

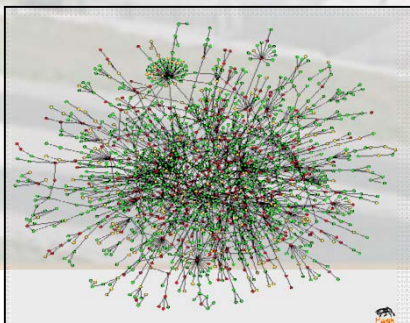
Colloquium Jacques Morgenstern

SysDiag, Montpellier, France

Biological Complex system modeling and engineering for diagnosis

**Ingénierie des systèmes biologiques synthétiques :
applications au diagnostic médical**

Franck Molina, CNRS





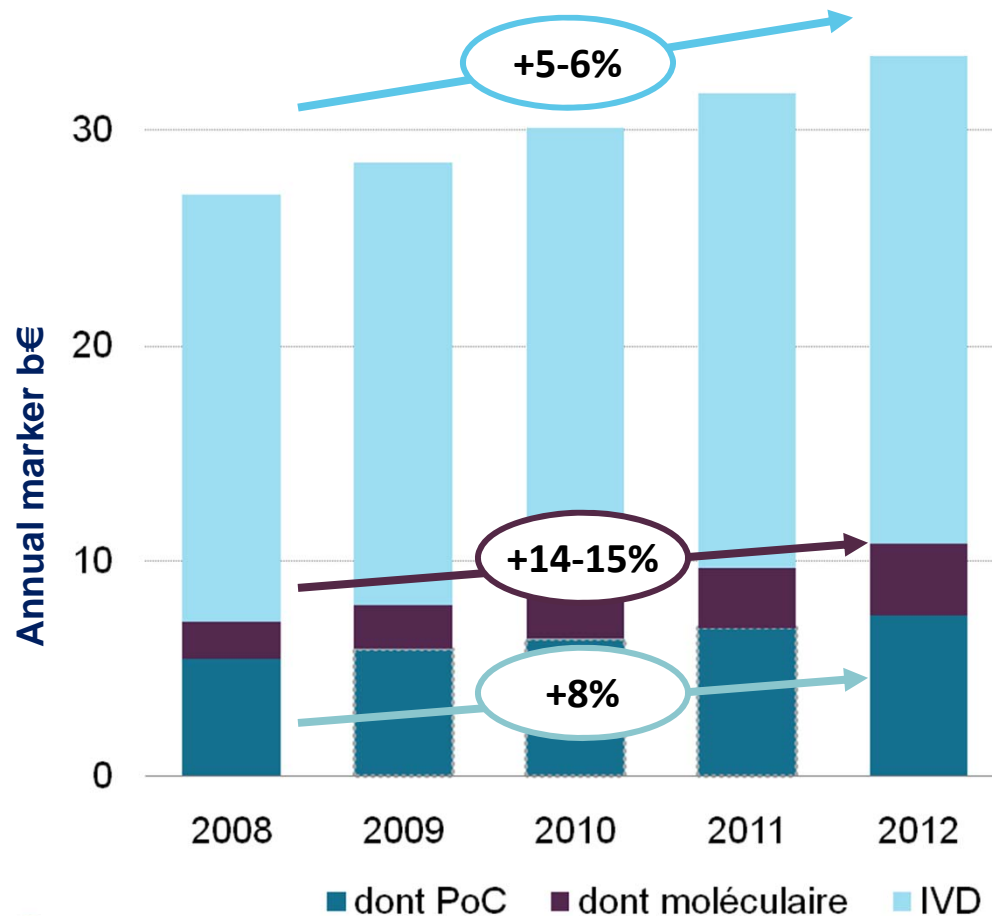
SysDiag's missions are :

- to understand the bases of chronic and multifactorial diseases (neurodegenerative and cardiovascular diseases, cancer, diabetes...)
- to identify new biomarkers associated with these pathological conditions
- to provide innovative solutions for clinicians that will contribute to improve the health and quality of life of their patients.





Diagnostic tests represent **less than 5%** of hospital cost and about **1.6 % of all** health cost. Their results influence up to **60-70% medical decision**



- ⇒ Change of paradigm with the convergence of diagnosis and therapy, going to personalized medicine and theranostic
- ⇒ New role of biomarker in biomedical and therapeutic, molecular assay development (nucleic acid, proteines, metabolites)
- ⇒ explosion of POC Technologies and of home monitoring/testing)
 - ⇒ Integration with IT
- ⇒ New interfaces electronic/biology (Biosensor)

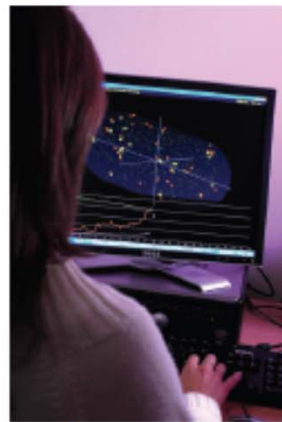


Interdisciplinarity : The SysDiag model

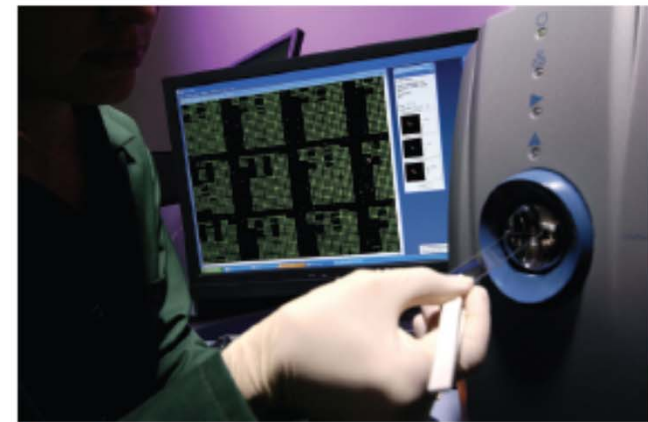
- Crossing disciplines for innovation in Diagnosis
- Combining experimental biology and complex systems modelling approaches
 - 70% experimentalists and 30% theoreticians



