

What is a cell cycle checkpoint ?

The TotemBioNet answer

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Abstract. TotemBioNet is a new software platform to assist the design of qualitative regulatory network models by combining “genetically modified Hoare logic”, temporal logic model checking and optimized enumeration techniques. TotemBioNet is particularly efficient to manage parameter identification, the most critical step of formal modelling. It is also remarkably flexible and efficient to check properties in order to explore biological assumptions. To illustrate this efficacy, we address the classical example of the cell cycle, where the passage from one phase to the next one, called *checkpoint*, is crucial but is usually a rather fuzzy informal concept. The cyclic behaviour of the cell cycle is specified by temporal logic and the order of individual events inside each phase is explored thanks to quantifiers introduced in Hoare logic. This way, TotemBioNet rapidly suggests a sensible formalization of the notion of checkpoint.

Keywords: Regulatory network · discrete modelling · parameter identification · Hoare logic · temporal logic · model-checking · cell cycle.

1 Formal Methods for Thomas Regulatory Networks

In the 70's, *qualitative models* based on discrete mathematics [10,17] have proved useful to understand the main causalities that govern observed phenotypes [19,18], and the multivalued framework of René Thomas and Houssine Snoussi has become a classic for biological regulatory networks. It gained new power with the introduction of formal methods in the early 2000s [4], concomitantly with [5] for signalling networks.

A qualitative model is an influence graph where the important actors for the biological question, as well as their interactions, have been inventoried on the basis of biological knowledge. Formally, the graph covers a huge set of different discrete models because the strengths of combined activations and inhibitions are unknown. They are encoded by a set of discrete *parameters* [15], which we need to identify: Any biological knowledge reduces the number of possible parameter values, by rejecting the parameterizations that do not comply. Parameter identification is the most difficult part of the modelling activity.

Simulations rapidly reach their limit, because of the non-determinism of trajectories and of a high number of parameterizations: A finite number of simulations cannot establish general properties. Formal methods solve this difficulty.

There are software platforms based on Thomas’ semantics: some study the *invariants* of a given model using Petri net tools [12] and some others perform *model checking* algorithms inspired from program verification techniques [1,11]. *Temporal logics* allow to check very general properties, including those universally quantified on traces, and SMBioNet [4], based on CTL, has made possible to *exhaustively* treat *sets of models* (sets of parameterizations) by optimized enumeration algorithms. Nevertheless, to find the *exhaustive* set of possible parameterizations, enumeration asks for exponentially growing computation time w.r.t. the size and connectivity of the regulatory network.

On another note, biological experiments directly request a set of traces in the model, and the “genetically modified Hoare logic” [3] is much more effective than temporal logic for this task. Instead of commonly enumerating parameterizations as in many approaches [14], [16] and [9], it produces a set of constraints on the parameters that characterizes those in which these traces exist. The program Hoare-*fol* [13] efficiently computes these constraints.

TotemBioNet stands out by combining all these approaches in such a way that modellers converge *rapidly* toward the exhaustive set of parameterizations satisfying all *formalized* biological knowledge.

Finding the set of parameterizations compatible with current knowledge does not end the modelling activity. It remains to use the model to study the *biological question*... which, most of the time, is *not yet formalized*. Each possible explanation constitutes an hypothesis, and formalizing the latter is necessary to, at least, check its *consistency*: if the set of compatible parameterizations is empty then the hypothesis is inconsistent. **TotemBioNet**’s ability to repeatedly question the model about its diverse properties, and to obtain quick answers, allows for a fast convergence towards a formalisation of the biological question.

2 The platform TotemBioNet

TotemBioNet supports two variants of temporal logics: CTL and a dedicated *fair-path* CTL, needed for certain reachability properties in the Thomas framework. Indeed, the quantifiers *A* and *E* in CTL often induce artifactual results because they consider *unfair* paths that cross an infinite number of times a given state but never fire a possible transition from this state. So, universally quantified CTL formulas to study, for instance, attraction basins become unfairly false. Fair-path CTL quantifiers *A* and *E* simply ignore unfair paths [13]. **TotemBioNet** automatically translates fair-path CTL formulas into (more complex) CTL formulas, allowing it to benefit from usual CTL model checking algorithms.

Besides, **TotemBioNet** Hoare triples contain: a pre-condition describing the possible initial states of a given biological experiment, a path, and an observed post-condition. The path abstracts the curves obtained from experimental observations: according to thresholds setting, a threshold crossing of a variable *v* along the curve is written *v+* if *v* increases, or *v-* if *v* decreases. When experimental conditions are not precise enough to know which variable passes its threshold first, *existential quantifiers* can express this uncertainty: $\exists(v_1+; v_2- , v_2- , v_1+)$

means that v_1 has increased and v_2 has decreased, but in an unknown order. Also, universal quantifiers permit to abstract together a collection of similar experimental observations. Genetically modified Hoare logic extends classical Hoare logic by formalising under which conditions on the parameters each $v+$ or $v-$ of the path can occur. Then, the usual *weakest pre-condition* is the constraint on the parameters that makes the abstract path possible [3].

