

Spiking Neural P Systems with Neuron Division

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background

efficient solutions to hard problems
eg. SAT

- nondeterminism
- exponential size resources (*precomputed*)

versus

- deterministic
- polynomial time
- polynomial size system (*initially*)

GhP: ‘linear parallel time’

‘PPP’ previous work

L. Pan, G. Păun, M.J. Pérez-Jiménez.

Spiking neural P systems with neuron division and budding.

7th Brainstorming Week on Membrane Computing, vol. II,
pp. 151–168 (2009)

goal uniform solution to SAT

cf. active membranes (using spiking)

instance given as ‘input’

size of problem as parameter

precomputed: polynomial size system

here: same methodology as PPP–paper

but – do not use budding

and – constant size initial system (vs. linear)

spiking neural P system with neuron division

neurons	nodes ‘places’
synapses	directed edges
spikes	objects ‘tokens’

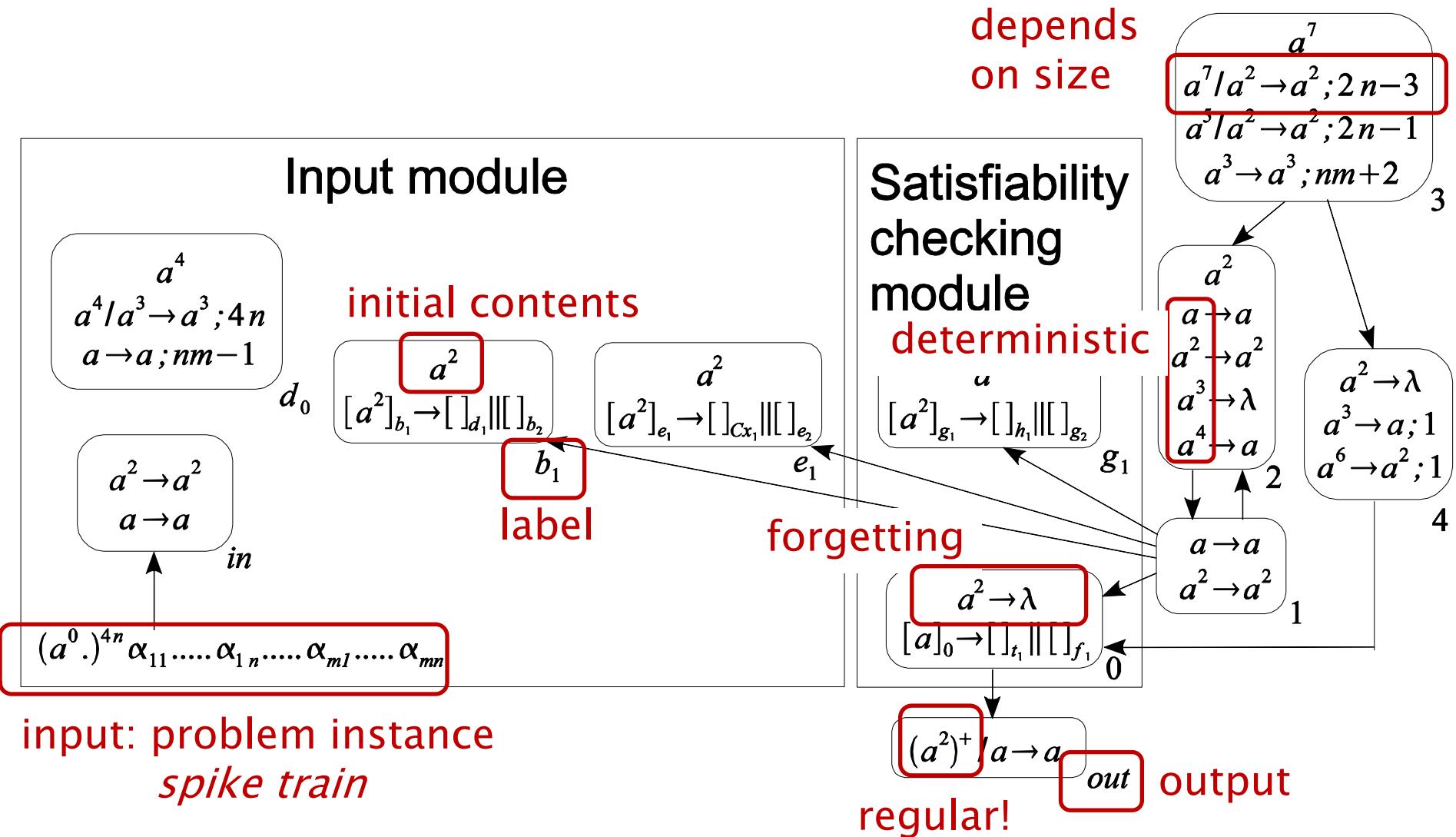
[$E / a^c \rightarrow a^p ; d$]