Cell culture automation



Bioinformatic & modelling

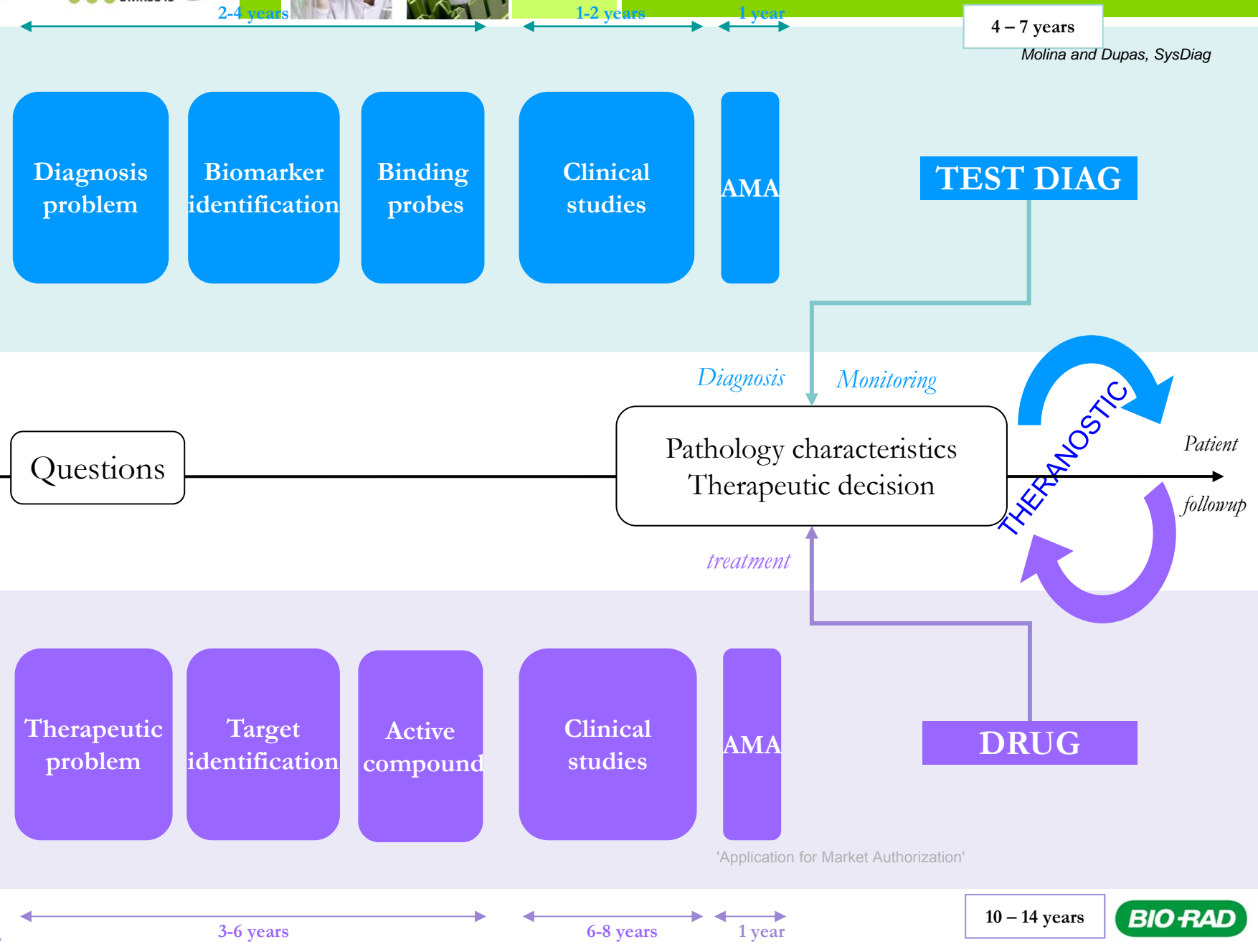


Experimentation

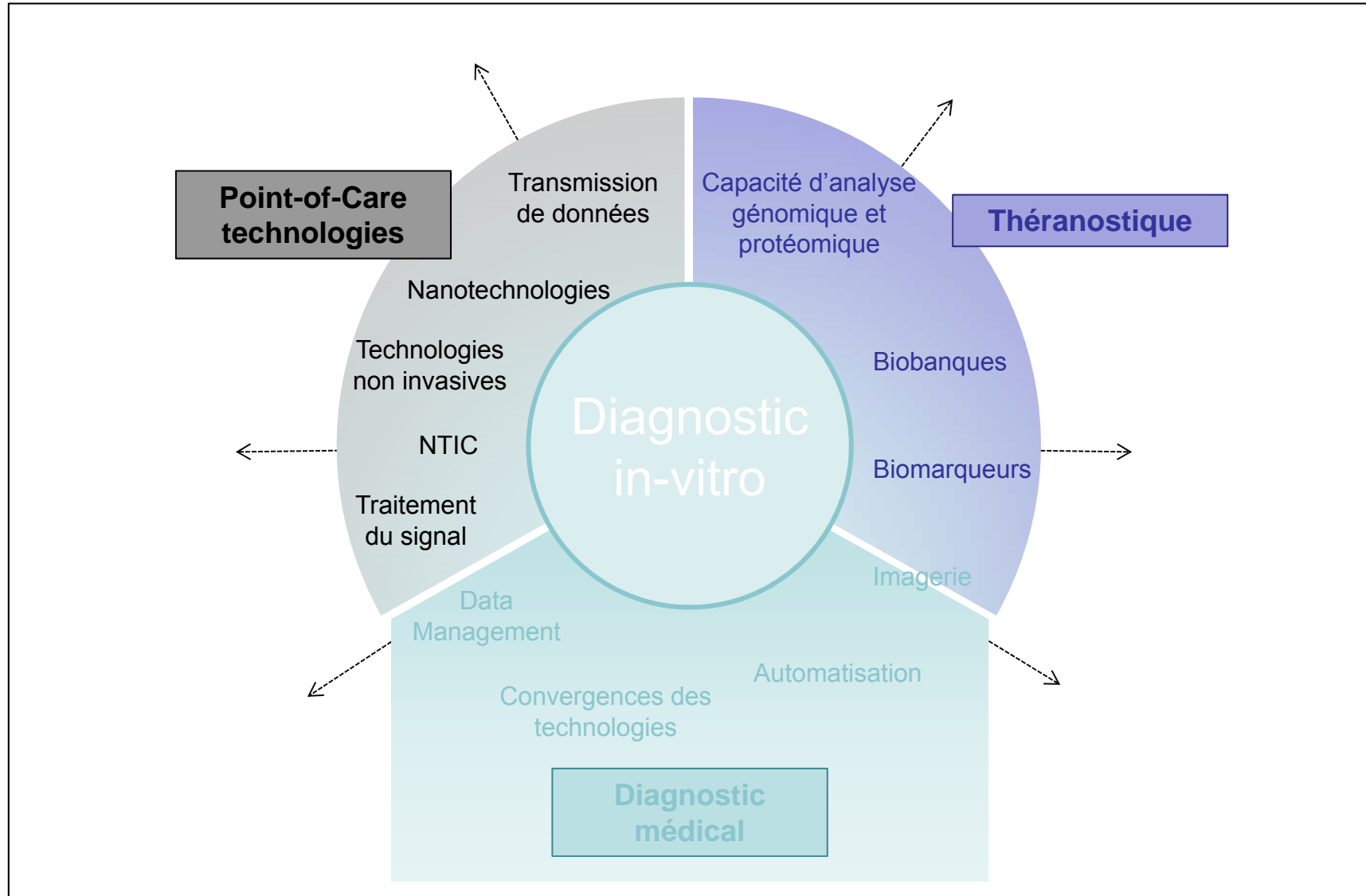
DIAG DISCOVERY

PRACTICIANS

DRUG DISCOVERY

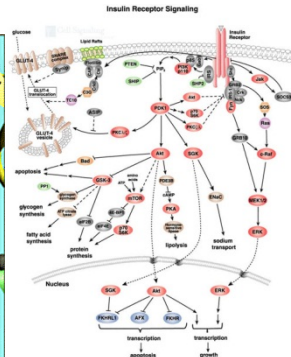
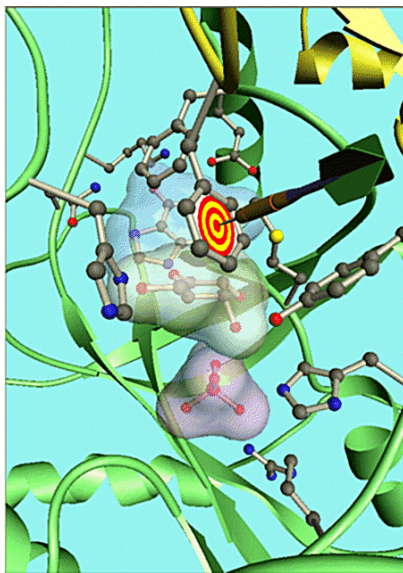


Various axes of development



Biomarkers

Parameters objectively measured (precise and reproducible), as an indicator of a biological process (physiological or pathological) or of a drug action.



Kinds of biomarkers :

Molecular (gene, RNA, Proteine, Chemical compound)

Physical

Imaging

others

Diagnostic biomarkers

Early detection biomarkers

Disease classification

Predictive biomarkers

Predict the response to a specific agent

Predict a particular adverse reaction

Metabolism biomarkers

Biomarkers that guide drug doses

Outcome biomarkers

Those that predict response

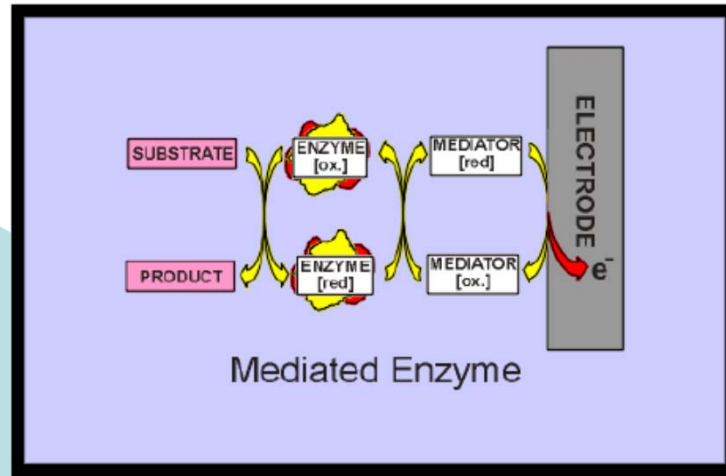
Those that predict progression

Those that forecast recurrence



Clinical question

Biomarkers



Undirect Cost

Medical

reliability
Mobility / miniaturization
Communication
control

Direct

Coupling Biology with electronics Biosensors

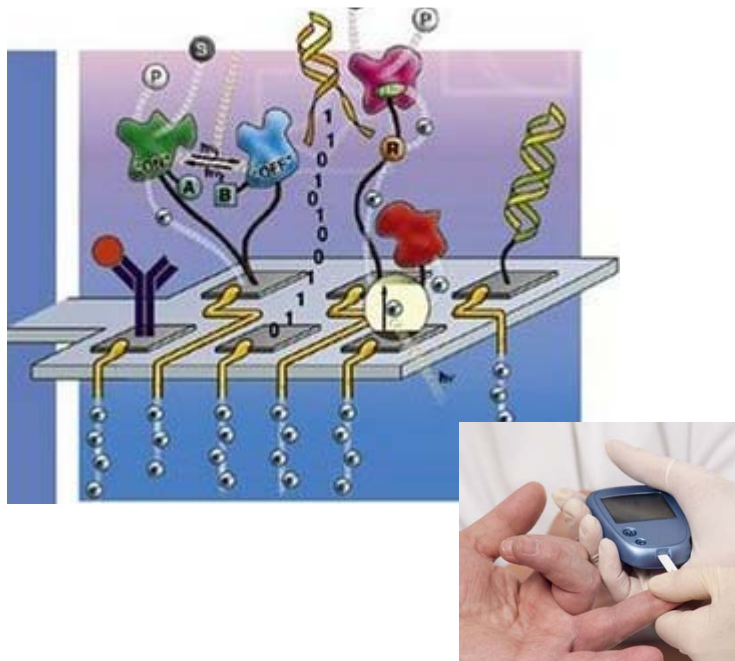


Screening
Diagnosis
Monitoring

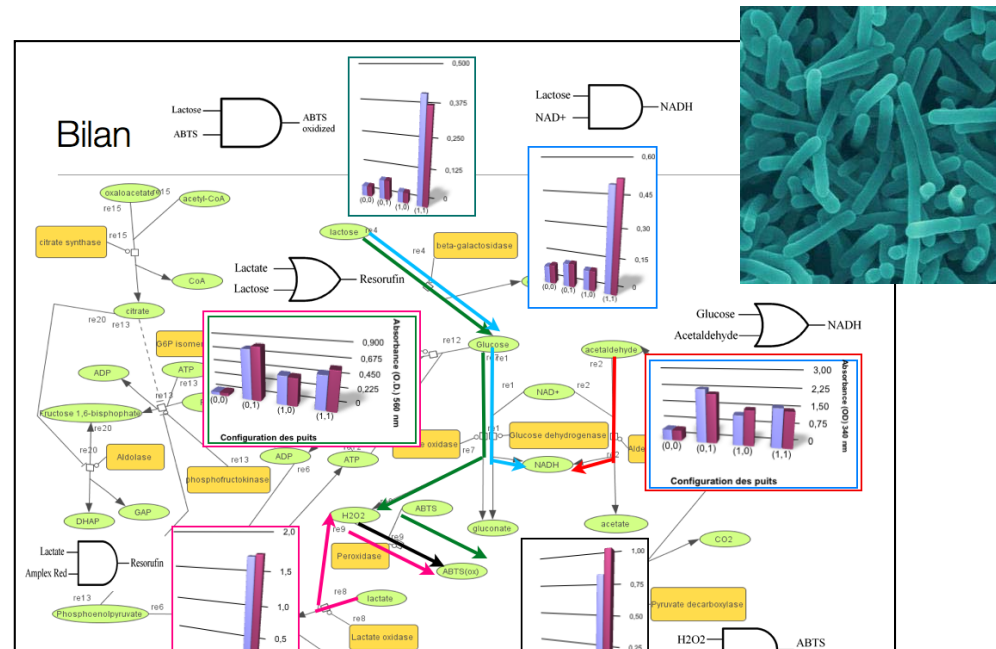


Engineered biology will transform diagnostic practices

From bioelectronics to artificial biological systems



Biosensor > Point of care and home testing



synthetic biology
Artificial biological systems



Systems Biology and Synthetic Biology

to address complexity

Systems biology

Integrative biology and modeling

To

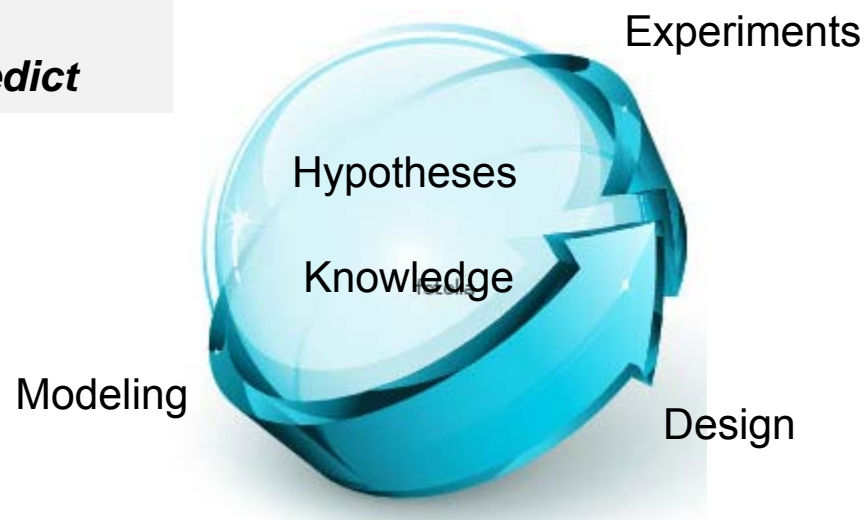
Understand and predict

Synthetic biology

Integrated prediction

To

Design





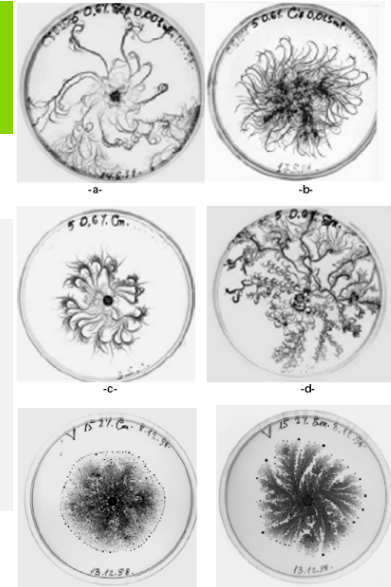
Synthetic Biology

- A) the design and construction of new biological parts, devices, and systems
- B) the re-design of existing, natural biological systems for useful purposes.