The inputs of TotemBioNet are: an influence graph, any knowledge on the parameter values, and properties on the dynamics of the system expressed using CTL, fair-path CTL, or Hoare triples. TotemBioNet integrates *Hoare-fol* [8] and an extended version of *SMBioNet* [4]. First, *Hoare-fol* computes and simplifies the weakest pre-condition wp w.r.t. genetically modified Hoare logic [3]. Then, the enumeration process of TotemBioNet is based on that of *SMBioNet*, which exploits self-influences and the *Snoussi constraint* (more resources cannot reduce the expression level) to greatly reduce the enumeration complexity.

TotemBioNet enumerates all parameterizations that satisfy wp : *i*) if $wp \equiv False$, the enumeration process stops, *ii*) if wp is a conjunction of atoms of the form $(K_{v_i} \leq s_i)$ or $\neg(K_{v_i} \leq s_i)$ where K_{v_i} is a parameter and s_i a threshold for variable v_i , then the enumeration domains of K_{v_i} are reduced, and *iii*) if wp contains disjunctions, the validity of wp is checked on the fly. This considerably reduces the search space of all possible parameterizations and TotemBioNet generates, for each remaining parameterization, one input file for the model checker *NuSMV* [6]. This file contains the conjunction of temporal formulas and an automaton which encodes the state transition graph for the current parameterization. TotemBioNet also offers *environment variables* used to freeze some variables according to an experimental environment (by the way, it also reduces the number of parameters).

TotemBioNet, see <https://gitlab.com/totembionet/totembionet>, comes with many examples that illustrate the combination of CTL properties, fair-path CTL properties and Hoare triples. TotemBioNet allows one to describe the influence graph with *yEd* graph editor (<https://www.yworks.com/products/yed>). A typical session consists in building the influence graph using *yEd*, in automatically generating the corresponding text file and then in adding temporal properties and Hoare triples in concrete syntax. TotemBioNet generates an output file (possibly in *csv* format) which contains all parameterizations, labeled with “OK” when the dynamic properties are verified, and if not with all the properties which are not satisfied. The global TotemBioNet process is illustrated in Figure 1.

3 TotemBioNet Use Case: a Simplified Cell Cycle Model

The cell cycle is a series of events leading to correct duplication of DNA of a cell (synthesis or S phase) and its division into two genetically identical daughter cells (mitosis or M phase). Gap phases G1 and G2 lie respectively before S and M. Progression through the cell cycle is driven by Cyclins/Cyclin-dependant kinases complexes (Cyc/Cdks) and their inhibitors. A 5-variables cell-cycle model has been designed in [2] where the variables sk , a and b are the main Cyc/Cdks

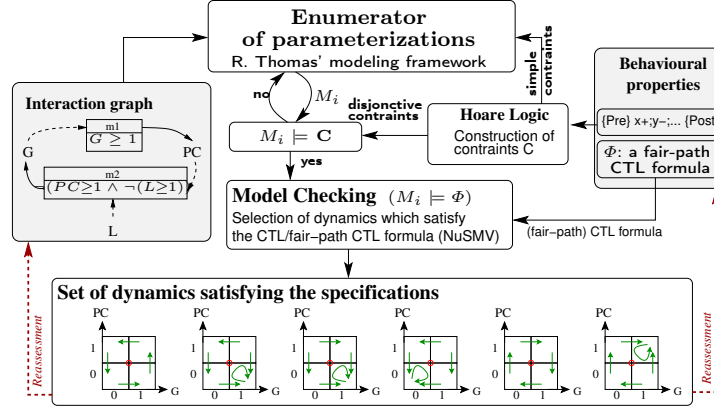


Fig. 1: TotemBioNet' processing flow

involved in the mammalian cell cycle and en and ep their inhibitors. The interaction graph and its variables are detailed in Appendix A. Moreover, the succession of phases, G1, S, G2 and M, has been described in [2] *via* the Hoare

$$\text{triple } \mathbf{H}_{init} : \left\{ G1_{init} \right\} \begin{array}{l} sk+; sk+; en-; \\ a+; sk-; sk-; b+; \\ a-; ep+; \\ en+; b-; ep-; \end{array} \left\{ G1_{init} \right\} \text{ where the pre- and post-condition}$$

$G1_{init}$ specifies the state $sk = 0, ep = 0, a = 0, b = 0, en = 1$, G1 is the blue subsequence, S the red one, G2 the dark gray one and M the green one.

Our main question is: *Is this small model powerful enough to represent checkpoints?* First, notice that the model assumes constant infusion of growth factors and consequently its dynamics must *always* be cyclic. The ability of TotemBioNet to mix several formal approaches allows us to combine H_{init} and this property specified using fair-path CTL: $\varphi_{cyclic} \equiv G1_{init} \Rightarrow AX(AF(G1_{init}))$ (where $AXAF$ means a strict future). TotemBioNet results are synthetized in the first line of Table 1: from the 100800 parameterizations satisfying Snoussi constraint, 676 of them satisfy the weakest precondition calculated from H_{init} . Then 609 out of the 676 also validate φ_{cyclic} . Notice the great efficiency of Hoare Logic: The use of the equivalent CTL formula φ_{init} (see Appendix B) instead of the H_{init} Hoare triple in the first experiment of the table would drastically increase the computation time: from 6.1s to 18.5min for the same result.

