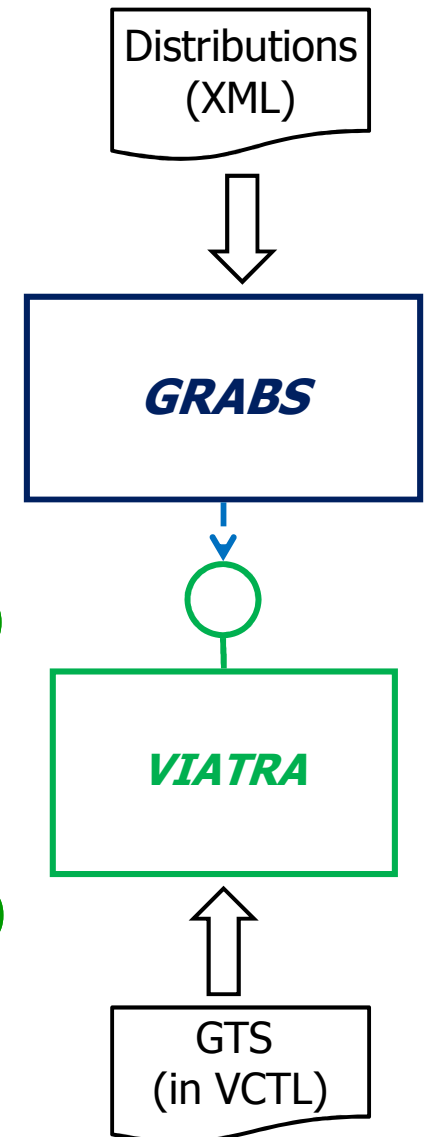
- ✗ Sampled by *stochastic simulation*

Initialisation

- ◆ compute all enabled events (rule, match)
- ◆ for each event determine randomly (based on distributions) their delay

Iteration

- ◆ select next event and apply (rule, match)
- ◆ update enabled events and remaining delays



## Properties of Runs

For  $r = (G_0 \xrightarrow{p_1, m_1, t_1} \dots \xrightarrow{p_n, m_n, t_n} G_n)$ ,  $ct(r) = t_n$  is *completion time*.

- $G_{i-1} \xrightarrow{p_i, m_i} G_i$  has application time  $at(i) = t_i$ , enabling time  $et(i)$
- If  $m_i$  exists at the start of the run,  $et(i) = 0$ .
- Otherwise, if enabled by  $G_{i-j} \xrightarrow{p_j, m_j} G_j$  with  $j < i$ ,  $et(i) = at(j)$
- $delay(i) = at(i) - et(i)$  is random variable with distribution  $F(p_i)$

Run  $r$  *follows* sequence  $s$  if both contain the same steps (rules and matches) in the same order.

Completion time distribution of (runs following) sequence  $s$  assigns conditional probability for a run to complete within time  $ctd(s)$  if the run follows  $s$ .

$$ctd(s)(x) = Prob\{ct(r) \leq x \mid \text{run } r \text{ follows } s\}$$

# Completion Time Distribution

**Proposition 1 (completion time distribution).** *Assume a sequence  $s = (G_0 \xrightarrow{p_1, m_1} \dots \xrightarrow{p_n, m_n} G_n)$  in  $\mathcal{SG}$ . The set of critical steps  $CS(s) \subseteq \underline{n} = \{1, \dots, n\}$  of  $s$  is the smallest subset of indices of steps such that*

- $n \in CS(s)$
- For all  $k \in CS(s)$ ,  $j \leq k$  with  $j \rightsquigarrow k$ , such that for all  $j' \leq k$  with  $j' \rightsquigarrow k$  implies  $j \geq j'$ , also  $j \in CS(s)$

*The completion time distribution of runs following  $s$  is given by*

$$ctd(s) = *_{i \in CS(s)} F(p_i)$$



# Lifting the Basic Theory of Graph Transformation

- ✗ Complex transformations via composed rules
  - Parallel transformation
  - Concurrent transformations
- ✗ General pattern
  - Define composition operation  $\#$
  - Define relation of sequences  $(\mathbf{G}\#)^*$  using composed rules to basic sequences  $\mathbf{G}^*$
- ✗ Stochastic GTS
  - Lift definition of  $\text{ctd}$  from  $\mathbf{G}^*$  to  $(\mathbf{G}\#)^*$
  - Fine delay distribution  $\mathbf{F}(\mathbf{c})$  for composed rules  $\mathbf{c}$  such that  $\mathbf{F}(\mathbf{c}) = \text{ctd}(\mathbf{G} \rightarrow_{\mathbf{c}} \mathbf{H})$

## Series-parallel Productions

- ✘ Using  $+$  for disjoint and  $;$  for dependent concurrent productions, series-parallel rule expressions are

$$c ::= p \mid c_{i \in e_1, e_2} c \mid c + c$$

- ✘ Assignments  $\pi$  of productions and  $\mathbf{F}$  of distribution functions are extended inductively
- ✘ Exact s-p productions, recursively
  - in  $\mathbf{c}; \mathbf{d}$  all components in  $\mathbf{d}$  depend on all components in  $\mathbf{c}$
  - in  $\mathbf{c} + \mathbf{d}$ , both  $\mathbf{c}$  and  $\mathbf{d}$  are s-p

## Series-parallel Transformations

- ✗ If  $\mathbf{d}$  is a step using an exact s-p production  $\mathbf{c}$ , then  $\mathbf{ctd}(\mathbf{d}) = \mathbf{F}(\mathbf{c})$
- ✗ If  $\mathbf{c}$  is not exact, then  $\mathbf{F}(\mathbf{c})$  is an upper bound, i.e., for each deadline  $\mathbf{t}$ ,  $\mathbf{F}(\mathbf{c})$  is less likely to meet it than  $\mathbf{ctd}(\mathbf{t})$

$$\mathbf{F}(\mathbf{c})(\mathbf{t}) \leq \mathbf{ctd}(\mathbf{d})(\mathbf{t})$$

**Proof:** Induction on the structure of rule expressions, using parallelism and concurrency theorems as induction steps.

# Conclusion

- ✓ Basic theory of algebraic GTS lifted
- ✓ Symbolic computation of completion time distribution for series-parallel processes
  
- ✗ Connections to be explored
  - Scheduling theory: *makespan* of a partially ordered network of tasks with stochastic durations
  - Concurrent semantics: deterministic graph processes, non-deterministic unfoldings
  
- ✗ Potential applications
  - Local optimisation of processes
  - Refinement of stochastic system