Structuring and linking of biological data guided by ontologies and the organizing principles of mathematical models InOvive Meeting

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People involved in the PhD

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#### Context

#### System biology, whole cell modeling, BioSys group @INRAE



- A modeling approach that is a powerfull principle to understand the organisation of biological networks in bacteria
- Exploit this principle in the field of knowledge representation
- Molecular scale, entities are molecules

H2O, Glucose, DNA, Proteins, Enzymes, ... are molecules In the cell molecules are transformed through **biochemical reactions** (BR)



One biochemical reaction

A network of reactions

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## Molecular Biology

Molecular knowledge is complex different types of molecules different types of processes different scales Experimental data

heterogeneous, different types fragmented, not always available

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Molecular knowledge is complex different types of molecules different types of processes different scales Experimental data are heterogeneous, different types fragmented, not always available



#### General questions

Taking into account the facts that (i) molecular biology provides complex knowledge and (ii) experimental data are heterogeneous and fragmented:

- Is it possible to link together these data and this knowledge?
- Is it possible to check the consistency of these data and knowledge?
- Is it possible to refine the links with fresh data and knowledge while keeping the consistency?

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# First Steps / State of art

Formal representations of knowledge and data



- 1. Represent the molecular knowledge
  - The ontologies BiPON and BiPOM
  - They use the systemic approach to tackle the complexity
- 2. Represent the experimental data
  - Organize the data (since they are heterogeneous and fragmented)
  - Link this representation with the representation of knowledge

### The systemic approach



- Aim: simulate the behavior of a given system (for us the system is a bacteria)
- An interesting property: tackle the complexity of a given system
- How?: break down this system into connected sub-systems
- A sub-system: inputs, outputs and a function to fulfill
- The transformation of inputs to outputs is done through a mathematical model

#### BiPON/BiPOM and the systemic approach

In BiPOM a sub-system is a Biological Process



The Biochemical Reaction (BR) transforms the molecule S to the molecule P and is mediated by the molecule E:



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## **BiPON/BiPOM**

The limit of BiPON/BiPOM to represent experimental data

- Individuals that are inputs (resp. outputs or mediators) of a BiologicalProcess are always a molecule
- Ex: in BiPOM, individuals of class Chemical, Gene, (non)GeneProduct, Metabolite are molecules



- Well suited to represent BR since by definition a BR transform molecules to another molecules
- But BR are not directly relied to experimental data.

#### Mathematical models are relied to experimental data Ex: the speed of $S \rightarrow P$ is given by the following model



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- A mathematical model: variables and parameters
- The values of variables and parameters represent (experimental) data
- The model organizes variables and parameters:
  V<sub>i</sub> = f(S<sub>0</sub>, E, K<sub>cat</sub>, K<sub>m</sub>)
  - 1. The inputs of the model are  $S_0$ , E,  $K_{cat}$ ,  $K_m$
  - 2. The output of the model is  $V_i$

# Mathematical models are relied to experimental data

The model organizes (experimental) data



- The values of variables and parameters represent (experimental) data
- The model organizes these data: fluxomics = f(proteomics, metabolomics, parameter\_value)

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BiPON/BiPOm: no classes to represent these values

#### Questions from a knowledge representation perspective A new ontology for experimental data

Through variables and parameters, models are data organizers:

- 1. Can we find <u>classes</u> that represent variables and parameters?
- 2. Can we organize these classes like variables and parameters are organized in the models?



#### The new ontology Input and Output are inferred



 $\begin{array}{l} \textit{Input} \equiv \textit{Concentration} \sqcap \exists \textit{is\_read\_by.BiologicalProcess} \sqcup \textit{Parameter} \sqcap \exists \textit{is\_read\_by.BiologicalProcess} \\ \textit{Output} \equiv \textit{Flow} \sqcap \exists \textit{triggered\_by.BiologicalProcess} \end{array}$ 

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## Summary and discussion



1. Since Inputs/Outputs are different classes, how to connect *BiologicalProcesses?*  $S \longrightarrow P \longrightarrow Q \longrightarrow R$ 

The pool connects flows and concentration <sup>1</sup>

- 2. Generalization of Inputs/Outputs:
  - Inputs are the information that are the cause of the process (T°, pression, etc ...)
  - Outputs are matter exchange between the process and the env

#### Discussion

In the example  $S \longrightarrow P$ :

- 1. Where does the information come from?
- 2. Where does the matter exchange come from?



The speed v<sub>i</sub> contains the information, the inputs

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The BR contains matter exchange, the outputs

### Discussion

- 1. Where does the information come from?
- 2. Where does the matter exchange come from?



- The matter exchange is also avalaible in another type of models: Ordinary Differential Equations (ODEs)
- A stricking fact: 2 sets of math expressions (M1 and M2) contains all the information we need to populate the ontology

## Questions



- 1. Given only *M*1 and *M*2 which model the dynamic of a network of reactions, can we populate the ontology?
- 2. Conversely, can we populate the ontology with only BR (in XML)?
- 3. What is the motivation to populate the ontology with 2 different sources?

## Questions



- 1. Given only *M*1 and *M*2 which models the dynamic of a network of reactions, can we populate the ontology?
- 2. Conversely, can we populate the ontology with only BR (in XML)?
- 3. What is the motivation to populate the ontology with different sources?