Another question to understand checkpoint is the order of transitions inside phases. [2] suggests that some transitions inside a phase may admit permutations: all transitions except the first one for G1 and S, all transitions for G2 and none for M. H_{forall} encodes the 12 possible paths owing to the *Forall* quantifier,

$$\mathbf{H}_{forall} : \left\{ G1_{init} \right\} \begin{array}{l} sk+; \text{ Forall}((sk+; en-), (en-; sk+)); \\ a+; \text{ Forall}((sk-; sk-; b+), (sk-; b+; sk-), (b+; sk-; sk-)); \\ \text{ Forall}((ep+; a-), (a-; ep+)); \\ en+; b-; ep-; \end{array} \left\{ G1_{init} \right\}$$

and surprisingly TotemBioNet returns the same 609 parameterizations!

Exp	Hoare triple	$ H_m $	Temporal logic formula	$ S_m $	Computation Time (s)
1	H_{init}	676	φ_{cyclic}	609	6.1
2	H_{forall}	676	φ_{cyclic}	609	6.1
3	H_{perm}	0	φ_{cyclic}	0	0.24
4	H_{permG1}	260	φ_{cyclic}	240	2.4
5	H_{permG1}	260	$\varphi_{cyclic} \wedge \varphi_{G2/M} \wedge \varphi_{G1/S}$	28	2.9

Table 1: Formal properties of a simplified 5-variables cell cycle model: H_m is the set of models satisfying Hoare and Snoussi constraints. S_m is the set of selected models after model-checking of a temporal logic formula on each element of H_m . (Performed on an Intel Core i7-8650U processor, 1.90GHz, 8 cores.)

This suggests that a *phase* could be simply a bag of transitions that can be performed in arbitrary order. We check this idea with the Hoare triple H_{perm} (Appendix C) and **TotemBioNet** returns a unsatisfiable weakest precondition. So, let us check individually for G1, S and M. H_{permG1} asks for all permutations in G1, whereas S, G2 and M are the same as in H_{forall} (Appendix D): **TotemBioNet** returns 240 parameterizations. For S and for M no parameterization is selected. We conclude that the order of transitions suggested in [2] is constrained within S and M but not within G1 and G2.

Now, having a better idea of what goes on within a phase, it appears that a *checkpoint* between two phases, p_1 and p_2 should ensure that none of the possible *first* transitions of p_2 can be performed before one of the transitions of p_1 . The most biologically important and the most studied checkpoints are G2/M and G1/S. They can be formalized using Hoare logic but the CTL formula is simpler if we remark that the first state of a phase is unique, whatever the order of the

$$\text{previous phases: } \varphi_{G2/M} \equiv \begin{pmatrix} sk = 0 \\ \wedge ep = 0 \\ \wedge a = 1 \\ \wedge b = 1 \\ \wedge en = 0 \end{pmatrix} \Rightarrow \neg \begin{pmatrix} EX(\mathbf{en} = 1 \wedge EX(a = 0 \wedge EX(ep = 1))) \\ \vee EX(\mathbf{en} = 1 \wedge EX(ep = 1 \wedge EX(a = 0))) \\ \vee EX(a = 0 \wedge EX(\mathbf{en} = 1 \wedge EX(ep = 1))) \\ \vee EX(ep = 1 \wedge EX(\mathbf{en} = 1 \wedge EX(a = 0))) \end{pmatrix}.$$

Similarly $\varphi_{G1/S}$ is given in Appendix E, and **TotemBioNet** returns 28 parameterizations satisfying $\varphi_{G2/M}$ and $\varphi_{G1/S}$ in addition to H_{permG1} and φ_{cyclic} . Thus, *checkpoints can be captured in a purely discrete framework*. The biologically less studied checkpoints, S/G2 and M/G1, have been also formalized (Appendix F). No parameterisation is selected suggesting that the current model is not detailed enough to satisfy S/G2 or M/G1 checkpoint, as defined.

4 Conclusion

TotemBioNet combines in an optimized manner Hoare logic and (different variants of) temporal logic. Two of our current works are to facilitate an incremen-

tal analysis of models, as well as to provide a more user-friendly interface with jupyter notebook as BioCHAM [7] and CoLoMoTo [11] do. *TotemBioNet* aims at offering a growing palette of formal methods to the modellers, so that each biological knowledge can be formalized according to the most suited one. Thanks to the versatility and efficacy of *TotemBioNet*, the general properties of qualitative Thomas models can be rapidly checked during their design. We showed on the small cell cycle model initially specified with a Hoare triple in [2] how the main properties of the phases can be explored, leading to a proper formalization of the notions of phase and checkpoint.

