

# On Probabilistic Parallel Programs with Process Creation and Synchronisation

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# Split-Join Systems: A Model for Programs with Process Spawning

An example run:

The rules of a SJS:

$X$

split:  $X \hookrightarrow \langle XX \rangle$

intern:  $X \hookrightarrow q$

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join:  $\langle qr \rangle \hookrightarrow X$

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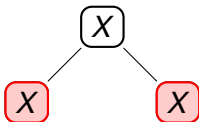
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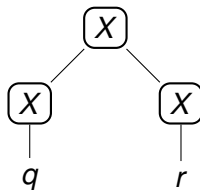
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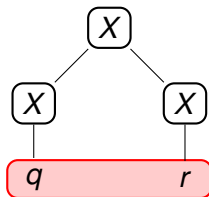
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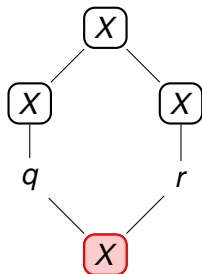
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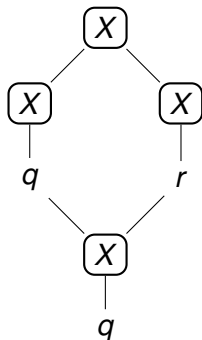
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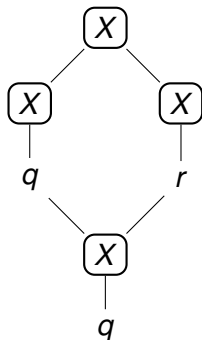
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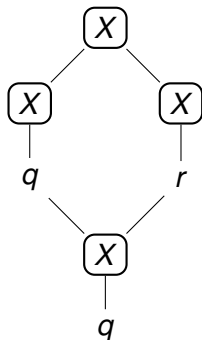
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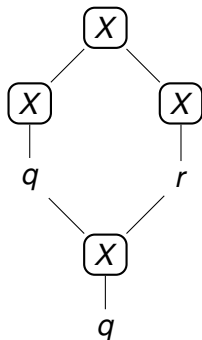
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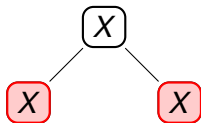
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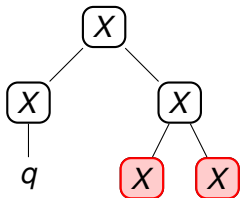
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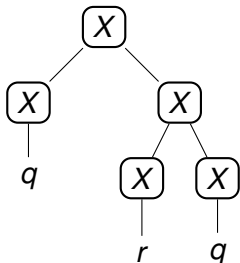
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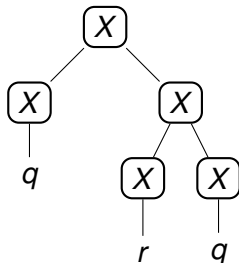
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Time  $T = 3$ , Work  $W = 5$ , Space  $S = 3$

# Why the Sibling Constraint?

The **sibling constraint** prevents arbitrary synchronisation.

- Arbitrary Synchronisation would lead to Petri-Nets.  
→ more difficult to analyse
- Sibling Synchronisation is enough for modeling purposes.



pSJSs subsume (probabilistic) pushdown systems (pPDSs).

Pushdown rules:

push:  $qX \hookrightarrow rYZ$

intern:  $rY \hookrightarrow sW$

pop:  $sW \hookrightarrow t$

pop:  $tZ \hookrightarrow u$

Run  $qX \Rightarrow rYZ \Rightarrow sWZ \Rightarrow tZ \Rightarrow u$

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View this run as:

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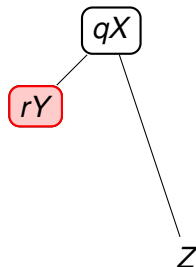
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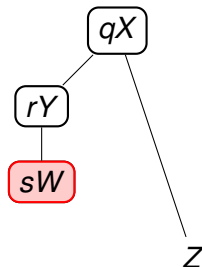
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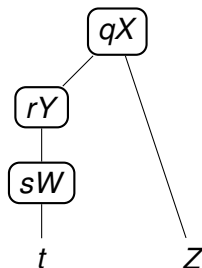
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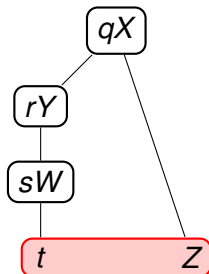
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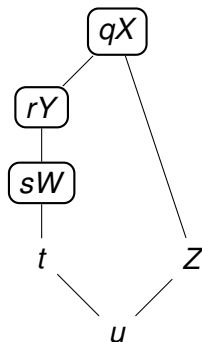
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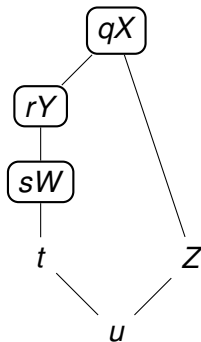
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Conversely, any pSJS can be **sequentialised**.