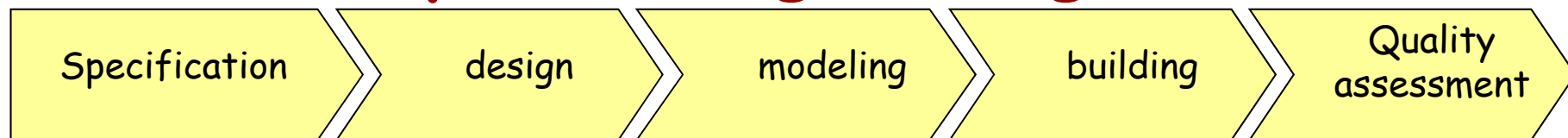
Zauner, 2006.

Biological molecules and biological systems are able of computer behavior in a robust and an auto-organized way

Bacteria Harnessing Complexity, E. Ben Jacob et al. 2002



Systems engineering



Standardization

Abstraction

Decoupling and modularity

Gulati et al., 2009

Endy, 2005

Synthetic Biology

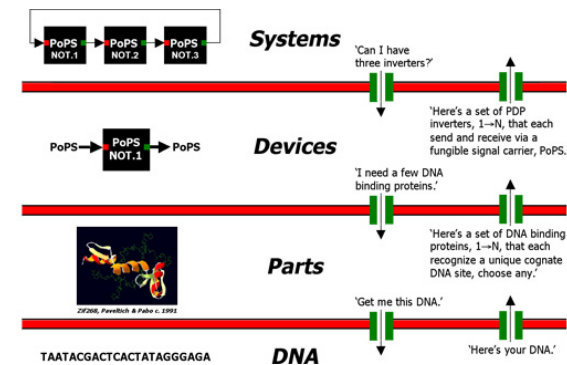
Abstraction :

Different hierarchical levels to manage complexity

'Parts'=Basic Biological Functions

'Devices'= Combination of parts to perform a desired function

'Systems'= Combination of 'Devices'



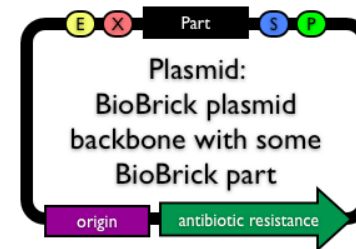
Standardization :

Concentrating information : description and characterization of components and conditions makes it easier for the user

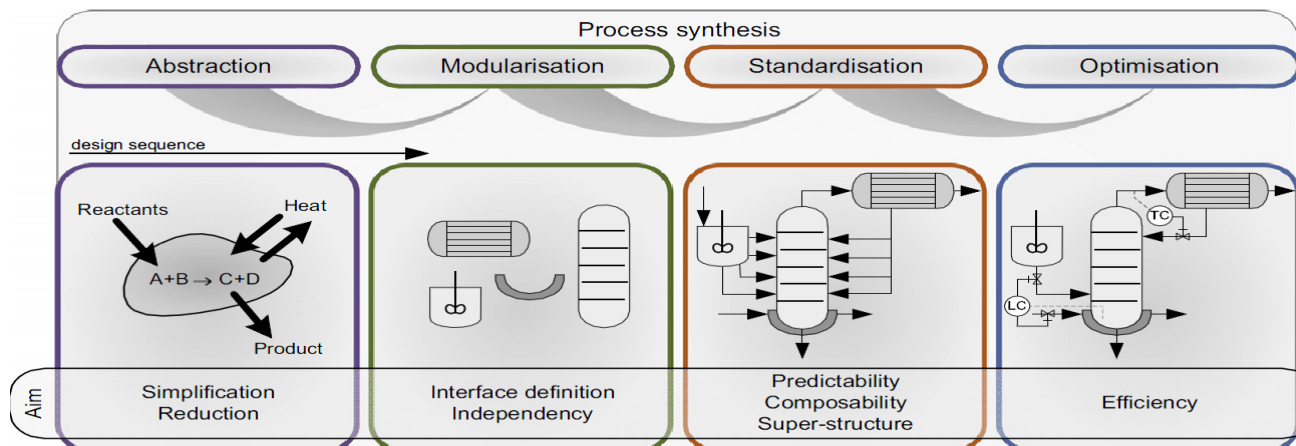
Decoupling and Modularity :

A complex problem is a set of tasks

A complex biological system is a set of separate devices

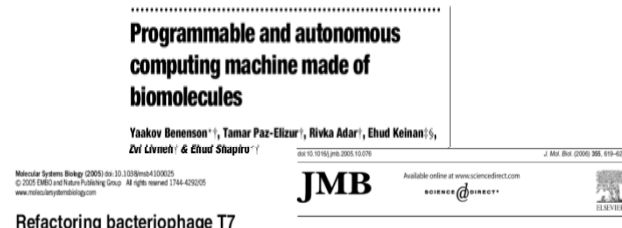


Catalog of parts & devices
partsregistry.org/





Various strategies in synthetic biology



Leon Y Char^{1,2}
¹ Department of Bio
Cambridge, MA, USA
² These authors con
* Corresponding aut
Tel.: +1 617 255 5

Ron Weiss, MIT, Harvard,
Design of a lentivirus able to target breast cancer intra-cellular biomarkers RNAi networks

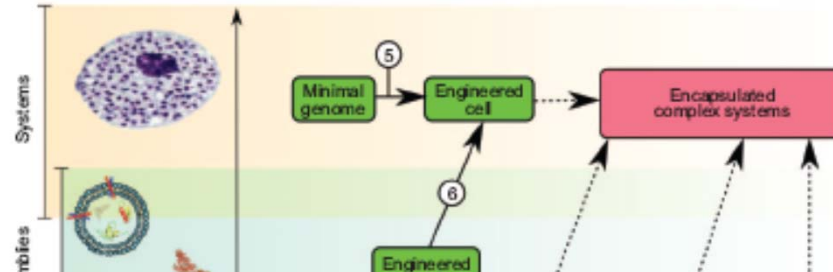
Stem cell reprogramming by bact. for tissus reconstruction

PL Luizi, Roma autocatalytic vesicules constructions, minimal cells.

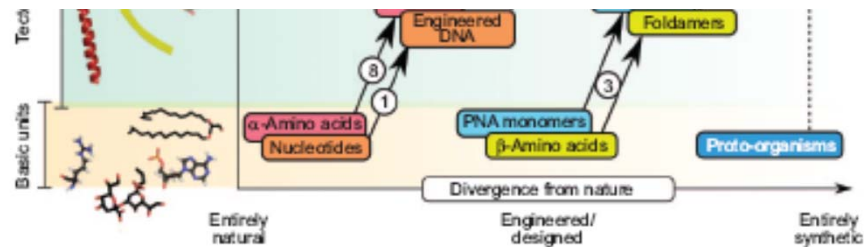
J. Stelling ETHZ, Zurich Electronic-like circuit design with composable parts (Bact.)

V. Dos santos, Helmutz Inst. Germany reprogrammed bact. to target cancer cell
Pseudomonas putida.

Jim Haseloff, Univ Cambridge UK , Plant reprogramming



DNA molec
both data
Yaakov Benenson^{1,2}, BI
Departments of ¹Computer Sci
Edited by Peter S. Davies, Cal





Synthetic Biology and Health : Proof of concept and applications

Drug production : metabolic engineering

Various "devices" have begun to emerge from Synthetic biology

- Orthogonal inducible promoters
- RBS libraries
- State sensors
- Spatiotemporal controllers...

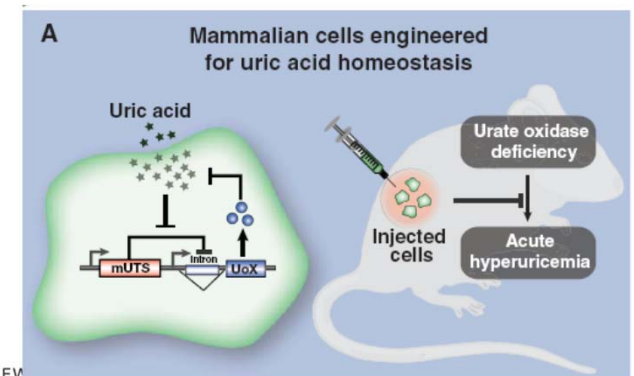
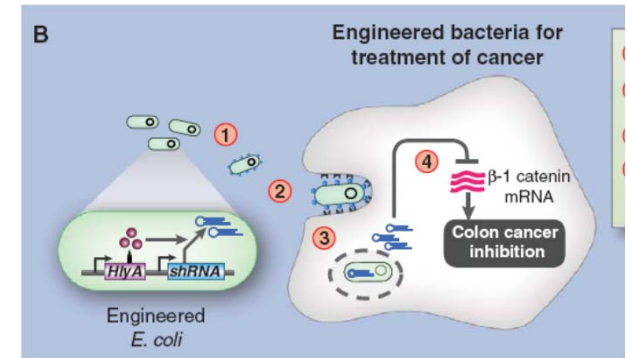
Synthetic biology devices as tools for metabolic engineering Eric Shiue

Engineering of "smart" cell for therapeutic purpose

- Bacteria invading cancer cells in vitro
- Delivering a functional RNAi
- Control the virulence of pathogens
- Eliminate cancer cells based on their expression pattern of micro-RNAs

Medical diagnostics :

Limited number of project



REVIEW

Synthetic Biology Moving into the Clinic

Warren C. Ruder,* Ting Lu,* James J. Collins†

Ruder et al Science 2011



What synthetic biology can bring to clinical diagnostic ?