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Appendix A: Static description of the cell cycle model

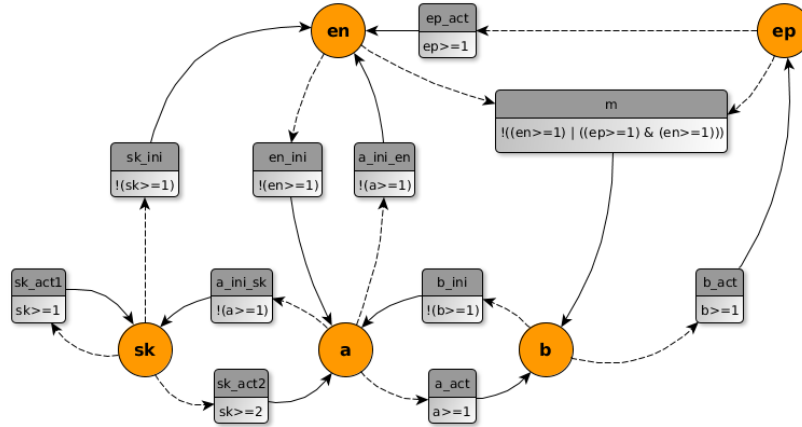


Fig. 2: A 5-variable interaction graph of the mammalian cell cycle, from [2]. Progression through the cell cycle is driven by 2 types of genetic entities: complexes of Cyclins/Cyclin-dependant kinases (Cyc/Cdks) and their inhibitors known as *enemies*. The 5 variables of the graph represent these entities, in orange. *sk* is the abstraction of both complexes CycE/Cdk2 and CycH/Cdk7, known as *starting kinases*. *a* and *b* respectively represent CycA/Cdk1 and CycB/Cdk1. *en* is the abstraction of the main Cyc/Cdks enemies: the anaphase-promoting complex APC/Cdh1, cyclin-kinase inhibitors *p21* and *p27*, and Wee1 protein. The variable *ep* is the anaphase-promoting complex APC/Cdc20, which is a Cyc/Cdks enemy involved in mitosis exit and so-called *exit protein*. Regulations between variables are described in [2]. This interaction graph was designed using the tool *yEd* (www.yworks.com/products/yed).

Appendix B: Equivalent specification of H_{init} using a Fair CTL formula

In the first experiment, the cell cycle is specified by the H_{init} Hoare triple. Here, the cell cycle is specified by the φ_{init} CTL formula depicting the H_{init} path.

$$\varphi_{init} \equiv \left(\begin{array}{l} (sk = 0 \wedge ep = 0 \wedge a = 0 \wedge b = 0 \wedge en = 1) \rightarrow EX(\mathbf{sk} = \mathbf{1} \wedge ep = 0 \wedge a = 0 \wedge b = 0 \wedge en = 1) \\ \wedge (EX(\mathbf{sk} = \mathbf{2} \wedge ep = 0 \wedge a = 0 \wedge b = 0 \wedge en = 1) \wedge (EX(sk = 2 \wedge ep = 0 \wedge a = 0 \wedge b = 0 \wedge en = 0) \\ \wedge (EX(sk = 2 \wedge ep = 0 \wedge a = \mathbf{1} \wedge b = 0 \wedge en = 0) \wedge (EX(\mathbf{sk} = \mathbf{1} \wedge ep = 0 \wedge a = 1 \wedge b = 0 \wedge en = 0) \\ \wedge (EX(\mathbf{sk} = \mathbf{0} \wedge ep = 0 \wedge a = 1 \wedge b = 0 \wedge en = 0) \wedge (EX(sk = 0 \wedge ep = 0 \wedge a = 1 \wedge \mathbf{b} = \mathbf{1} \wedge en = 0) \\ \wedge (EX(sk = 0 \wedge ep = 0 \wedge a = \mathbf{0} \wedge b = 1 \wedge en = 0) \wedge (EX(sk = 0 \wedge \mathbf{ep} = \mathbf{1} \wedge a = 0 \wedge b = 1 \wedge en = 0) \\ \wedge (EX(sk = 0 \wedge ep = 1 \wedge a = 0 \wedge b = 1 \wedge \mathbf{en} = \mathbf{1}) \wedge (EX(sk = 0 \wedge ep = 1 \wedge a = 0 \wedge \mathbf{b} = \mathbf{0} \wedge en = 1) \\ \wedge (EX(sk = 0 \wedge \mathbf{ep} = \mathbf{0} \wedge a = 0 \wedge b = 0 \wedge en = 1)))))))))) \end{array} \right)$$

Appendix C: Specification of H_{perm}

This Hoare triple encodes a cell cycle in which phases are described by all permutations of their respective transitions. G1 is specified in blue, S in red, G2 in grey and M in green.

