

# Exhaustive dynamic analysis of reachability and stable states of biological regulatory networks modeled in Process Hitting using Answer Set Programming

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**Abstract.** The Process Hitting is a recently introduced framework to model concurrent processes. It is notably suitable to model biological regulatory networks with partial knowledge of co-operations by defining the most permissive dynamics. In this paper, we explain the methods we developed with ASP to find the fixed points or states in which it is not possible any more to have evolutions of the model. We also aim at solving the problem of reachability that consists of deciding if, starting from a given initial state, it is possible to reach a given local state. Finally, we illustrate the merits of our methods by applying them to various biological examples and comparing the results with existing approaches. We show that our approach succeeds in processing large models.

**Keywords:** Process Hitting, stable states, fix points, reachability, Answer Set Programming

## 1 Introduction

As regulatory phenomena play a crucial role in biological systems, they need to be studied accurately. Biological Regulatory Networks (BRNs) consist in sets of either positive or negative mutual effects between the components. With the purpose of analyzing these systems, they are often modeled as graphs which make it possible to determine the possible evolutions of all the interacting components of the system. Indeed, in order to address the formal checking of dynamical properties within very large BRNs, we recently use a new formalism, named the “Process Hitting” (PH) [5], to model concurrent systems having components with a few qualitative levels. A PH describes, in an atomic manner, the possible evolutions of a “process” (representing one component at one level) triggered by the hit of at most one other “process” in the system. This particular structure makes the formal analysis of BRNs with hundreds of components tractable.

PH is suitable, according to the precision of this information, to model BRNs with different levels of abstraction by capturing the most general dynamics. The objectives of the work presented in this paper are the following.

Firstly, we show that starting from one PH model, it is possible to find all possible stable states (fixed points [6]). We perform an exhaustive search of the possible states, combination processes, one process from each sort and then check if it is a fixed point.

The second phase of our work consists in computing the dynamics. It consists in determining from a known initial state the possible next states of the PH model. Finally we verify if we can reach a specific state of one or several component (gene or protein). The results are ensured to respect the PH dynamics.

Our contribution is from the results that allowed to determine the stable states, we propose to evaluate the benefits of the Answer Set Programming (ASP) [2] to compute them. ASP has been proven efficient to tackle models with a large number of components and parameters. Our aim here is to assess its potential w.r.t. the computation of some dynamical properties of the PH model. In this paper, we show that ASP turns out to be effective for these enumerative searches which justifies its use. The benefit of our approach is that it makes possible to get the minimal paths to reach our goal(s) also we can verify if it is possible after a given number of steps.

## 2 Preliminary definitions

In order to develop the new methods we have used two main frameworks. The first is a logic programming language: the Answer Set Programming (ASP). Precisely it is a logic programming language on which I wrote my bibliographic research report. The second framework is the Process Hitting (PH) a new formalism for the representation of biological regulatory networks. We will see later in this section more details about these two frameworks.

### 2.1 Answer Set Programming

The ASP is a declarative programming paradigm with known semantics as the semantics of answer sets. This paradigm allows the programmer to specify what the problem to be solved and not how to solve it. ASP programs are written in AnsProlog\* (short for "Answer Set Programming in Logic" with superscript \*). These programs are composed of a set of facts and a set of rules from which other facts can be derived. A consistent set of facts that can be derived from a program using the rules is called "answer set" for the program. The sets of possible responses to an AnsProlog\* program are calculated with a program called a solver.

**Logic program** Consider an ASP logic program, as each rule is an ordered pair, [2] :

$$head \leftarrow body. \tag{1}$$

with *Head* and *Body* sets of literals.

This simple language has many advantages that make it one of the most effective in terms of knowledge representation, reasoning and declarative problem solving.