Possibility for local molecular measurements

high sensibility, Early diagnosis
Accessibility to difficult region, less invasive

Perturbations from local environment
Specificity, accessibility issues

Capabilities for integrated processing

Reduce need of heavy technology
low cost compatible to systematic screening,
personnalized medicine etc.
Ability for sophisticated assays Supported by
Modelling

Composability issues
Control and robustness problems
Complex experimental and Clinical validation

Biocompatibility

can be disposable
human compatible

Stability issues,
Human and environmental interaction issues

Can be interfaced with other supports

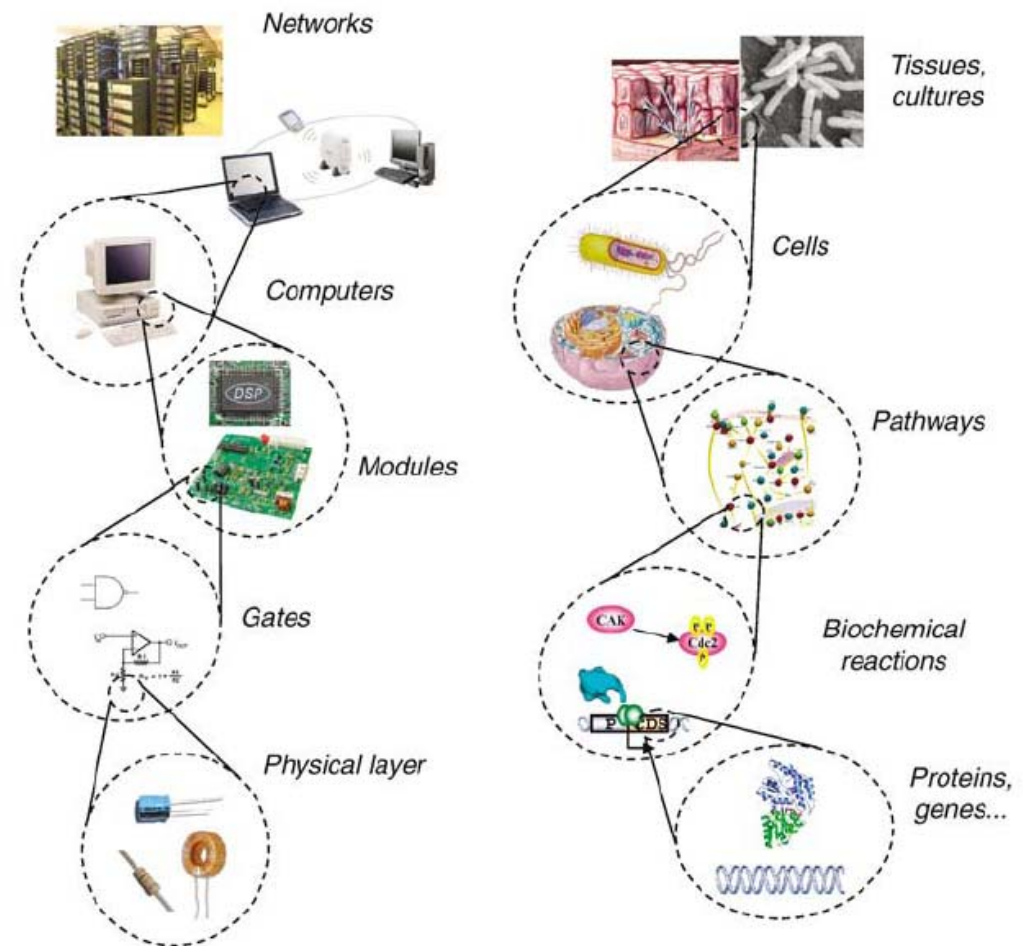
Electronics, physics,
chemistry, automation, etc.

Linearity, Kinetics,
normalization, etc.



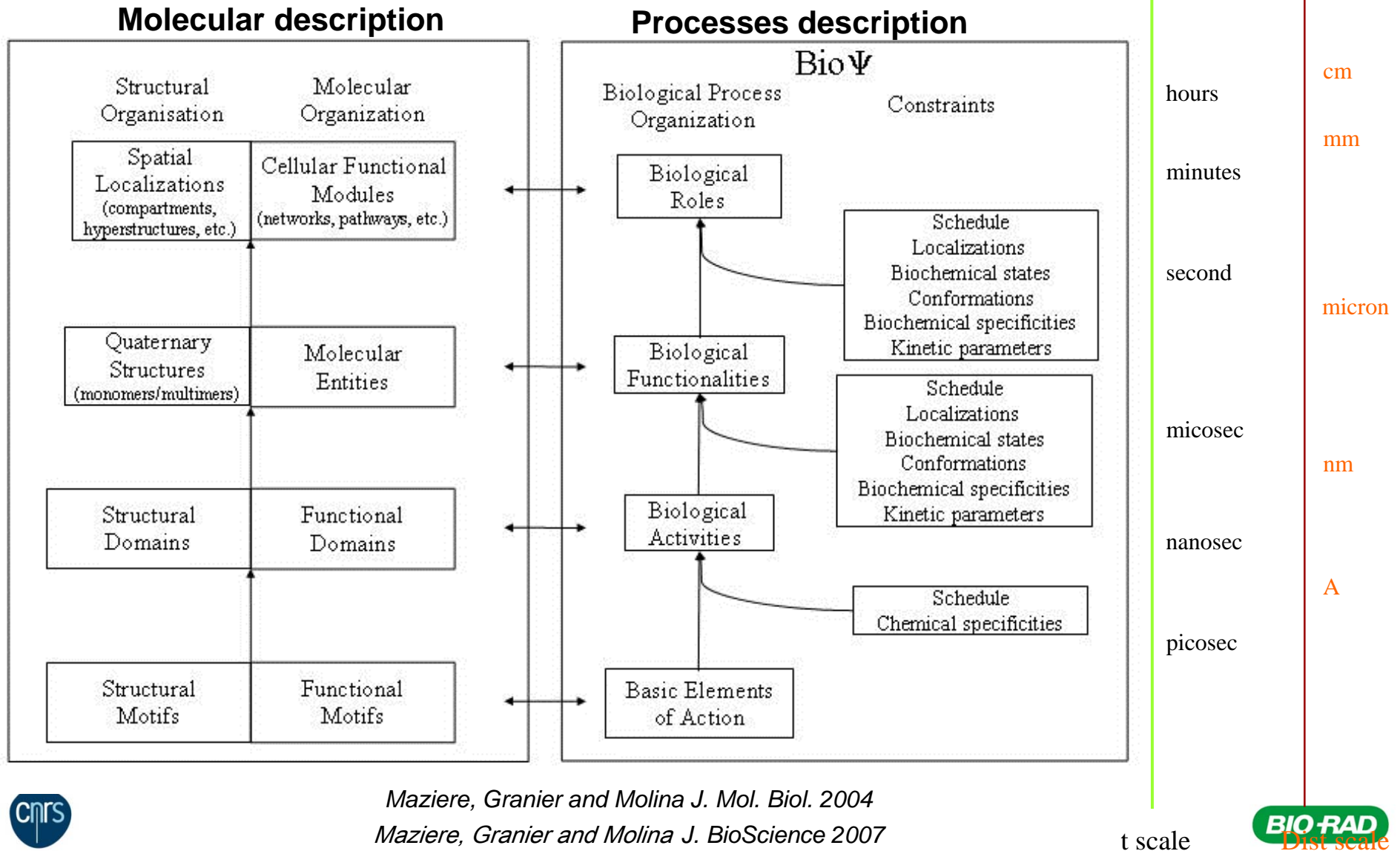
Design synthetic biological system based on functional parts

The modeling issue





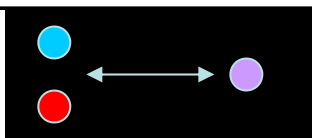
Biological processes are made by combination of a limited number of elementary actions





A
Bond Modifiers

- a split/link
- a acting on C-C
- b acting on C-O
- c acting on C-N
- d acting on C-S
- e others



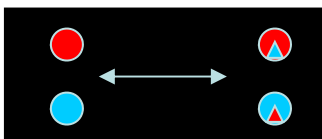
B

Transferts

- a Transferors
- a acting on C-C
- b acting on C-O
- c acting on C-N
- d acting on C-S
- e others

b Oxidoreductors

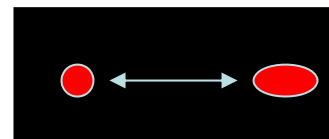
- a acting on C-C
- b acting on C-O
- c acting on C-N
- d acting on C-S
- e acting on S
- f acting on N-O
- g acting on S-O
- h others



C

Intramolecular modifications

- a Isomerors
- a chirality
- b cis/trans
- c bond moves
- d others



Classification of 97 basic elements of actions BEA (for all known processes)

D

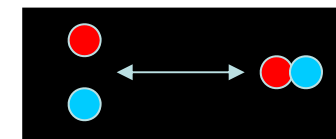
Non covalent interactions

a Binding:

- a Protein-Protein
- b Protein-Nucleic Acid
- c Protein-Other
- d Nucleic Acids-Nucleic Acids
- e Nucleic Acid-Others

b Transport:

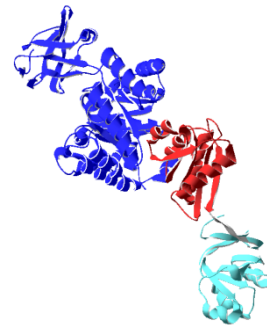
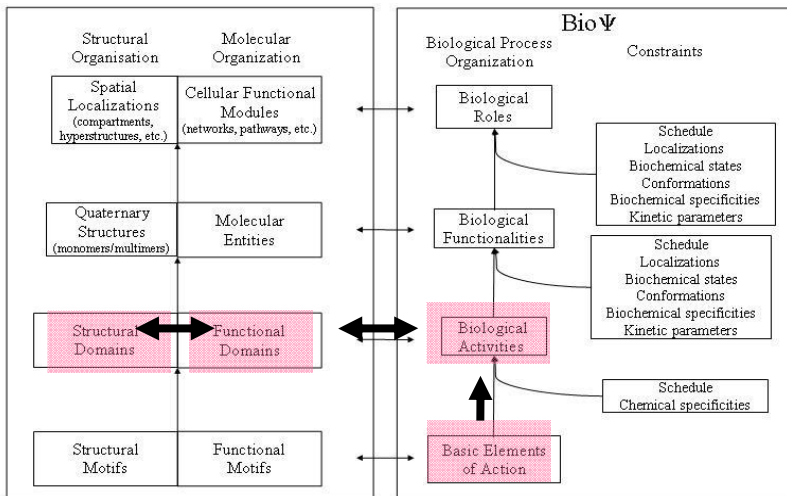
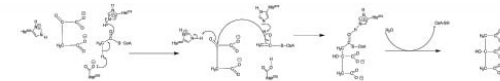
- a Tunnel
- b Cargo