The resulting pPDS is equivalent with respect to

- returned value
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Bottomline:

- pPDSs: **sequential** programs
- pSJSs: **parallel** programs

# Relationship to Branching Processes

A **branching process** is a pSJS without join rules.

## Example

$$X \xrightarrow{0.5} \langle XY \rangle$$

$$Y \xrightarrow{0.3} X$$

$$X \xrightarrow{0.5} q$$

$$Y \xrightarrow{0.7} q$$

A single synchronisation state suffices.

classical mathematical model

→ biology, physics, natural language processing, ...

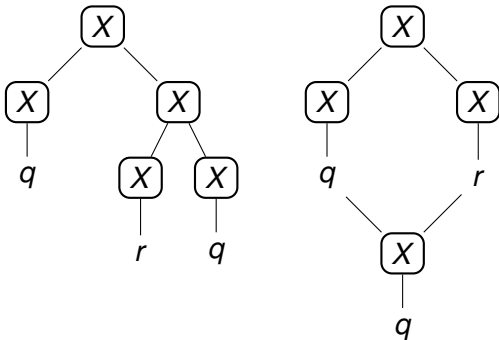
Our pSJS model **generalises** pPDSs and branching process.

- **procedures** from pPDSs
  - may be recursive
  - may return values
- parallel **spawns** from branching processes
- new: **joins** for process synchronisation and communication

# Termination

terminating runs:

split:  $X \xrightarrow{0.5} \langle XX \rangle$   
intern:  $X \xrightarrow{0.3} q$   
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One can transform the pSJS so that  
all terminating runs terminate in a single state.

# Termination Probability

$[X \downarrow q]$  = Prob. that  $X$  terminates in  $q$

## Example

$$X \xrightarrow{2/3} \langle XX \rangle \quad X \xrightarrow{1/3} q \quad \langle qq \rangle \xrightarrow{1} q$$

$X$  can only terminate in  $q$ . We have  $[X \downarrow q] = 1/2$ .

## Theorem

*The  $[X \downarrow q]$  are the solution of a system of polynomial equations.*

*Deciding whether  $[X \downarrow q] = 0$  is in  $P$ .*

*Deciding whether  $[X \downarrow q] < 1$  is PosSLP-hard even for pPDSs.*

# Probability of Finite Space

A pSJS may be useful even if it does not always terminate.  
→ operating systems, network servers, system daemons, ...

## Theorem

Let  $r$  be the *probability* that  
a computation started in  $X$  needs only *finite space*.  
Then  $r$  can be “efficiently expressed”.  
Deciding whether  $r = 0$  is in P.  
Deciding whether  $r < 1$  is PosSLP-hard even for pPDSs.

Proof: By transforming the pSJS so that  
a run terminates if and only if it needs finite space.

Applied to pPDSs, the theorem improves on [EKM05].

# Work and Time

Given a pSJS and a start process  $X$ , one can compute  $\mathcal{P}(W = k)$  and  $\mathcal{P}(T = k)$  for  $k = 1, 2, \dots$  iteratively.

This allows to **approximate**  $\mathbb{E}W$  and  $\mathbb{E}T$ .



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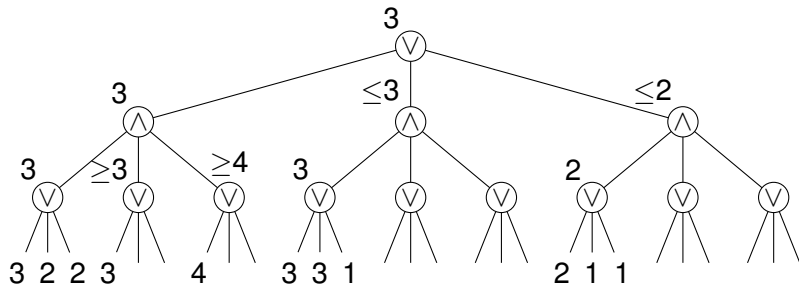
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Proof sketch for the upper bound:

- transform the pSJS to a **branching process** with similar distribution of  $W$  and  $T$  (uses  $[X \downarrow q]$ )
- set up a **matrix**  $A$  for the branching process such that  $A_{X,Y}$  = expected number of spawned  $Y$ -processes when applying an  $X$ -rule
- compute the **spectral radius** of  $A$  and compare with 1