## The syntax and semantic of ASP

### 1. L'alphabet

According [2], the alphabet of the axiom (or just the alphabet) of the framework of the answer set is comprised of seven classes of symbols:

- Variables,
- Constants,
- Symbols of functions,
- Symbols of predicates,
- Connectors,
- Symbols of punctuation, and
- Special symbol  $\perp$

All these classes vary from one alphabet to another, but the sets of 5<sup>th</sup> and 6<sup>th</sup> classes (Connectors and symbols of punctuation) are defined as follows:

- Connectors with  $\{\neg, \text{or}, \leftarrow, \text{not}, ', '\}$
- Symbols of punctuation with  $\{'(', ')', '., '\}$

The other classes remain constant as they meet some informal agreement. In general, variables start with a capital letter and contain letters and numbers (X,Y,...). The constants, symbols and predicates follow the same rule, but they begin with a lowercase letter (f,g,a,b,...). Sometimes there is the addition of a supplementary agreement which covers the letters used [2] :

What seems a little fuzzy is the concept of predicate. Indeed the word predicate can be an innovation of the new grammar. Let's consider the following sentence:

*Socrates was an athenian.*

*Socrate* is the subject and *was an athenian* is the predicat. This sentence can be noted in ASP by :

*athenian(Socrate)  $\leftarrow$   $\top$ .* with  $\top$  is the symbol of True.

### 2. The rules Comme nous l'avons déjà mentionné, une règle est de la forme:

$$head \leftarrow body. \quad ((1))$$

$$A \leftarrow B^+, \text{ not } B^-. \quad (2)$$

This rule is then equivalent to:

$$L_0 \leftarrow L_1, \dots, L_m, \mathbf{not} L_{m+1}, \dots, \mathbf{not} L_n. \quad (3)$$

And this rule states:

$$\left. \begin{array}{l} \text{Si } L_1, \dots, L_m \text{ are } \mathbf{true} \\ \text{et } L_{m+1}, \dots, L_n \text{ are } \mathbf{false} \end{array} \right\} \text{Then } L_0 \text{ is } \mathbf{true}$$

with  $L_i$  literals and  $k \geq 0$ ,  $m \geq k$  and  $n \geq m$ .

**Head** is also called conclusion

**Body** is also called premise

Special cases rules [4,?] :

- A rule is **constant** if all literals are constants (noted with **ground**);
- A **fact** or a **reality** : it is a rule with an empty body. It can be written even without the arrow  $\leftarrow$  :

$$L_0. \text{ or } L_0 \leftarrow \top. \quad (4)$$

- A **constraint**: This false symbol to the head (*head*) is often eliminated and the constraint will be written generally as:

$$\leftarrow L_1, \dots, L_m, \mathbf{not} L_{m+1}, \dots, \mathbf{not} L_n. \quad (5)$$

We say that a set of literals  $X$  violates the constraint (5) if  $\{L_1, \dots, L_m\} \subseteq X$  and  $\{L_{m+1}, \dots, L_n\} \not\subseteq X$ . If we have a program contain this type of rule 5, then  $X$  is an answer set of the program  $\pi$  if and only if  $X$  is an *answer set* of  $\pi \setminus \{r\}$ , with  $r$  such a constraint 5 and  $X$  does not violate the constraint 5.

- Cardinality constraints:  
It is literals extended in the following form:

$$l \{q_1, \dots, q_m\} u. \quad (6)$$

with  $m \geq 1$ ,  $l$  an integer and  $u$  can be an integer or by default the infinite if it does not exist.  $l$  and  $u$  are the lower and upper limits of the cardinality of the subsets of  $\{q_1, \dots, q_m\}$  that satisfy the answer sets. These literals are constrained ( $q_i$ ) may occur in the head and body of the rule. Cardinality constraint is satisfied in an answer set  $X$ , if the number of atoms of  $\{q_1, \dots, q_m\}$  belonging to  $X$  is between  $l$  and  $u$ . In other words:

$$l \leq |\{q_1, \dots, q_m\} \cap X| \leq u$$

with  $\cap$  symbol of intersection and  $|A|$  is the cardinality of the set  $A$ .

**Modelisation of a problem with ASP** We can consider that the construction of models is one of the fundamental components of the scientific process. It concerns all systems we seek to control. A model has two main features [1]:

- it is a simplification of a given system
- it allows an action on the system

Models offer the possibility to provide a solution to a problem identified as such. This concept model can address another angle issues related to representation process.

1. **Modeling steps** The modeling process can be regarded as a special form of representation whose operations are detailed in [2]. Logical ASP programs follow the strategy of "generate and test". This strategy includes four steps:
  - Enumerate with facts;
  - Explain with the rules;
  - Generate all the possibilities with cardinalities, and finally;
  - Filter with constraints.

2. **Resolving an ASP program**

From [4] and [3], the couple Grounder and Solver usually work together: the grounder used to remove variables in order to achieve a constant program and the solver computes all answer sets for stabilized programs generated by the grounder.