Formal structure-function description based on elementary actions

- Enzymatic activity on metabolites relies on chemical processes.
- New** Bio Ψ formalism allows describing detailed elementary actions (BEA) at the chemical level



Bio Ψ description of CS

$BF_{CS}(ACoA, OAA, H_2O) =$
 $BA_{Bind}(OAA, CS),$
 $BA_{Bind}(ACoA, CS),$
 $BA_{thioesterhydrolysis}(ACoA, OAA, H_2O),$
 $BA_{Bind}^{-1}(CoASH, citrate)$

Set of BEA

Da:misc.1,
 Da:misc.1,
 Ba:CO.1⁻¹,
 Aa:CC.2⁻¹,
 Ba:lab.1,
 Ba:lab.2⁻¹,
 Ba:CS.2,
 Da:misc.1⁻¹ and
 Da:misc.1⁻¹

Aa:CC.2(R1,R2,R3): R1-C(OH)R2-R3 \rightarrow R1-CO-R2 + R3-H
 Ba:CO.1(R): C-O-R \rightarrow C^o + R-O^o

12 SysDiag

Bio Ψ DataBase

- Allows the Bio Ψ description
- Contains all the relationship between EC number and BEA
- Will contains the relationship between folds and BEA

http://www.sysdiag.cnrs.fr/publications/supplementary-materials/BioPsi_Manager/

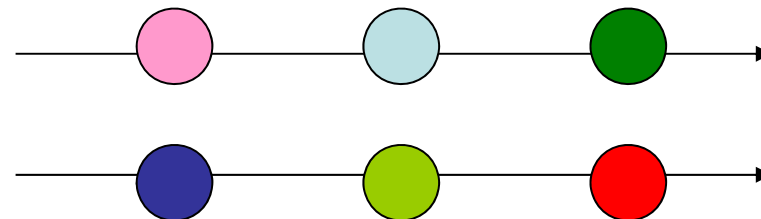
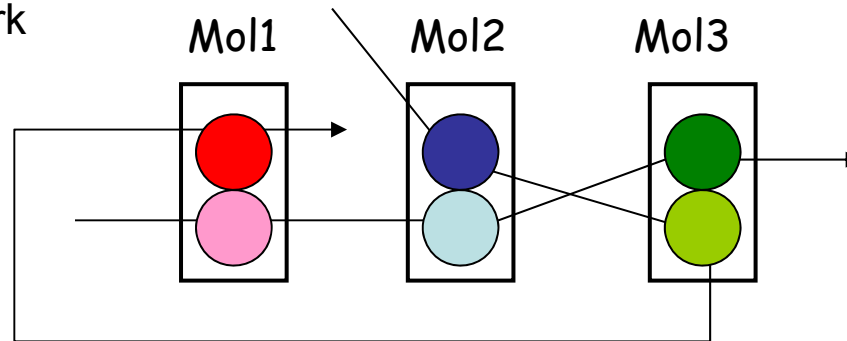
A functional unit has a sequence of BEA

Bio Ψ better describes functions than GO, EC
Bio Ψ formalism allows calculating on functions



Paradigm shift for network modeling

Molecule based network



Functional unit based network

Functional units are defined by 3D fold and BEA sequences

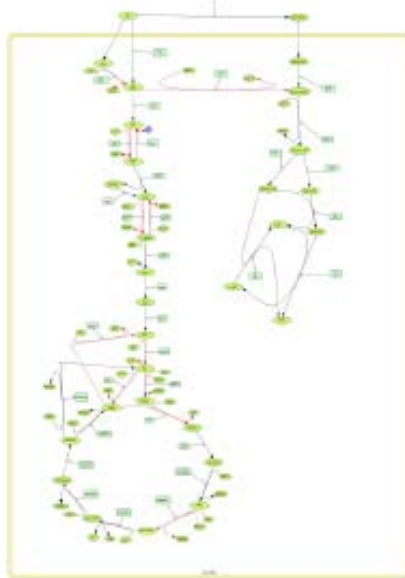
Pérès et al Bioinformatics 2010, Buesher, et al. Science 2012



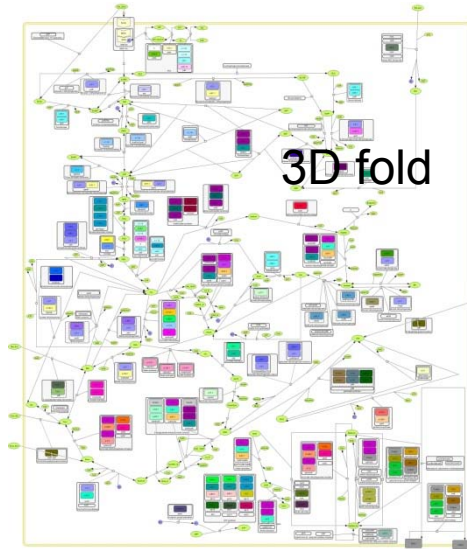
New paradigm for functional network representation

Consequences on :

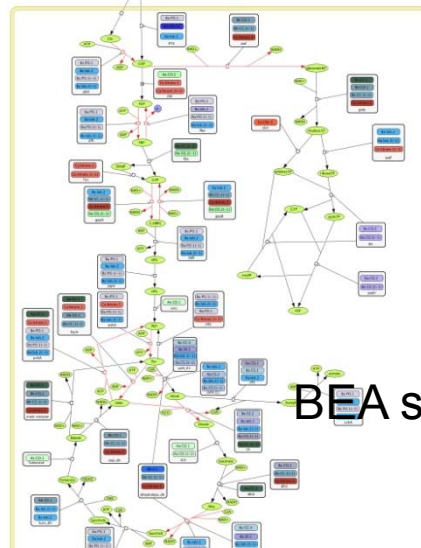
- Network topology analyzes
- Dynamic network simulation
- Functional interpretation



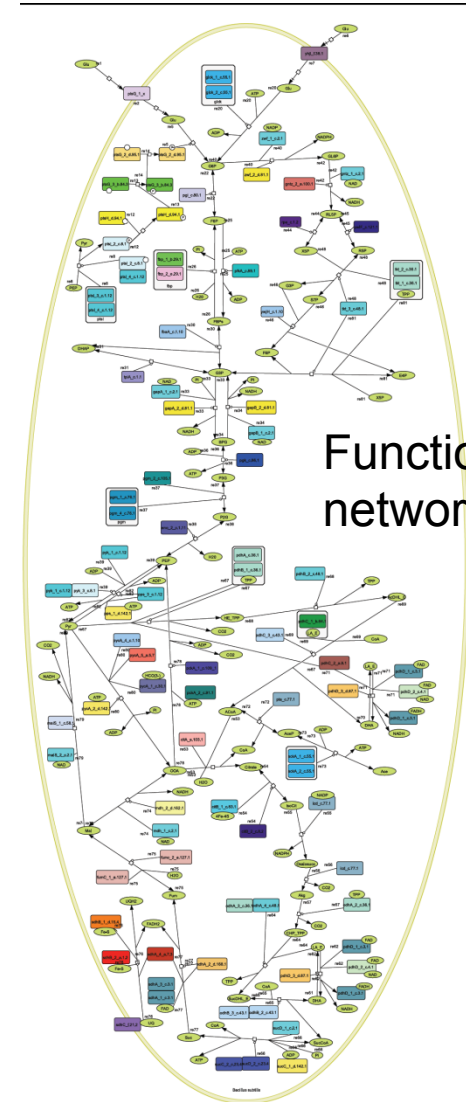
Molecular
Network
CCM *B. subtilis*



3D fold



BEA seq

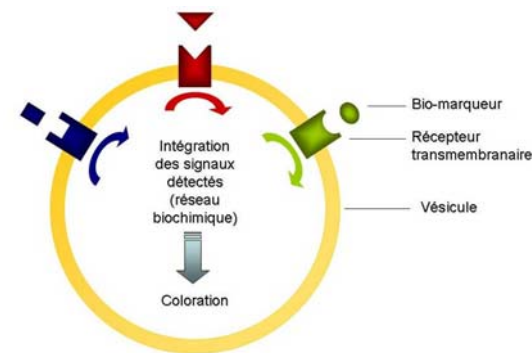


Functional
network



Design a cell-free synthetic biological system based on functional parts :

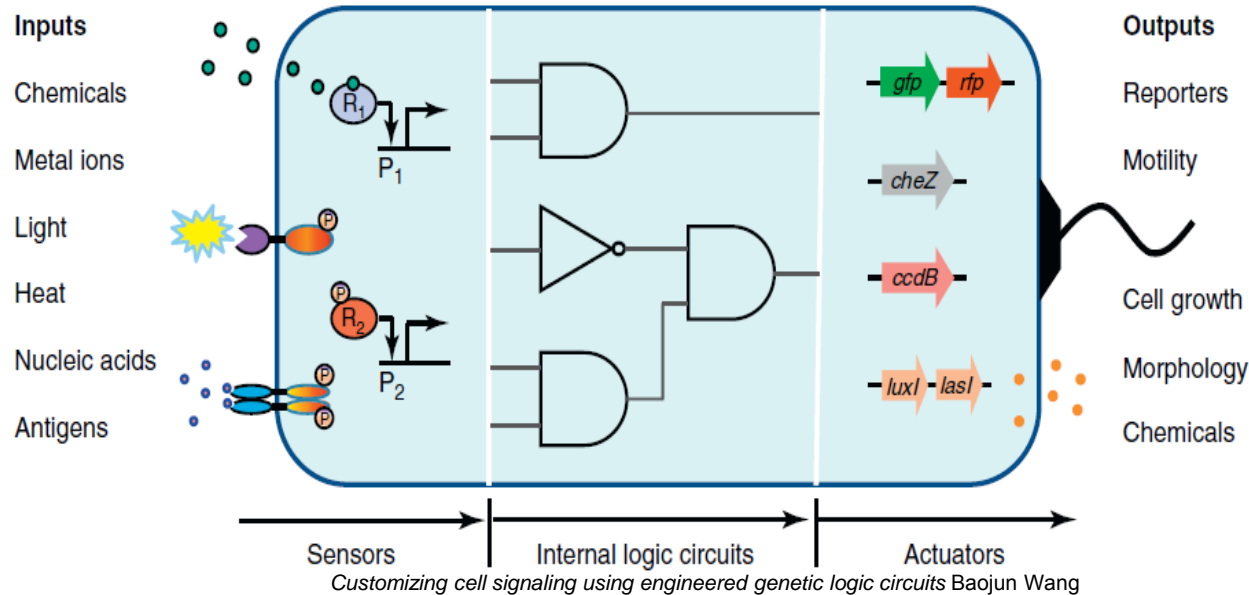
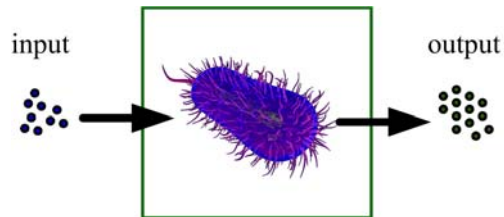
The composability issue