extended firing

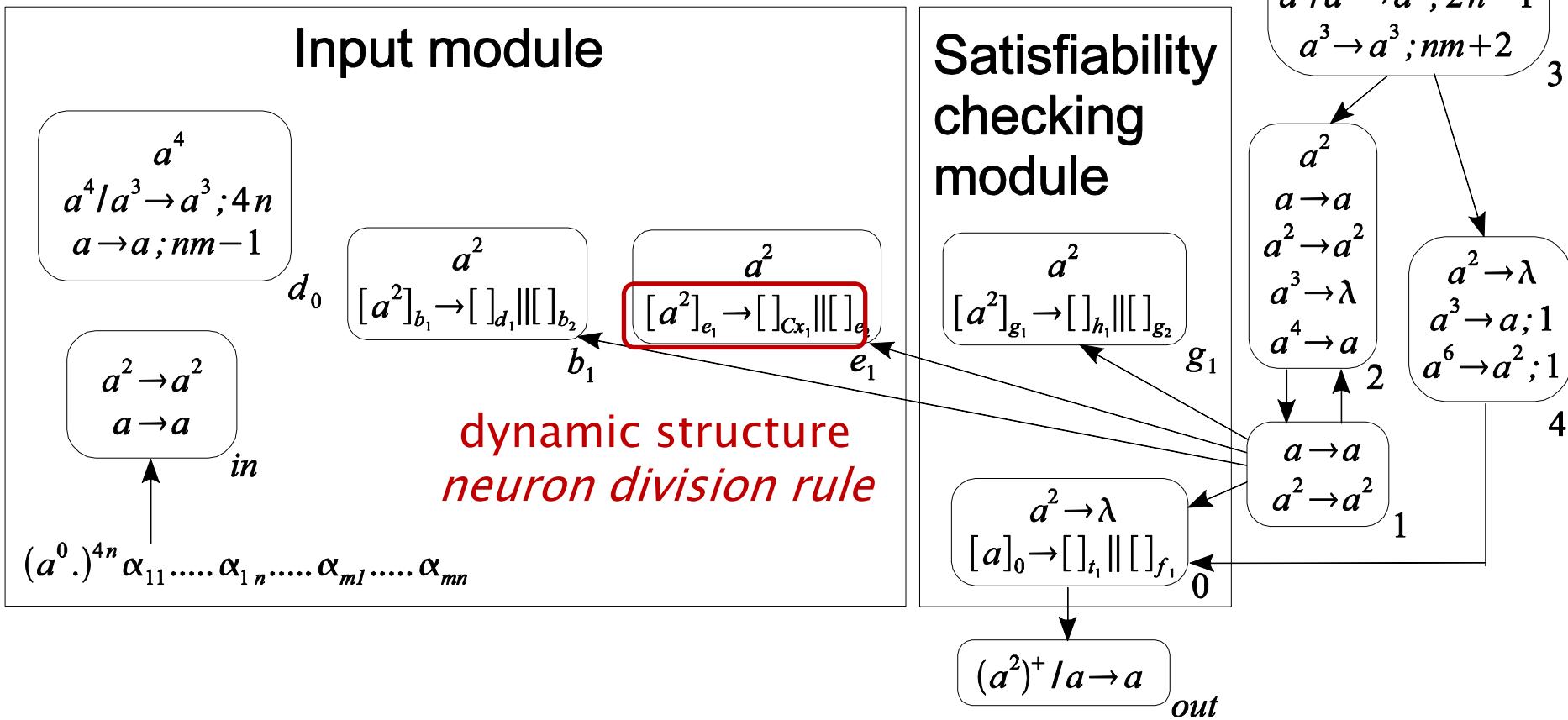
E ‘test’ regular expression over a
 $c \geq 1, p \geq 0, c \geq p$
consume, produce
 $d \geq 0$ *delay* ‘blocks’