Example of Grounder: GRINGO, DLV, LPARSE

Example of ASP Solver : SMODELS, DLV, CMODELS, CLASP...

The combined use of the Grounder and the Solver specifies major programs in a compact, using rules with schematic variables and other abbreviations. Both systems employ grounding algorithms that work quickly and simplify the program. In developing these new methods we worked with CLINGO which is a combination of grounder GRINGO and solver clasp.

## 2.2 Process Hitting

Definition 1 introduces the Process Hitting (PH) [5] which allows to model a finite number of local levels, called *processes*, grouped into a finite set of components, called *sorts*. A process is noted  $a_i$ , where  $a$  is the sort's name, and  $i$  is the process identifier within sort  $a$ . At any time, exactly one process of each sort is *active*, and the set of active processes is called a *state*.

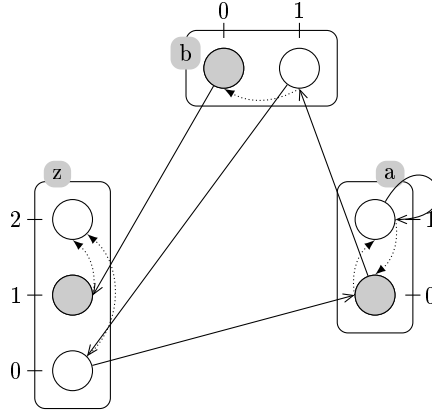
The concurrent interactions between processes are defined by a set of *actions*. Actions describe the replacement of a process by another of the same sort conditioned by the presence of at most one other process in the current state. An action is denoted by  $a_i \rightarrow b_j \uparrow b_k$ , which is read as " $a_i$  hits  $b_j$  to make it bounce to  $b_k$ ", where  $a_i, b_j, b_k$  are processes of sorts  $a$  and  $b$ , called respectively *hitter*, *target* and *bounce* of the action. We also call a *self-hit* any action whose hitter and target sorts are the same, that is, of the form:  $a_i \rightarrow a_i \uparrow a_k$ .

**Definition 1 (Process Hitting).** A Process Hitting is a triple  $(\Sigma, L, \mathcal{H})$ :

- $\Sigma = \{a, b, \dots\}$  is the finite set of sorts;
- $L = \prod_{a \in \Sigma} L_a$  is the set of states with  $L_a = \{a_0, \dots, a_{l_a}\}$  the finite set of processes of sort  $a \in \Sigma$  and  $l_a$  a positive integer, with  $a \neq b \Rightarrow L_a \cap L_b = \emptyset$ ;
- $\mathcal{H} = \{a_i \rightarrow b_j \uparrow b_k \in L_a \times L_b^2 \mid (a, b) \in \Sigma^2 \wedge b_j \neq b_k \wedge a = b \Rightarrow a_i = b_j\}$  is the finite set of actions.

**Definition 2 (Action).** An action is noted  $a_i \rightarrow b_j \uparrow b_k$  where  $a_i$  is a process of sort  $a$  and  $b_j, b_k$  two processes of sort  $b$ . When  $a_i = b_j$ , such an action is referred as a self-action and  $a_i$  is called a self-hitting process.

Example. Figure 1 represents a PH  $(\Sigma, L, \mathcal{H})$  with three sorts ( $\Sigma = \{a, b, c\}$ ) and:  $L_a = \{a_0, a_1\}$ ,  $L_b = \{b_0, b_1, b_2\}$ ,  $L_c = \{c_0, c_1\}$ .



**Fig. 1.** A PH model example with three sorts:  $a, b$  and  $z$  ( $a$  is either at level 0 or 1,  $b$  at either level 0 or 1 and  $z$  at either level 0, 1 or 2). Circles represent the processes, boxes represent the sorts, and the actions are drawn by pairs of arrows in solid and dotted lines. The grayed processes stand for a possible initial state.

**Definition 3 (Next state).** Let  $(\Sigma, L, \mathcal{H})$  be a Process Hitting and  $s \in L$  be one of its states. The set of the next possible states for  $s$  are computed as follows:

$$next(s) = s[b_k/b_j] \mid \exists (a_i, b_j) \in s^2, \exists b_k \in L_b, a_i \rightarrow b_j \uparrow b_k \in \mathcal{H}$$

**Definition 4 (Stable state or fixed point).** Let  $PH = (\Sigma, L, \mathcal{H})$  be a Process Hitting and  $s \in L$  be a state,  $s$  is a stable state for  $PH$  if and only if  $next(s) = \emptyset$ .