$$\mathbf{H}_{perm} : \left\{ G1_{init} \right\} \begin{array}{l} \text{Forall}((sk+; sk+; en-), (sk+; en-; sk+), (en-; sk+; sk+)); \\ \text{Forall}((a+; sk-; sk-; b+), (a+; sk-; b+; sk-), \\ (a+; b+; sk-; sk-), (sk-; a+; sk-; b+), \\ (sk-; a+; b+; sk-), (b+; a+; sk-; sk-), \\ (sk-; sk-; a+; b+), (sk-; b+; a+; sk-), \\ (b+; sk-; a+; sk-), (sk-; sk-; b+; a+), \\ (sk-; b+; sk-; a+), (b+; sk-; sk-; a+)) \\ \text{Forall}((ep+; a-), (a-; ep+)); \\ \text{Forall}((en+; b-; ep-), (en+; ep-; b-), \\ (ep-; b-; en+), (ep-; en+; b-), \\ (b-; en+; ep-), (b-; ep-; en+)); \end{array} \left\{ G1_{init} \right\}$$

Appendix D: Specification of H_{permG1}

This Hoare triple describes the cell cycle in which G1 in addition to G2 allows all permutations of its transitions.

$$\mathbf{H}_{permG1} : \left\{ G1_{init} \right\} \begin{array}{l} \text{Forall}((sk+; sk+; en-), \\ (sk+; en-; sk+), (en-; sk+; sk+)); \\ a+; \\ \text{Forall}((sk-; sk-; b+), (sk-; b+; sk-), \left\{ G1_{init} \right\} \\ (b+; sk-; sk-)); \\ \text{Forall}((ep+; a-), (a-; ep+)); \\ en+; b-; ep-; \end{array}$$

Similarly, the premise of $\varphi_{M/G1}$ formula (see below) encodes the first state of M. $sk+$ and $en-$ are the 2 possible first events of G1 according to H_{permG1} . Thus the 8 paths enabling these events to occur before completion of M events must not exist, starting from the state in premise. $\varphi_{M/G1}$ is then defined as:

$$\left(\begin{array}{l} sk=0 \\ \wedge ep=1 \\ \wedge a=0 \\ \wedge b=1 \\ \wedge en=0 \end{array} \right) \Rightarrow \neg \left(\begin{array}{l} EX(en=1 \wedge EX(b=0 \wedge EX(sk=1 \wedge EX(ep=0 \wedge EX(sk=2 \wedge EX(en=0)))))) \\ \vee EX(en=1 \wedge EX(sk=1 \wedge EX(b=0 \wedge EX(ep=0 \wedge EX(sk=2 \wedge EX(en=0)))))) \\ \vee EX(sk=1 \wedge EX(en=1 \wedge EX(b=0 \wedge EX(ep=0 \wedge EX(sk=2 \wedge EX(en=0)))))) \\ \vee EX(en=1 \wedge EX(b=0 \wedge EX(sk=1 \wedge EX(ep=0 \wedge EX(en=0 \wedge EX(sk=2)))))) \\ \vee EX(en=1 \wedge EX(sk=1 \wedge EX(b=0 \wedge EX(ep=0 \wedge EX(en=0 \wedge EX(sk=2)))))) \\ \vee EX(sk=1 \wedge EX(en=1 \wedge EX(b=0 \wedge EX(ep=0 \wedge EX(en=0 \wedge EX(sk=2)))))) \\ \vee EX(en=1 \wedge EX(b=0 \wedge EX(en=0 \wedge EX(ep=0 \wedge EX(sk=1 \wedge EX(sk=2)))))) \\ \vee EX(en=1 \wedge EX(en=0 \wedge EX(b=0 \wedge EX(ep=0 \wedge EX(sk=1 \wedge EX(sk=2)))))) \end{array} \right)$$

TotemBioNet extracts no model (Table 2) for each of these checkpoints, from which we conclude that the model is not precise enough to capture them.

Exp	Hoare triple	$ H_m $	Temporal logic formula	$ S_m $	Computation Time (s)
6	H_{permG1}	260	$\varphi_{cyclic} \wedge \varphi_{G2/M} \wedge \varphi_{G1/S} \wedge \varphi_{S/G2}$	0	3.5
7	H_{permG1}	260	$\varphi_{cyclic} \wedge \varphi_{G2/M} \wedge \varphi_{G1/S} \wedge \varphi_{M/G1}$	0	3.2