Synthetic Biology and Whole-Cell Biosensors

Whole-cell biosensors are intrinsically modular : consisting of a recognition element coupled to an arbitrarily chosen reporter



Advantages:

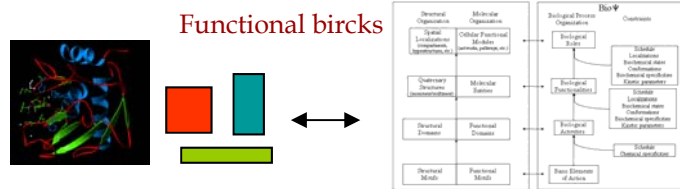
- Simple use
- Cost-effective
- Self-replication
- Fast, short analysis time
- Disposable
- High sensibility (signal amplification) and selectivity
- No need for sample preparation
- Multiplexing



Our Strategy

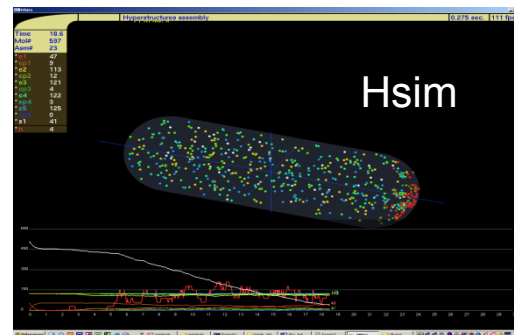
Design - Simulation/modélisation - Experimental validation

Identification and characterization of molecular compounds

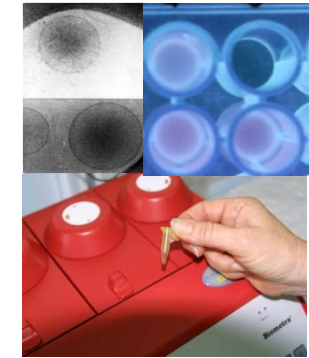
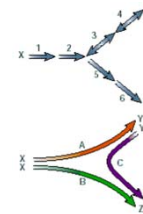


BioΨ
Formalization

Simulation
Stochastic Cellular automaton, Multi-Agent



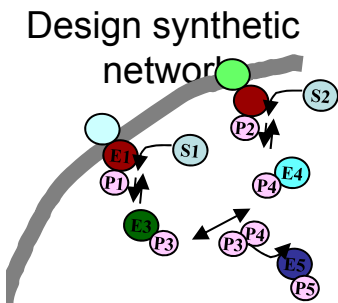
Modelling:
Control Th.
flux analyses
ODE
Elementary modes



Biotechnology

Auto-organisation/
Robustness

Validation *in vitro*



BioNetCAD network Computer Assisted Design

Rialle, S., Felicori, L., Dias-Lopes, C., Peres, S., El Atia, S., Thierry, A. R., Molina, F. (2010). BioNetCAD: Design, simulation and experimental validation of synthetic biochemical networks. *Bioinformatics*, 26(18):2298-304



Standardized catalog of functional biological compounds:

Reusable, non toxic

- ↪ Processes are formalized (ready for modelling)
- ↪ Biological behavior characterized experimentally (ready to use in a synthetic system)
 - × « compound » properties
 - × robustness
 - × stability
 - × « composability » score

```

Generic BioPsi description
Biological Roles description
BFEnzyme-LinkedTransmembraneConditionalTest :=
  BFTransmembraneReceptor(Ligand) IM transmembrane |
  BFLigand(TransmembraneReceptor) IM extracellular,
  BFTransmembraneReceptorLigand(Substrate) IM transmembrane |
  BFSubstrate(TransmembraneReceptor/Ligand) IM cytosol,
  BFProduct IM cytosol

Biological Functionalities description
BFTransmembraneReceptor(Ligand) :=
  BA_Binding(Ligand)
  IM extracellular

BFTransmembraneReceptorLigand(Substrate) :=
  BA_ConformationalChange,
  BA_EnzymaticActivity(Substrate) -> Product
  IM cytosol

BFLigand(TransmembraneReceptor) :=
  BA_ProteinBinding(TransmembraneReceptor)
  IM extracellular

BFSubstrate(TransmembraneReceptor/Ligand) :=
  BA_ProteinBinding(TransmembraneReceptor)
  IM cytosol

BFProduct
  BA_ProteinBinding-1(TransmembraneReceptor)
  IM cytosol
    
```

SysDiag UMR3145 Compubiotech project [Contact](#)

Reusable Molecular Elements Catalog

Catalog home | References | External links

Catalog Home

Actions at the Compartmental environment scale

- redox conditions keeper
- pH conditions keeper
- temperature conditions keeper

Actions at the System scale

- timekeeper
- switch
- killer
- scaffold controller
- fuel

Actions at the Module scale

- conditional sensor
- revealing
- amplifier
- inhibitor
- distributor
- revealing
- cargo
- conductor

Actions at the Molecule scale

Welcome to the Molecular Elements for Synthetic Biology Catalog website

Introduction

Devices can act at different levels in the synthetic system :

- The whole system scale
- A group of molecules scale
- A single molecule scale

SysDiag UMR3145 Compubiotech project [Contact](#)

Reusable Molecular Elements Catalog

Catalog home | References | External links

Catalog > Module scale > Conditional Sensor (CS) > Enzymatic Transmembrane CS

1 TransMembrane CS

1.1 Enzymatic TransMembrane CS

An enzymatic transmembrane sensor is a receptor, where the binding of an extracellular ligand leads to an enzymatic activity of the receptor himself, in the intracellular side.

List of BioPsi processes used in the current module

Biological Role

- BR_TransmembraneConditionalTest

Biological Functionalities

- BF_TransmembraneReceptor
- BF_Ligand



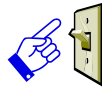
Molecular, modular compounds Identification, characterization

- Modules with defined role

Redox conditions monitoring

pH conditions monitoring

Temperature conditions
monitoring



Timekeeper (Time
counter)

Switch On/Off

Killer (destruction
component)

Scaffold controller

Fuel



Conditional Sensor

Amplifier

Inhibitor

Distributor

Revealing

Cargo

Conductor

- Proteins and small molecules useful for synthetic biology

Ex : Peroxidase + substrat



Revealing role

Ex : Glucose oxydase + glucose



Switch on role, Conditional sensor role

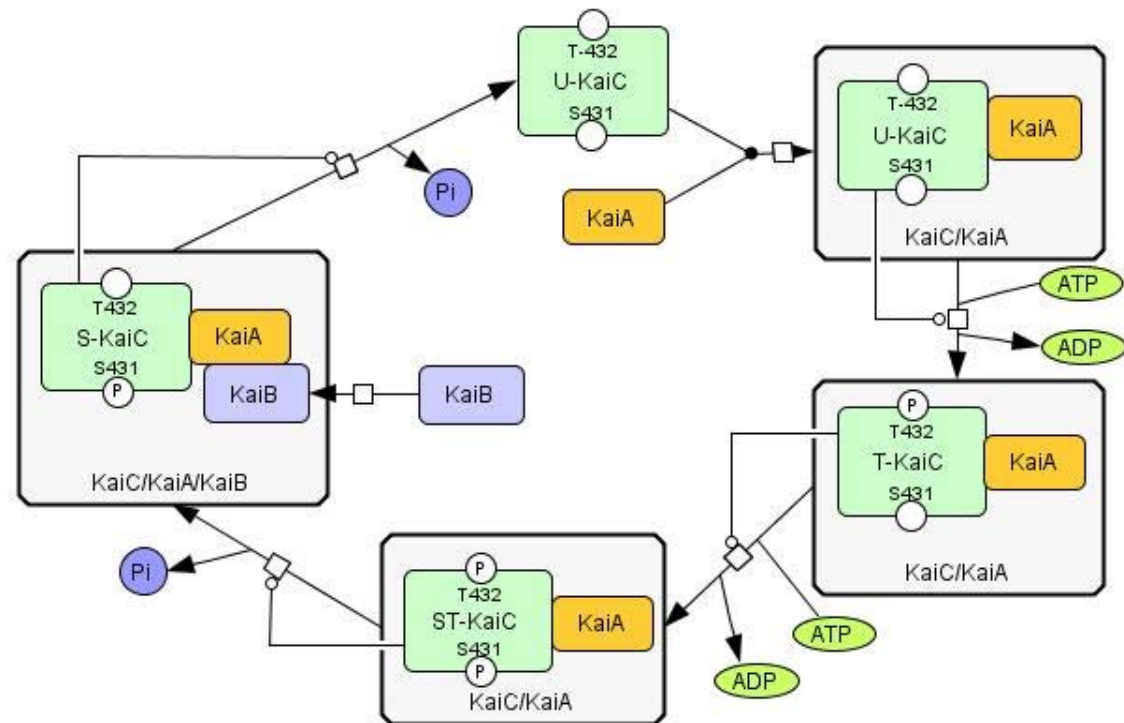


Oscillator

Type 1: Circadian oscillation

Implementation 1: Cyanobacterial KaiC phosphorylation

- 3 protéines : KaiA, KaiB, KaiC
- Phosphorylation de KaiC oscillante (période = 24h)



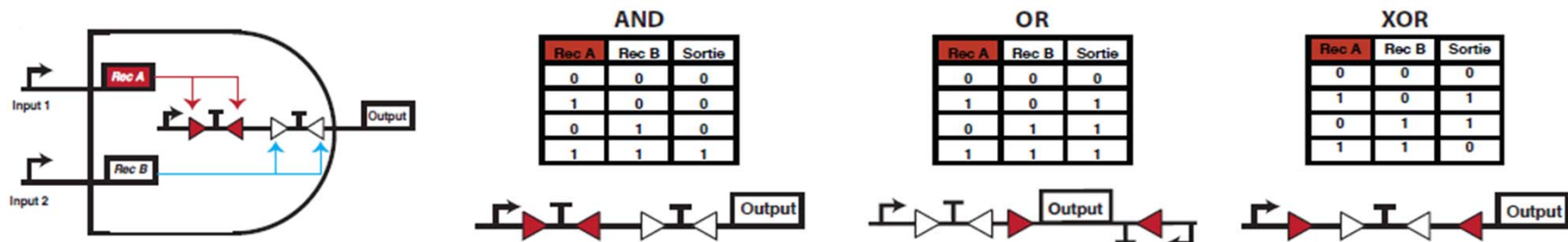


Rewritable digital data storage in live cells via engineered control of recombination directionality

Jerome Bonnet, Pakpoom Subsoontorn, and Drew Endy¹

Department of Bioengineering, Room 269B, Y2E2 Building, 473 Via Ortega, Stanford University, Stanford, CA 94305

First reliable and rewritable DNA inversion-based data storage system that works in vivo