**Definition 5 (Lemme: Stable state or fixed point).** Let  $PH = (\Sigma, L, \mathcal{H})$  be a Process Hitting and  $s \in L$  be a state,  $s$  is a stable state for  $PH$  there is no playable action at the state  $s$ :

$$\forall h \in \mathcal{H}, frappeur(h) \notin s \vee cible(h) \notin s$$

**Definition 6 (Reachability question).**

If  $\varsigma \in L$  is a state and  $A \in \mathbf{Proc}$  a set of processes, we note  $\mathcal{P}(\varsigma, A)$  pour la question d'accessibilité:

$$\exists ?\delta \in \mathbf{Sce}(\varsigma), A \subseteq ([\varsigma] \cdot \delta) \text{ (i.e. } \forall a_i \in A, (\varsigma \cdot \delta)[a] = a_i).$$

With  $\mathbf{Sce}(\varsigma)$  is the set of successively playable actions from the state  $\varsigma$ .

**Definition 7 (Playable action).** Let  $PH = (\Sigma, L, \mathcal{H})$  be a Process Hitting and  $s \in L$  a state of  $PH$ . We say that the action  $h = a_i \rightarrow b_j \uparrow b_k \in \mathcal{H}$  is playable at the state  $s$  if and only if  $\text{hitter}(h) = a_i \in s$  and  $\text{target}(h) = b_j \in s$  (i.e.  $s[a] = a_i$  et  $s[b] = b_j$  )

The resulting state after playing the action  $h$  at  $s$  is denoted by  $(s \cdot h)$  or  $(s \cdot h)[b] = b_k$  and  $\forall c \in \Sigma, c \neq b, (s \cdot h)[c] = s[c]$ .

### 3 Fixed point implementation

The study of fixed points (or basins of attraction) provides an important understanding of the different behaviors of a BRN (Biological Regulatory Network) [6]. The fixed point is a stable state of the BRN in which it is not possible any more to have new changes. Let  $(\Sigma, L, H)$  be a Process Hitting . It has been shown that a state  $s \in L$  is a fixed point of the Process Hitting if and only if  $s$  has no next state [5] i.e. there is no playable action at this state. In fact a steady state of a Process Hitting network is a set of processes with exactly one process of each sort as well as every process has no hit with the other selected ones (process with self-hit cannot be apart of a stable state).

#### 3.1 Process Hitting network traduction

In order to handle a PH biological network, it was necessary first to present it with ASP. To do this we chose the predicates: *sort*, *process* and *action*. Below is an example of PH network shown in ASP.

Example (Example PH network with ASP). If we try to present the network of Figure 1 we will have:

```

1 sort("a"). sort("b"). sort("z").
2 process("a", 0..1). process("b", 0..1). process("z", 0..2).
3 action("a",0,"b",1,0). action("a",1,"a",1,0). action("b",1,"z",0,2).
4 action("b",0,"z",1,2). action("z",0,"a",0,1).
```

Line 1 shows every sort of network with the predicate comes out and in parentheses the name of the sort. In line 2 there is a list of processes corresponding to each *sort* for example the sort "z" has 3 process numbered from 0 to 2, this numbering is provided by the  $2^{nd}$  parameter of the predicate *process*("z", 0..2). Finally we find all network actions that are defined in lines 3 and 4, for example the first action *action*("a",0,"b",1,0) is an action from  $a_0$  to  $b_1$  to bound so  $b$  from  $b_1$  into  $b_0$ .

### 3.2 Search of fixed points

Parler du point fixe et ses interrêts.