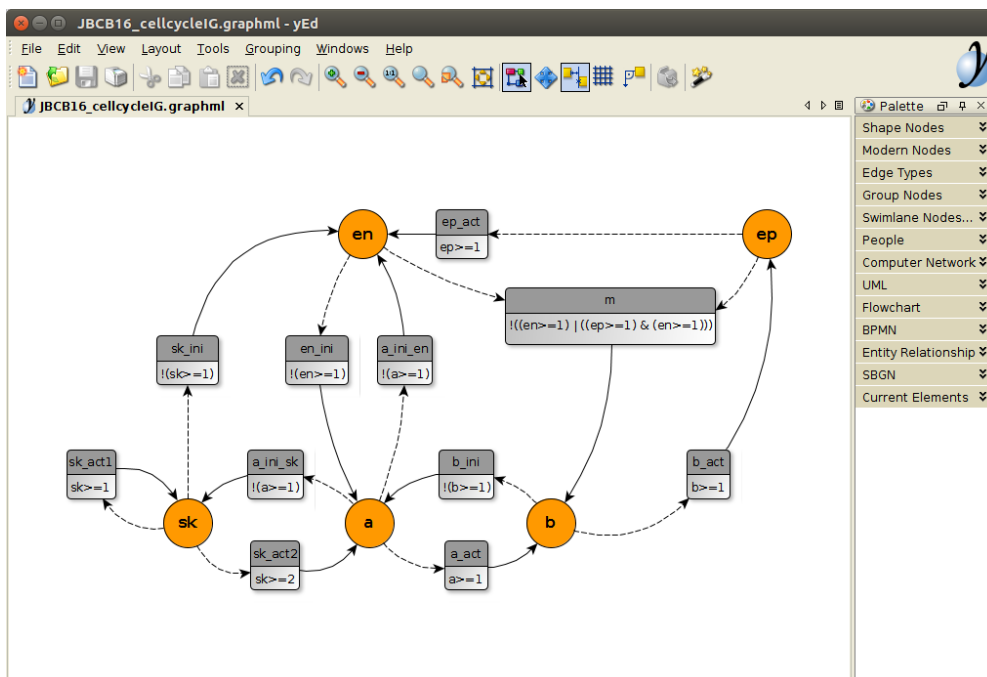
Table 2: **Verification of S/G2 and M/G1 checkpoints.** H_m is the set of models satisfying Hoare and Snoussi constraints. S_m is the set of selected models after model-checking of a temporal logic formula on each element of H_m .

A step-by-step user guide of TotemBioNet applied to the tool paper's use case

This scenario is based on the 5th experiment of use case, tool paper's table 1. TotemBioNet and its user manual are available at <https://gitlab.com/totembionet/totembionet>. Associated benchmarks are available at <http://www.i3s.unice.fr/~comet/DOCUMENTS/BenchTotemBioNet.zip>.

Step 1: Build the interaction graph with yEd

The yEd graph editor can be downloaded at : <https://www.yworks.com/products/yed>.



↓
JBCB16_cellcycleIG.graphml

Step 2: Generate static part of TotemBioNet's .smb input file

```
$ totembionet -yed ./JBCB16_cellcycleIG.graphml
```



```
JBCB16_cellcycleIGFromYed.smb
```

```
VAR
sk = 0 .. 2;
a = 0 .. 1;
b = 0 .. 1;
ep = 0 .. 1;
en = 0 .. 1;

REG
sk_act1 [sk>=1] => sk;
a_ini_sk [!(a>=1)] => sk;
sk_act2 [sk>=2] => a;
b_ini [!(b>=1)] => a;
en_ini [!(en>=1)] => a;
a_act [a>=1] => b;
m[!((en>=1)|((en>=1)&(ep>=1)))]=>b;
b_act [(b>=1)] => ep;
sk_ini [!(sk>=1)] => en;
a_ini_en [!(a>=1)] => en;
ep_act [ep>=1] => en;
```

The tool automatically returns variables' upper bounds equal to their outgoing degree. Each variable's upper bound must be manually decreased when several regulation thresholds are equal.

Step 3: Add properties into TotemBioNet's .smb input file

5_Experiment.smb

Content of JCB16_cellcycleIGFromYed.smb

+

```
#####
HOARE
PRE : {sk=0,ep=0,a=0,b=0,en=1}

TRACE:
Forall((sk+; sk+; en-),
(sk+; en-; sk+),
(en-; sk+; sk+));
a+; sk-; sk-; b+;
Forall((a-;ep+), (ep+;a-));
en+; b-; ep-;

POST: {sk=0,ep=0,a=0,b=0,en=1}

#####
FAIRCTL
phicyclic=((sk=0 & ep=0 & a=0 & b=0 & en =1)->
AX(AF(sk=0 & ep=0 & a=0 & b=0 & en =1)));

###
CTL
phiG2M=((sk=0 & ep=0 & a=1 & b=1 & en=0) ->
(!((EX((en=1) & EX((a=0) & EX((ep=1) ))))
| (EX((en=1) & EX((ep=1) & EX((a=0) ))))
| (EX((a=0) & EX((en=1) & EX((ep=1) ))))
| (EX((ep=1) & EX((en=1) & EX((a=0) )))))));

phiG1S=((sk=0 & ep=0 & a=0 & b=0 & en=1) ->
(!((EX((a=1) & EX((sk=1) & EX((en=0) & EX((sk=2))))))
| (EX((sk=1) & EX((a=1) & EX((en=0) & EX((sk=2))))))
| (EX((sk=1) & EX((en=0) & EX((a=1) & EX((sk=2))))))
| (EX((a=1) & EX((en=0) & EX((sk=1) & EX((sk=2))))))
| (EX((sk=1) & EX((a=1) & EX((sk=2) & EX((en=0))))))
| (EX((sk=1) & EX((sk=2) & EX((a=1) & EX((en=0))))))
| (EX((a=1) & EX((sk=1) & EX((sk=2) & EX((en=0))))))
| (EX((en=0) & EX((a=1) & EX((sk=1) & EX((sk=2))))))
| (EX((en=0) & EX((sk=1) & EX((a=1) & EX((sk=2))))))));
```