Recombinase Logic gates memory offers multiples advantages :

- i. Transient signals of low intensity can be stored
- ii. Able to perform sequential logic
- iii. The result of the diagnostic test may be stored in ADN and read by different methods

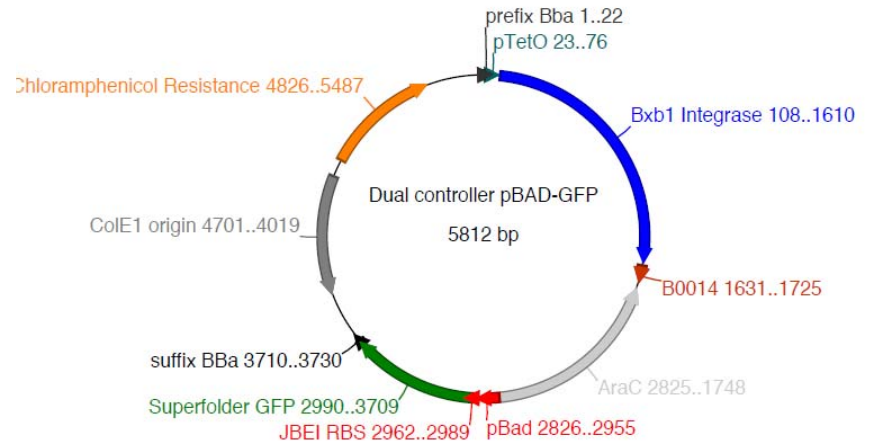
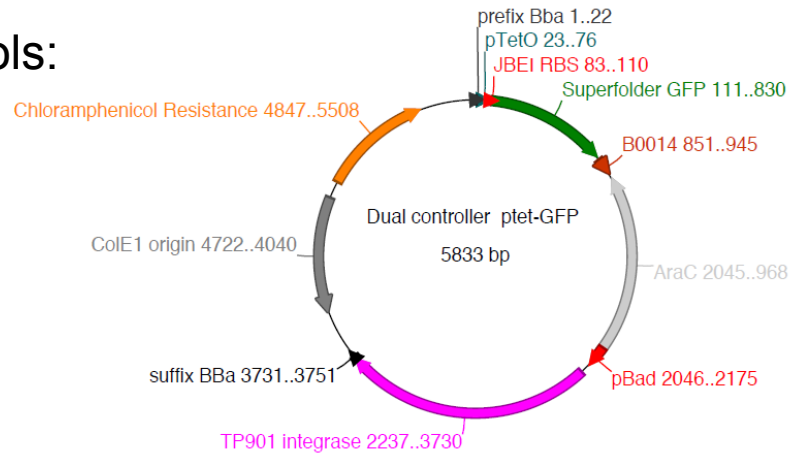
Control plasmids expressing two serine integrases adapted from bacteriophage TP901 and Bxb1 under the control of exogenous arabinose (ara) and anhydrotetracycline (aTc)

Plasmids encoding AND, OR, XOR, NAND, NOR, and XNOR logic elements placed between a standardized strong prokaryotic promoter and a GFP expression cassette

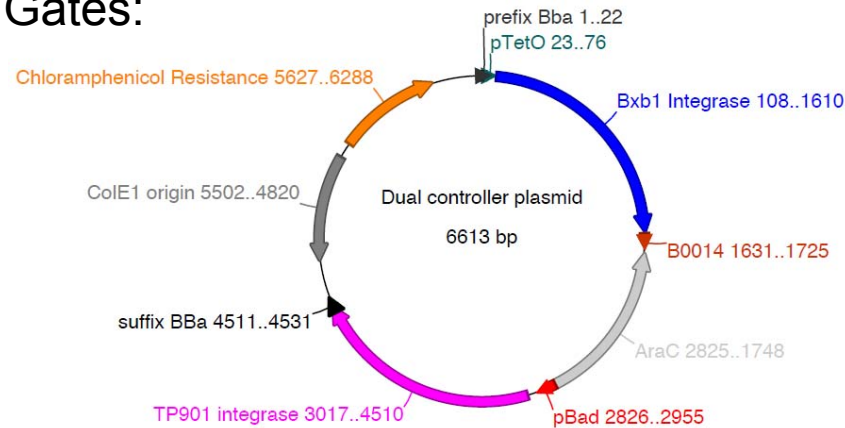


Biosensor in clinical context : testing function in samples

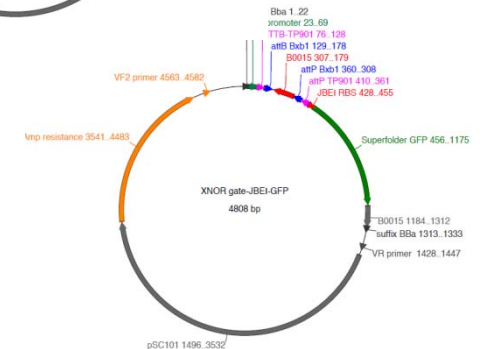
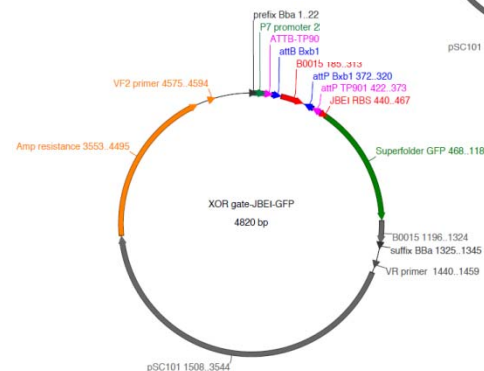
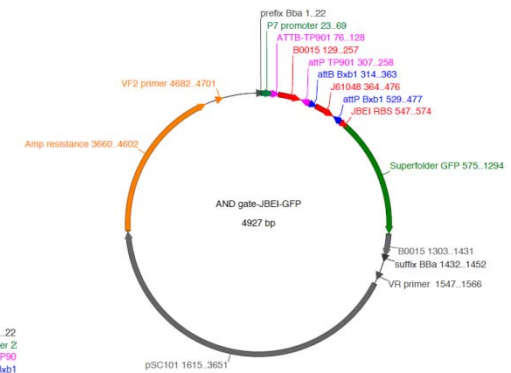
Controls:



Gates:



+

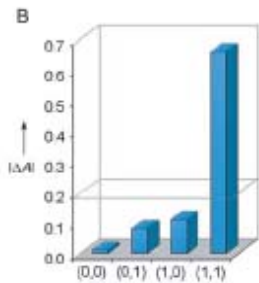
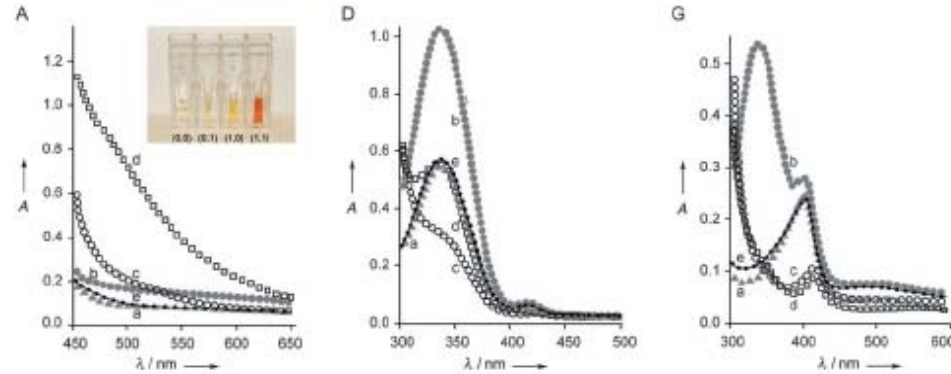


- >Monitoring of logic through :
- Bulk GFP levels from bacterial cultures
 - Single cell fluorescence distributions

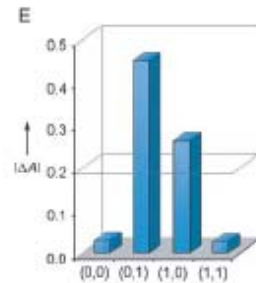


Biological logic gates identification (AND, OR, XOR, etc.) In cell-free biological nextworks

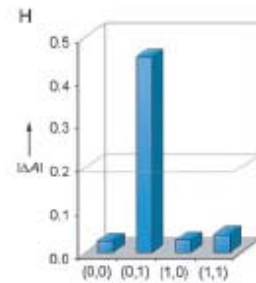
It is possible, Willner et al. 2006, 2009



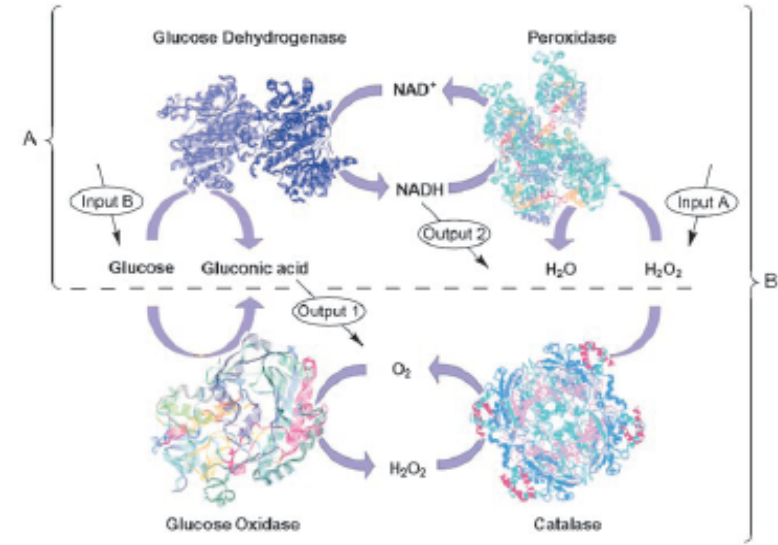
Input A	Input B	Output AND
0	0	0
0	1	0
1	0	0
1	1	1



Input A	Input B	Output XOR
0	0	0
0	1	1
1	0	1
1	1	0



Input A	Input B	Output InhibAND
0	0	0
0	1	1
1	0	0
1	1	0

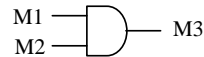
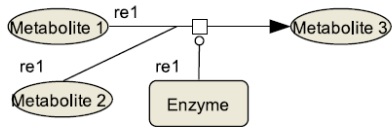


Scheme 1. A) Logic gates based on two coupled enzymes. B) Half-adder based on four coupled biocatalysts.



Rules for Bio-logic gates devoted to diagnosis use

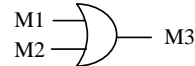
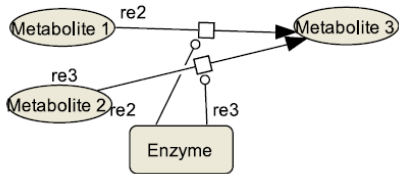
A



M1	M2	M3
0	0	0
1	0	0
0	1	0
1	1	1

Schematic representation corresponding to the different ways of obtaining each type of logic gate in a biochemical reaction.