### General questions

What is the motivation to populate the ontology with different sources?

Taking into account the facts that (i) molecular biology provides complex knowledge and (ii) experimental data are heterogeneous and fragmented:

- Is it possible to link together these data and this knowledge?
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- Is it possible to refine the links with fresh data and knowledge while keeping the consistency?

Since the information of different sources are represented with the same classes and relations, a linkage can be considered.

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### Perspectives

1. Consider a linkage and (possibly) evaluate the consistency: we need two sources of informations represented with the ontology

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2. Refine the links with another source of information

## Perspectives

Brief description of 2 examples of sources of informations

- > XML representation of reactions, Source 1 Monk et al. 2016
- M1 and M2, Source 2 Chassagnole et al. 2002
- 1. **Source 1** describes the reactions of the entire metabolic network of the bacteria *E.coli*.
- 2. **Source 2** is a model that study the dynamic of a sub-network of the metabolic network of the bacteria *E.coli*.

The ontology has been populated with each source, we have obtained 2 datasets:



#### In the perspective to link biological processes

The process named "PGI" in each dataset



Specific to dataset 2

We want to investigate contextual identity links

### Conclusion

#### Contributions:

- 1. An ontology that was originally designed to structure data but that would allow us to consider the linkage between different sources of informations
- 2. A RDF vocabulary for the representation and the querying of complex math expressions: the ODEs
- Short term perspective: investigate contextual identity links with the aid of datasets produced by us
- Middle term perspective, consider a new source of information that presents interesting characteristics:
  - 1. Another modelisation paradigm: constraints based instead of ODEs

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2. Covers a wider range biological processes

Thank you for your attention

Backup slides

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### A deeper look at the models

The experts help us to design the classes



## The function f is a process



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### The ontology



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#### The new ontology Input and Output are inferred



 $\begin{array}{l} \textit{Input} \equiv \textit{Concentration} \sqcap \exists \textit{is\_read\_by.BiologicalProcess} \sqcup \textit{Parameter} \sqcap \exists \textit{is\_read\_by.BiologicalProcess} \\ \textit{Output} \equiv \textit{Flow} \sqcap \exists \textit{triggered\_by.BiologicalProcess} \end{array}$ 

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Problem: query math expressions to extract biological processes, concentrations, flows

$$M_1 = \{r_1 = f_1(a), r_2 = f_2(b)\}$$

$$M_2 = \begin{cases} \text{Evolution} = \text{Production} & - \text{Degradation} \\ \frac{da}{dt} = & - & r_1 \\ \frac{db}{dt} = & r_1 & - & r_2 \\ \frac{dc}{dt} = & r_2 \end{cases}$$



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$$M_1 = \{r_1 = f_1(a), r_2 = f_2(b)\}$$

expr = "'Equals'('ex:r1', 'Function'('ex:a'))"
expr = "'Equals'('ex:r2', 'Function'('ex:b'))"

```
[] a math:Equals ;
   rdf:_1 <ex:r1> ;
   rdf:_2 [ a ode:Function ;
        rdf:_1 <ex:a> ] .
[] a math:Equals ;
   rdf:_1 <ex:r2> ;
   rdf:_2 [ a ode:Function ;
        rdf:_1 <ex:b> ] .
```

(	Evolution	=	Production	_	Degradation
	da dt	=		_	<i>r</i> <sub>1</sub>
$M_2 = \left\{ \right.$	$\frac{db}{dt}$	=	<i>r</i> 1	_	<i>r</i> <sub>2</sub>
	$\frac{dc}{dt}$	=	<i>r</i> <sub>2</sub>		

```
'expr'='''Ode'(\
''Equation'('Derivative'('ex:a','t'),\
''Function'('t',\
''Addition'('SignedTerm'('Minus'('Term'('ex:r1'))),\
'))),\
'Equation'('Derivative'('ex:b','t'),\
'Function'('t',\
'Addition'('SignedTerm'('Plus'('Term'('ex:r1'))),\
'SignedTerm'('Minus'('Term'('ex:r2')))\
))),\
'Equation'('Derivative'('ex:c','t'),\
'Function'('t',\)
```

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```
@prefix math: <http://www.example.org/math#> .
@prefix ode: <http://www.example.org/ode#> .
@prefix rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#> .
@prefix rdfs: <http://www.w3.org/2000/01/rdf-schema#> .
[] a ode:Ode :
    rdf:_1 [ a ode:Equation ;
            rdf:_1 [ a ode:Derivative ;
                    rdf:_1 <ex:a> ;
                    rdf: 2 :t ] :
            rdf:_2 [ a ode:Function ;
                    rdf:_1 _:t ;
                    rdf:_2 \Gamma a math:Addition ;
                             rdf:_1 [ a ode:SignedTerm ;
                                     rdf:_1 [ a math:Minus ;
                                             rdf:_1 _:_7 ] ] ] ] ;
    rdf:_2 \ \ a \ ode:Equation :
            rdf:_1 [ a ode:Derivative ;
                    rdf:_1 <ex:b> ;
                    rdf:_2 _:t ] ;
            rdf:_2 [ a ode:Function ;
```