parallelism – one rule for each neuron
(non)determinism

initial system



initial system



neuron division

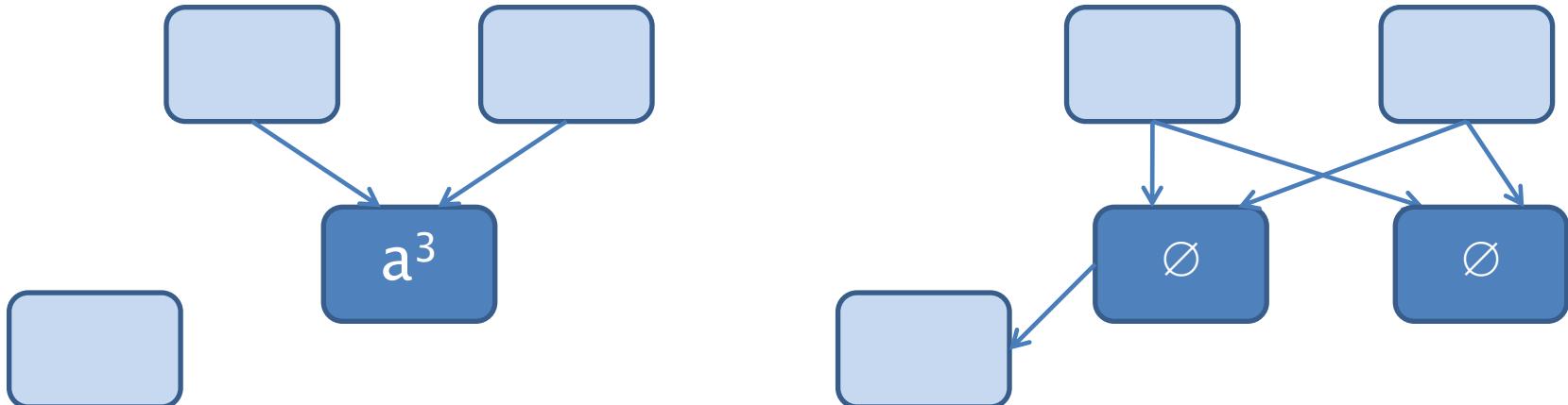
$$[E]_i \rightarrow []_j \parallel []_k$$

neuron division

E regular expression ‘test’

children inherit synapses
+ *synapse dictionary* (based on label)

no initial spikes



solving SAT

uniform encoding, parameters:
n variables and m clauses

SAT(n,m)

X = { x₁, x₂, … , x_n } variables

Y = C₁ ∧ C₂ ∧ … ∧ C_m proposition
conjunction ‘and’ of

C_j clause,
disjunction ‘or’ of
x_i or $\neg x_i$ literals

$$(x_1 \vee x_2 \vee x_3) \wedge (\neg x_2 \vee x_4) \wedge (\neg x_1 \vee x_3 \vee \neg x_4)$$

1	0	?	1	0	?	1
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satisfiability – assignment to variables

system strategy

closely follows Pan, Păun, Pérez-Jiménez

Generation Stage

generates structure of system
division

Input Stage

reads instance γ into system

Satisfiability Checking Stage

test all valuations on clauses

Output Stage

any valuation left?