Step 4: Call TotemBioNet with .smb input file

```
$ totembionet -csv ./5_Experiment.smb
```



```
./hoare/5_Experiment.out  
./5_Experiment.out  
./5_Experiment.csv
```

Step 5: Content of TotemBioNet output files

The weakest precondition (*wp*) calculated from the Hoare triple is in:

```
./hoare/5_Experiment.out
```

```
((sk=0) & (ep=0) & (a=0) & (b=0) & (en=1)))  
& (((K_sk:a_ini_sk>0) & (K_sk:sk_act1:a_ini_sk>1)  
& (K_en:a_ini_en<1) & (K_a:sk_act2:b_ini:en_ini>0)  
& (K_sk:sk_act1<2) & (K_sk:sk_act1<1)  
& (K_b:a_act:m>0) & ((K_a:en_ini<1)  
& (K_ep:b_act>0) & (K_en:sk_ini:a_ini_en:ep_act>0)  
& (K_b<1) & (K_ep<1)))))) & (K_ep:b_act>0)  
& (K_a:en_ini<1) & (K_en:sk_ini:a_ini_en:ep_act>0)  
& (K_b<1) & (K_ep<1)))))))))) & (K_sk:a_ini_sk>0)  
& (K_en:a_ini_en<1) & (K_sk:sk_act1:a_ini_sk>1)  
& (K_a:sk_act2:b_ini:en_ini>0) & (K_sk:sk_act1<2)  
& (K_sk:sk_act1<1) & (K_b:a_act:m>0)  
& ((K_a:en_ini<1) & (K_ep:b_act>0)  
& (K_en:sk_ini:a_ini_en:ep_act>0) & (K_b<1)  
& (K_ep<1)))))) & (K_ep:b_act>0) & (K_a:en_ini<1)  
& (K_en:sk_ini:a_ini_en:ep_act>0) & (K_b<1)  
& (K_ep<1)))))))))) & (K_en:sk_ini:a_ini_en<1)  
& (K_sk:a_ini_sk>0) & (K_sk:sk_act1:a_ini_sk>1)  
& (K_a:sk_act2:b_ini:en_ini>0) & (K_sk:sk_act1<2)  
& (K_sk:sk_act1<1) & (K_b:a_act:m>0)  
& ((K_a:en_ini<1) & (K_ep:b_act>0)  
& (K_en:sk_ini:a_ini_en:ep_act>0)  
& (K_b<1) & (K_ep<1)))))) & (K_ep:b_act>0)  
& (K_a:en_ini<1) & (K_en:sk_ini:a_ini_en:ep_act>0)  
& (K_b<1) & (K_ep<1))))))))))
```

The conjunctive or disjunctive form of wp and results of model-checking are in:

./5_Experiment.out

```

Hoare result is a conjunction of conditions on K.
100540 models have been removed.

##### Result of model checking #####
# Total number of models 260
# Selected models: 28
# Computation time: 2s999ms

```

Selected (OK) and rejected (KO) parameterizations are saved in ./5_Experiment.csv. The first line contains the complete list of K parameters. For example $K_{sk:a_ini_sk:sk_act1}$ is the parameter of sk variable that denotes the presence of two resources: a_ini_sk and sk_act1 . 28 parameterizations are selected in the 5th experiment, so the file contains 28 *OK models* and 232 *KO models*.