Decrire la nouvelle méthode qui permettent de determiner les états stables d'un réseau

In fact we have to translate the definition of a stable state in a method developped in ASP. So first we eliminate all processes with self-hit and we save them in the predicate "*shownProcess*":

```

5 hiddenProcess(A,I) ← action(A,I,B,J,K), A=B, process(A,I),process(B,J),
6                      process(B,K).
7 shownProcess(A,I) ← not hiddenProcess(A,I), process(A,I).
```

Then we have to browse this graph and extract all possible combinations of shown processes by choosing a process from each sort.

```

8 1 { selectProcess(A,I) : showProcess(A,I) } 1 ← sort(A).
```

It now remains to check each combination of processes whether it is a fix point. For this we use a special type of ASP rules: a **constraint**. The idea of constraints is based on the fact that a solution would be eliminated if it does not satisfy the constraint. For our problem a combination is eliminated if there is an action between two of the selected process:

```

9 ← hit(A,I,B,J), selectProcess(A, I), selectProcess(B, J), A!= B.
```

*Example 1.* Considering the last graphical presentation of a precess hitting Figure 1, there are 3 sorts *a*, *b* have 2 levels and *z* has 3 so we can find  $2 * 2 * 3 = 12$  states (whatever they can be reached or not). If we verify wether there exist fixed points we deduce that we have only one:  $\langle b\_0, z\_2, a\_0 \rangle$ . It is clear in Figure 1 that there is no action between each two processes. Besides our new ASP method proof this and returns also the same answer:

```

Answer 1 : fixProcess(a, 0), fixProcess(b, 0), fixProcess(z, 2)
```

## 4 Dynamic network evolution

In this section, we will present firstly how to determine using ASP the possible evolution of a biological network after a finite number of steps. Then we answer to the reachability question; which evolutions that allow the achievement of goals (future active processes) from a known initial state?

### 4.1 Future states identification

From an initial known state, a PH network can evolve into several new states after a few steps. The predicate "*time(0..n)*" sets the number of steps we want to play. For example if the biologist wants to know which states are reachable after 10 stages, it has only to replace *n* by 10, and he will have "*time(0..10)*". To compute the future states it is necessary to define an initial state of the network.



So for initializing the active processes we add in the ASP script representing the PH network the following rule for each sort. We note that this rules are generated automatically at the time of translating PH network to ASP and by default the level of all active processes is 0.

```
10 init(activeProcess("a",0)).
```

with  $a$  is the name of the sort and 0 the index of the active process.

The dynamic of a network is realised thanks to its actions. So to identify the future states we should start by identifying the playable actions for each state. We recall that an action is playable when both precesses: hitter and taget are active (Definition 7). In ASP the predicate "*playableAction*( $A, I, B, J, K, T$ )" it is true when the processes ( $A, I$ ) and ( $B, J$ ) are active at the step  $T$ . Our approach is to study the asynchronous dynamic so only one action is playable at the same step ( $T$ ). That's why between two successive states there will be one change of one sort. We present this by the predicate "*activeFromTo*( $B, J, K, T$ )". It means that in the sort  $B$  the active process change from index  $J$  to  $K$  at time step  $T$ . At the same step we can find many possible changes , that's why the rule (line 11) offers a set of all these possible changes thanks to braces. In fact the rule is encoded with a count atom at its head, which makes it a choice rule. Rule in line 12 filters any answer with more than 1 change at the same time  $T$  to ensure the asynchronous evolution.

```
11 {activeFromTo(B,J,K,T)} ← playableAction(A,I,B,J,K,T), instate(activeProcess(A,I),T),
12 instate(activeProcess(B,J),T), J!=K, time(T).
13 ← 2{ activeFromTo(B,J,K,T)}, time(T).
```

In order to determinate the next active processes at  $T+1$  we use the following rules :

```
14 instate(activeProcess(B,K),T+1) ← activeFromTo(B,J,K,T), time(T).
15 instate(activeProcess(A,I),T+1) ← instate(activeProcess(A,I),T), activeFromTo(B,J,K,T),
16 A!=B, time(T).
```

At the next step  $T + 1$  we find the new active process resulted from the predicate *activeFromTo* (line 13) as well as all the unchanged processes that correspond to the other sorts (line 15).

As a result it will be displayed all possible evolutions of the networks.

## 4.2 Reachability verification

- **known steps number** Parler de la méthode qui vérifie l'atteignabilité après un nombre fini de changements
- **unknown steps number: incremental approach (iclingo)** Parler de la méthode qui vérifie l'atteignabilité après un nombre indéfini de changements et qui retourne le chemin le plus court s'il existe pour atteindre ces objectifs

## 5 Experiments analysis

### 5.1 Benchmarks

### 5.2 Evaluation

### 5.3 Limitation

## 6 Conclusion and future directions

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