input encoding

$X = \{ x_1, x_2, \dots, x_n \}$ variables

$\gamma = C_1 \wedge C_2 \wedge \dots \wedge C_m$ clauses

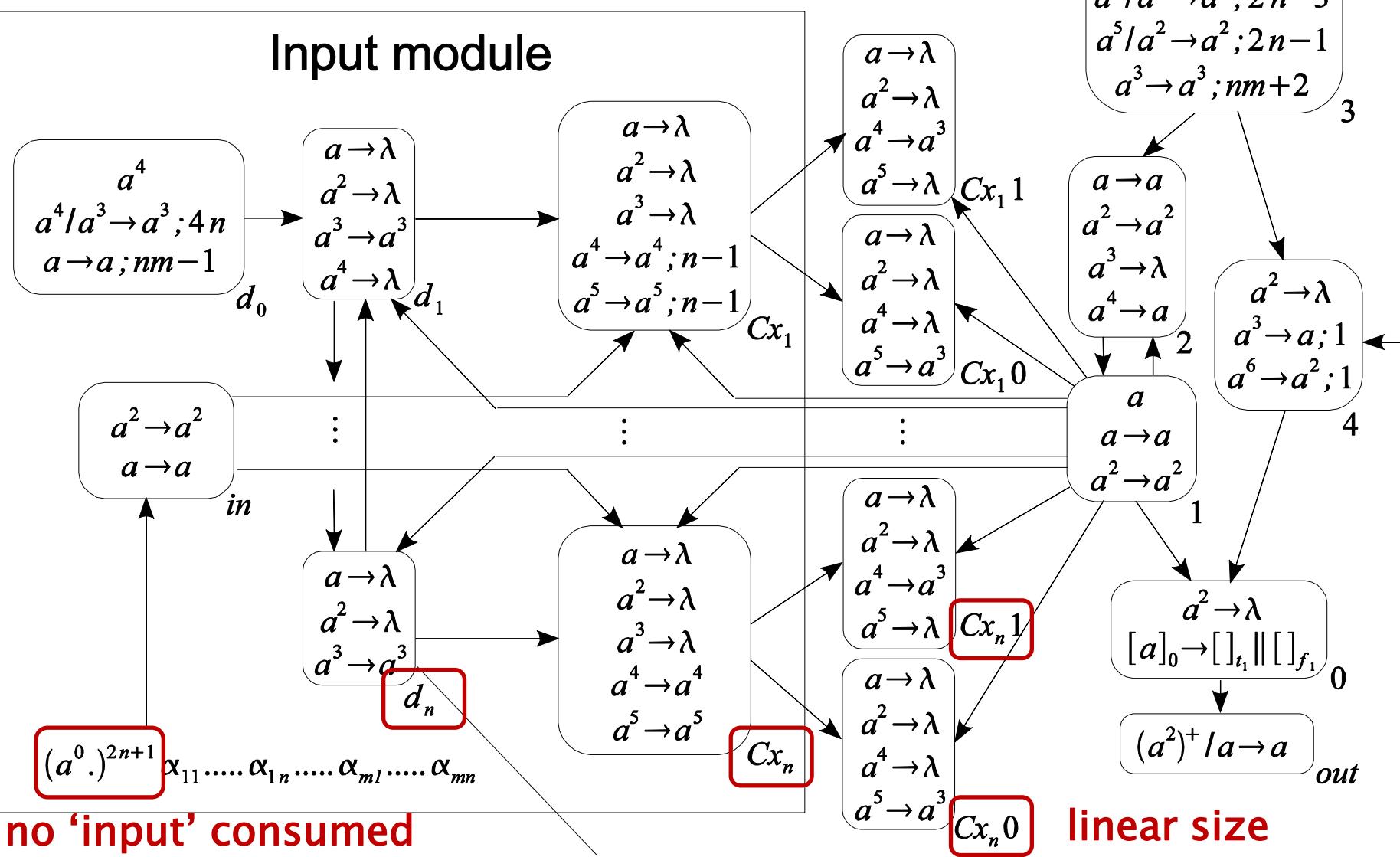
$$\alpha_{ij} = \begin{cases} a & x_j \text{ occurs in } C_i \\ a^2 & \neg x_j \text{ occurs in } C_i \\ a^0 & \text{otherwise} \end{cases}$$