./5_Experiment.csv: OK model section

n°	valid	K_{sk}	$K_{sk:a_ini_sk}$	$K_{sk:sk_act1}$	$K_{sk:a_ini_sk:sk_act1}$	K_a	$K_{a:b_ini}$	$K_{a:en_ini}$	$K_{a:sk_act2}$	$K_{a:b_ini:en_ini}$	$K_{a:en_ini:sk_act2}$	$K_{a:b_ini:sk_act2}$	$K_{a:b_ini:en_ini:sk_act2}$	K_b	$K_{b:a_act}$	$K_{b:m}$	$K_{b:a_act:m}$	K_{ep}	$K_{ep:b_act}$	K_{en}	$K_{en:a_ini_en}$	$K_{en:ep_act}$	$K_{en:sk_ini}$	$K_{en:a_ini_en:ep_act}$	$K_{en:ep_act:sk_ini}$	$K_{en:a_ini_en:sk_ini}$	$K_{en:a_ini_en:ep_act:sk_ini}$	
1	OK	0	1	0	2	0	0	0	0	0	0	0	1	0	0	0	1	0	1	0	0	0	0	0	0	0	1	
2	OK	0	1	0	2	0	0	0	0	1	0	0	1	0	0	0	0	1	0	1	0	0	0	0	0	0	0	1
3	OK	0	1	0	2	0	0	0	0	0	1	0	1	0	0	0	1	0	1	0	0	0	0	0	0	0	0	1
4	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	0	0	1	0	1	0	0	0	0	0	0	0	0	1
5	OK	0	1	0	2	0	0	0	0	0	0	0	1	0	0	1	0	1	0	1	0	0	0	0	0	0	0	1
6	OK	0	1	0	2	0	0	0	0	1	0	0	1	0	0	1	0	1	0	1	0	0	0	0	0	0	0	1
7	OK	0	1	0	2	0	0	0	0	0	1	0	1	0	1	0	1	0	1	0	0	0	0	0	0	0	0	1
8	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	1	0	1	0	1	0	0	0	0	0	0	0	0	1
9	OK	0	1	0	2	0	0	0	0	0	1	0	1	0	0	1	1	0	1	0	0	0	0	0	0	0	0	1
10	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	0	1	1	0	1	0	0	0	0	0	0	0	0	1
11	OK	0	1	0	2	0	0	0	0	1	0	1	0	1	1	1	0	1	0	1	0	0	0	0	0	0	0	1
12	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	1	1	1	0	1	0	0	0	0	0	0	0	0	1
13	OK	0	1	0	2	0	0	0	0	0	0	0	1	0	0	0	1	0	1	0	0	0	0	0	1	0	0	1
14	OK	0	1	0	2	0	0	0	0	1	0	0	1	0	0	0	1	0	1	0	0	0	0	0	1	0	0	1
15	OK	0	1	0	2	0	0	0	0	0	1	0	1	0	0	0	1	0	1	0	0	0	0	0	1	0	0	1
16	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	0	0	1	0	1	0	0	0	0	0	1	0	0	1
17	OK	0	1	0	2	0	0	0	0	0	0	0	1	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
18	OK	0	1	0	2	0	0	0	0	1	0	0	1	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
19	OK	0	1	0	2	0	0	0	0	0	1	0	1	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
20	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	1	0	1	0	1	0	0	0	0	0	1	0	0	1
21	OK	0	1	0	2	0	0	0	0	0	0	0	1	0	0	1	1	0	1	0	0	0	0	0	1	0	0	1
22	OK	0	1	0	2	0	0	0	0	1	0	0	1	0	0	1	1	0	1	0	0	0	0	0	1	0	0	1
23	OK	0	1	0	2	0	0	0	0	0	1	0	1	0	0	1	1	0	1	0	0	0	0	0	1	0	0	1
24	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	0	1	1	0	1	0	0	0	0	0	1	0	0	1
25	OK	0	1	0	2	0	0	0	0	0	0	0	1	0	1	1	1	0	1	0	0	0	0	0	1	0	0	1
26	OK	0	1	0	2	0	0	0	0	1	0	0	1	0	1	1	1	0	1	0	0	0	0	0	1	0	0	1
27	OK	0	1	0	2	0	0	0	0	0	1	0	1	0	1	1	1	0	1	0	0	0	0	0	1	0	0	1
28	OK	0	1	0	2	0	0	0	0	1	1	0	1	0	1	1	1	0	1	0	0	0	0	0	1	0	0	1

The last column of `./5_Experiment.csv`, named *Explanation*, is filled in only for KO parameterizations. It provides an explanation of model rejection by giving the unsatisfied formulas. Only 5 out of the 232 *KO models* are shown below:

`./5_Experiment.csv`: extract of KO model section

n°	valid	K_sk	K_sk:a_ini_sk	K_sk:sk_act1	K_sk:a_ini_sk:sk_act1	K_a	K_atb_ini	K_a:en_ini	K_a:sk_act2	K_atb_ini:en_ini	K_a:en_ini:sk_act2	K_atb_ini:sk_act2	K_atb_ini:en_ini:sk_act2	K_b	K_b:a_act	K_b:m	K_b:a_act:m	K_ep	K_ep:b_act	K_en	K_en:a_ini_en	K_en:ep_act	K_en:sk_ini	K_en:a_ini_en:ep_act	K_en:ep_act:sk_ini	K_en:a_ini_en:sk_ini	K_en:a_ini_en:ep_act:sk_ini	
29	KO	0	1	0	2	0	0	0	0	0	0	1	1	0	0	0	1	0	1	0	0	0	0	0	0	0	0	1
47	KO	0	1	0	2	0	0	0	0	0	0	0	1	0	0	1	1	0	1	0	0	0	0	0	0	0	0	1
105	KO	0	1	0	2	0	0	0	0	0	0	0	1	0	0	0	1	0	1	0	0	0	0	0	0	1	0	1
109	KO	0	1	0	2	0	0	0	0	0	0	1	1	0	0	0	1	0	1	0	0	0	0	0	1	0	1	
135	KO	0	1	0	2	0	0	0	0	0	0	1	1	0	0	1	1	0	1	0	0	0	0	0	1	0	1	

`./5_Experiment.csv`: extract of KO model explanations

n°	Explanation
29	[phiG1S]
47	[phicyclic]
105	[phiG2M]
109	[phiG2M, phiG1S]
135	[phicyclic, phiG2M, phiG1S]