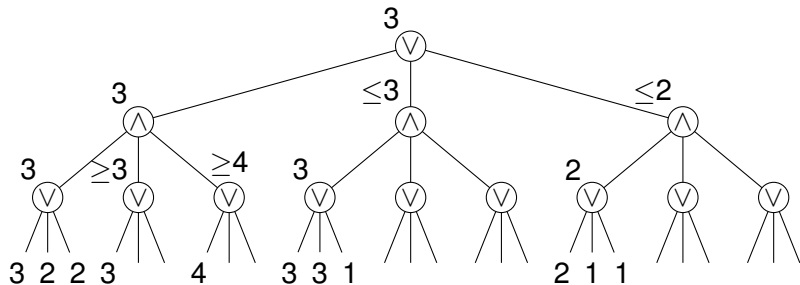
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We wish to analyse programs that evaluate min-max trees:



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(assume 0 or 3 children per node)

# A Straightforward Parallel Program

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function parMax(node)
  if node.leaf() then return node.val()
  else parallel < val1 := parMin(node.c1),
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  return max{val1, val2, val3}
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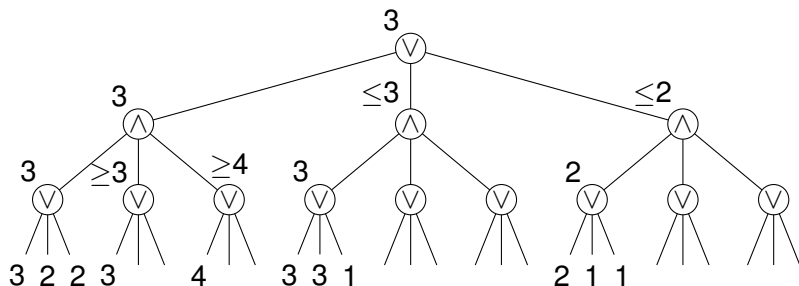
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→ parallel, but a lot of unnecessary work  
(maybe deep in the tree)

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# A Smart Sequential Program

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function seqMax(node,  $\alpha$ ,  $\beta$ )      (initially:  $\alpha = -\infty$ ,  $\beta = +\infty$ )  
  if node.leaf() then if node.val()  $\leq \alpha$  then return  $\alpha$   
    elseif node.val()  $\geq \beta$  then return  $\beta$   
    else return node.val()  
else val1 := seqMin(node.c1,  $\alpha$ ,  $\beta$ )  
  if val1 =  $\beta$  then return  $\beta$       (*cut-off after 1st child*)  
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→ no unnecessary work, but sequential

# A Smart Parallel Program

## “Young Brothers Wait”:

Idea: cut-offs usually occur after the 1st child (“oldest brother”)

- do the oldest brother
- only if no cut-off: do the young brothers **in parallel**

```
function YBWMax(node,  $\alpha$ ,  $\beta$ )  
  if node.leaf() then if node.val()  $\leq \alpha$  then return  $\alpha$   
    elseif node.val()  $\geq \beta$  then return  $\beta$   
    else return node.val()  
  else val1 := YBWMin(node.c1,  $\alpha$ ,  $\beta$ )  
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    else parallel  $\langle$  val2 := YBWMin(node.c2, val1,  $\beta$ ),  
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    return max{val2, val3}
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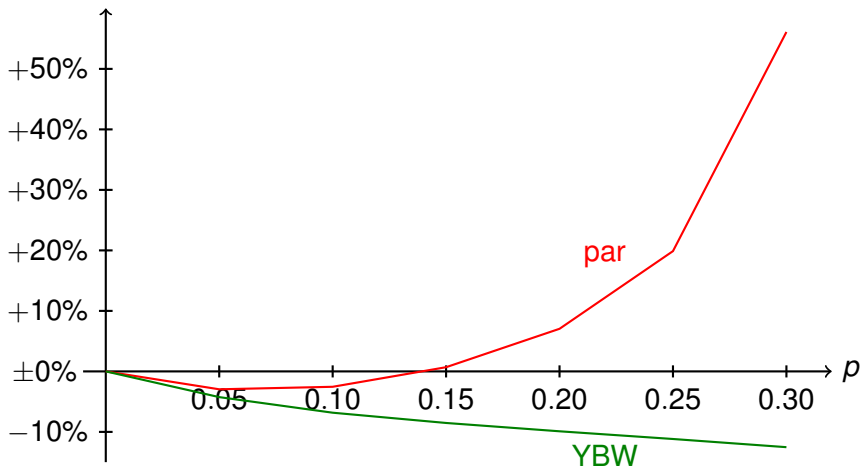
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This all fits naturally in the pSJS model.

We analyse performance under prob. assumptions on the input.

# Performance Analysis

Expected runtime of **par** and **YBW** compared to seq,  
as a function of  $p$  ( $p$  controls the expected tree size)



# Conclusions

- **pSJS: new model** for probabilistic parallel programs with **process spawning and synchronisation**
- Basic quantitative analysis is **as expensive as for pPDSs**.
- We can **model, analyse and compare** parallel programs under probabilistic assumptions.

Thank you!