spike train, sent into initial neuron

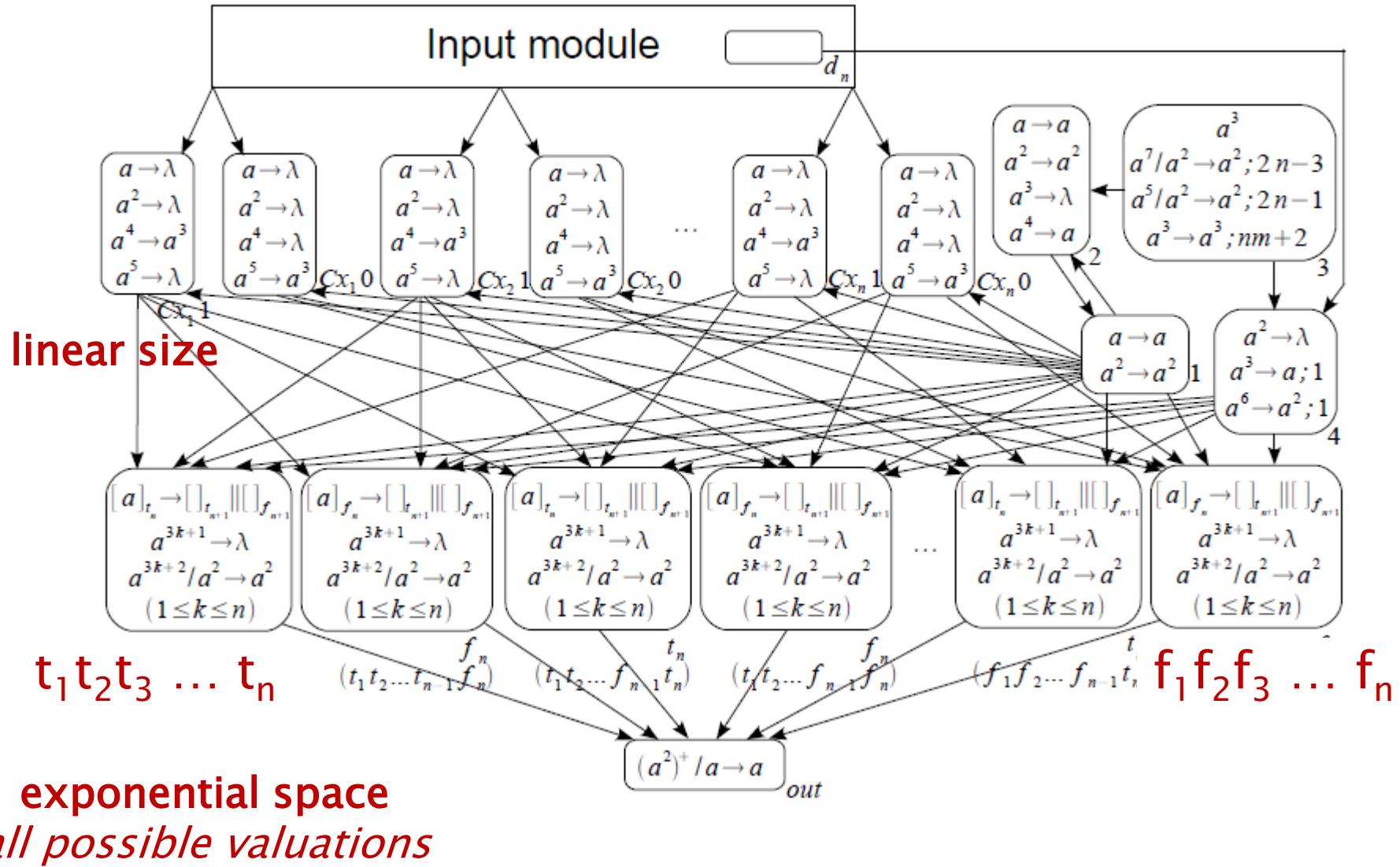
$$(a^0)^{4n} \alpha_{11} \dots \alpha_{1n} \dots \underbrace{\alpha_{i1} \alpha_{i2} \dots \alpha_{in}}_{\text{delay}} \dots \alpha_{i1} \dots \alpha_{in}$$

clause C_i

generation: step $2n-1$



generation: step 4n-1

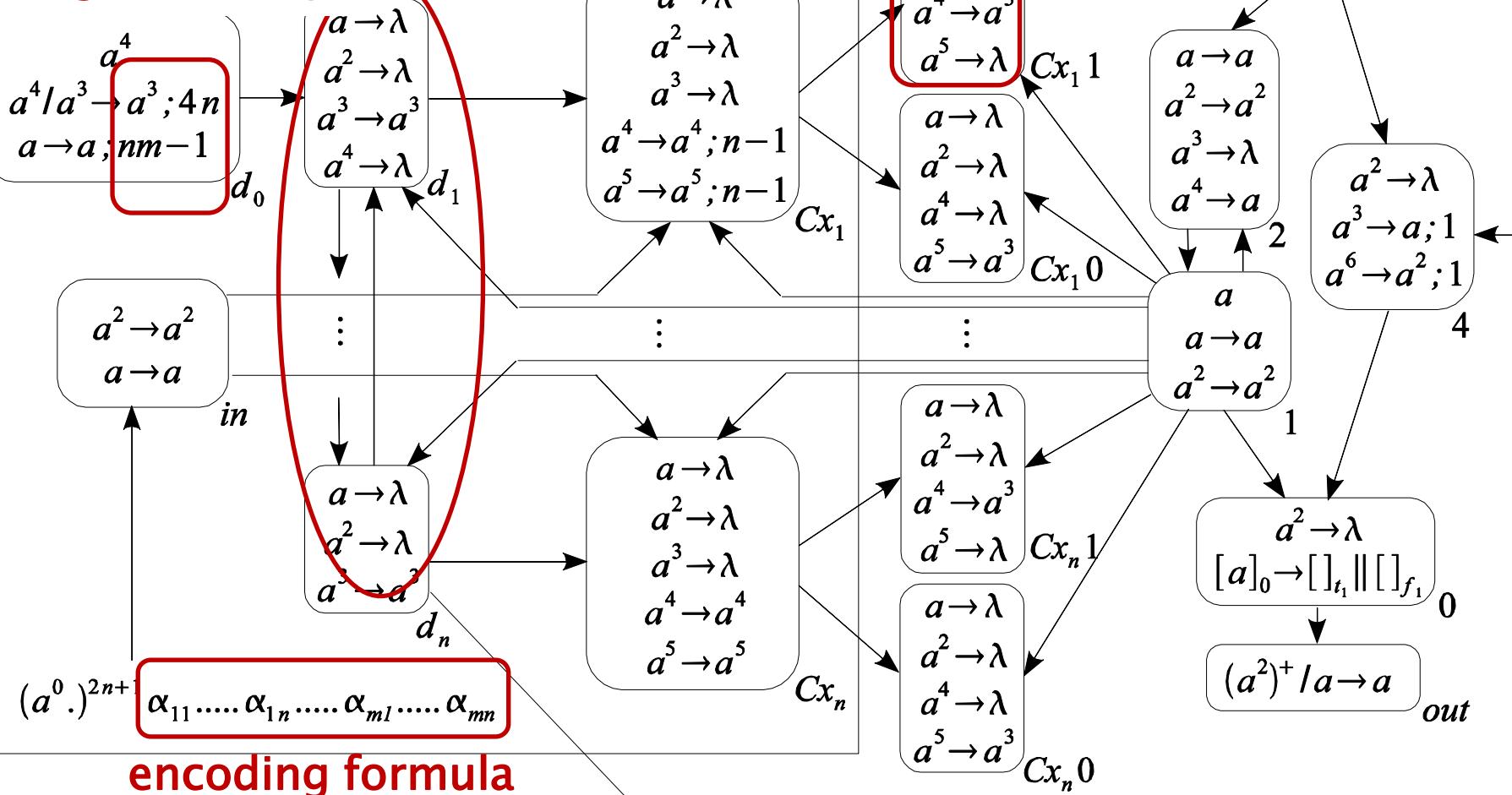


input. step $4n+1 \dots 4n+nm+1$

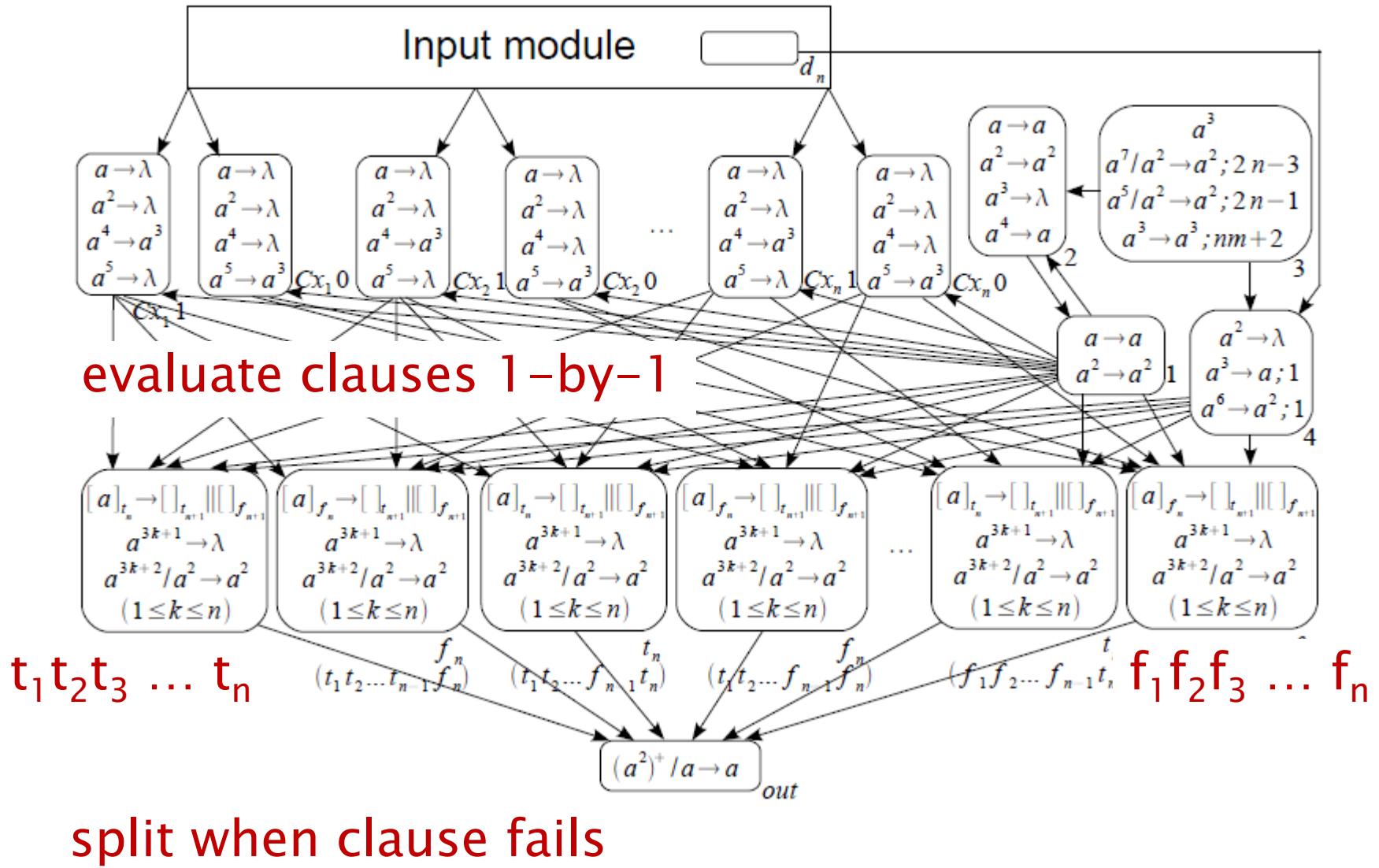
input module

a^3 loop – ‘timing’
for right variable

timing start/stop



satisfiability and output



comparing resources

Resources	this work	“PPP”
Initial number of neurons	11	$4n + 7$
Initial number of spikes	20	9
Number of neuron labels	$10n + 7$	$6n + 8$
Size of synapse dictionary	$6n + 11$	$7n + 6$
Number of rules	$2n^2 + 26n + 26$	$n^2 + 14n + 12$

too much names/labels !
can we reuse & get right structure ?

conclusion

perhaps improved ‘timing’ can eliminate labels
for consecutive ‘linear’ steps

division and/or budding

budding → linear size growth so ...

initial system (PPP)

Input module