B

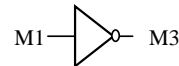
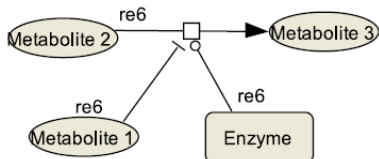


M1	M2	M3
0	0	0
1	0	1
0	1	1
1	1	1

(A) An enzymatic AND can be defined as two necessary and different metabolites (the inputs) lead to a product (the output) through one or several enzymatic reactions.

(B) An enzymatic OR gate could be defined by two different metabolites (inputs) that can individually produce (through one or several enzymatic reactions) the same metabolite (output) in identical external conditions.

C

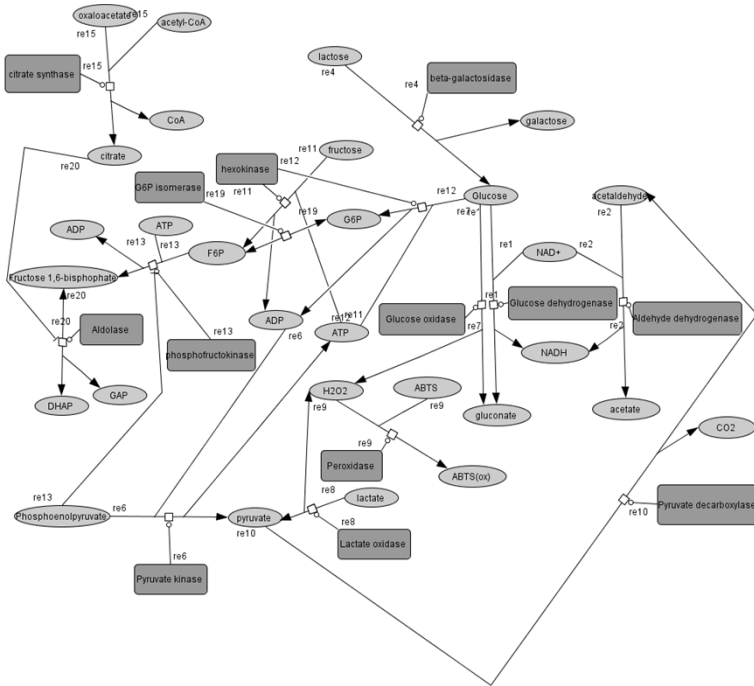


M1	M3
0	1
1	0

(C) An enzymatic NOT could be defined by a metabolite (input, an inhibitor typically) that prevents the production of another metabolite (the output).



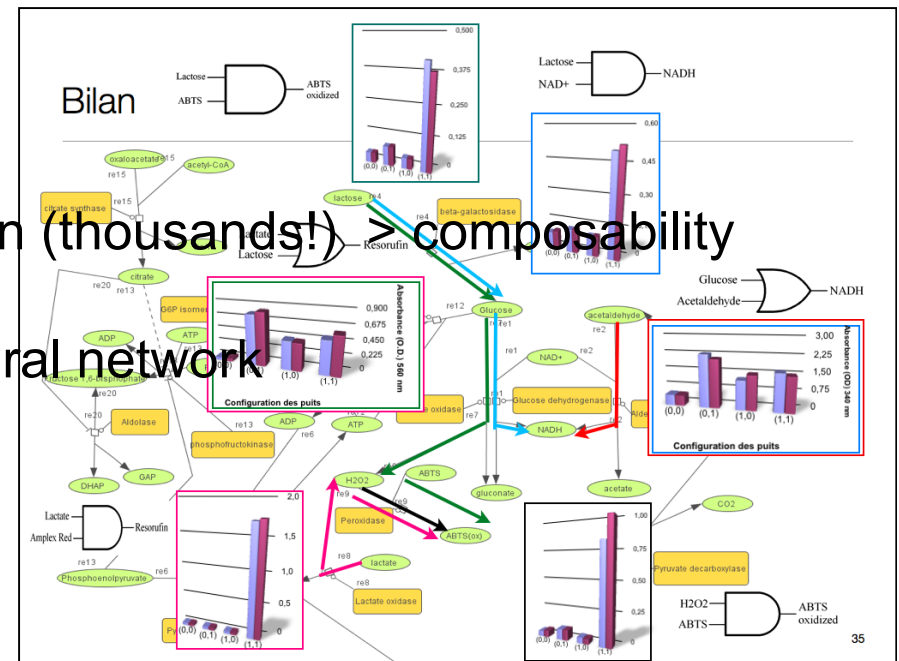
Natural biological network

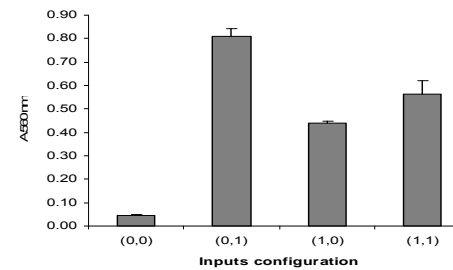
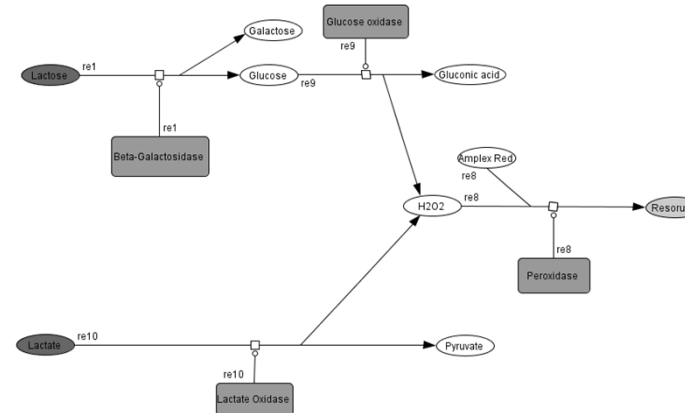
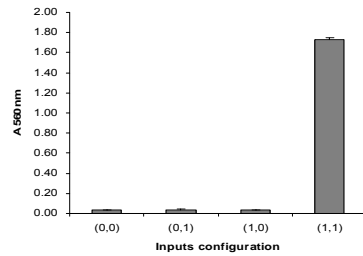
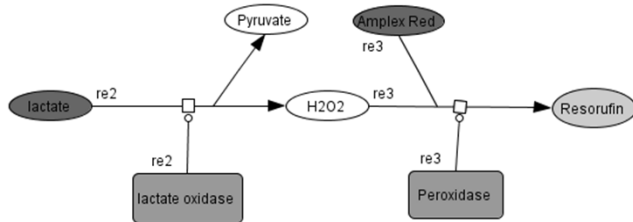
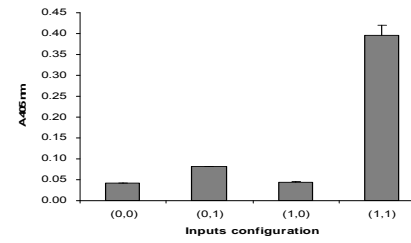
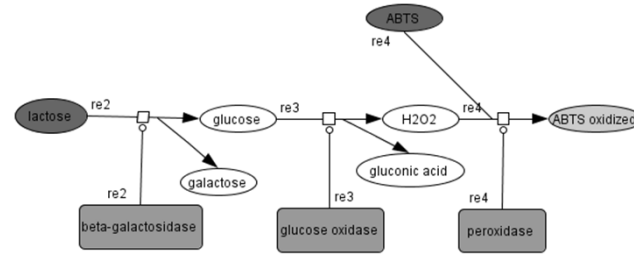
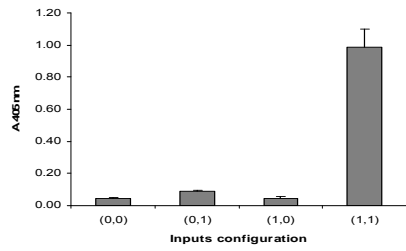
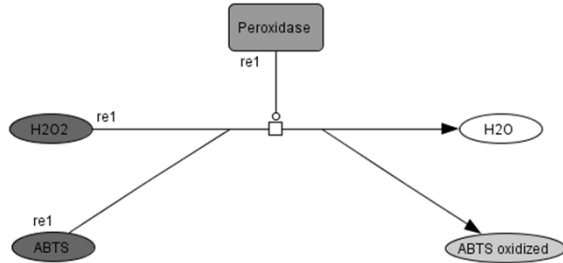


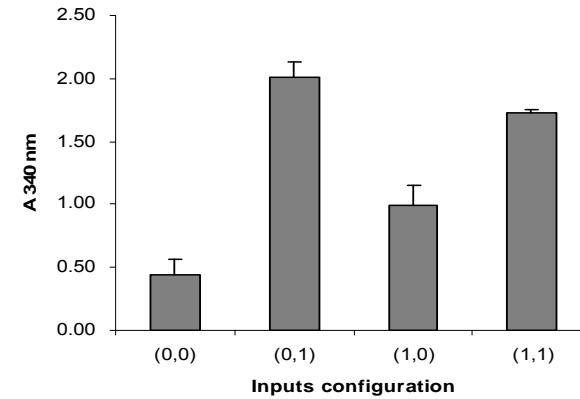
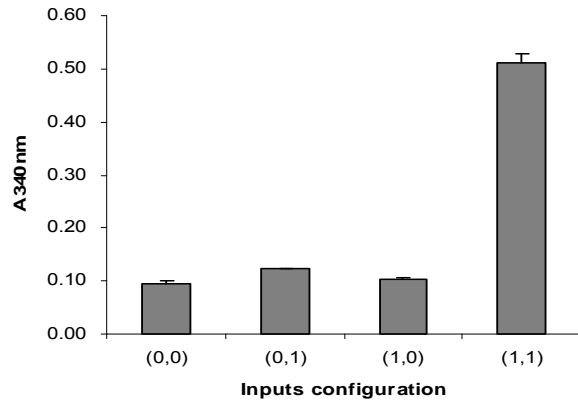
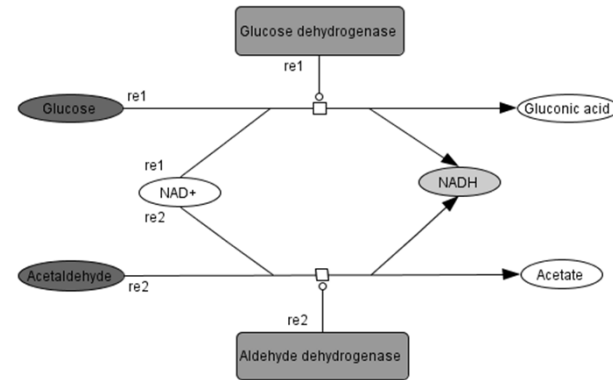
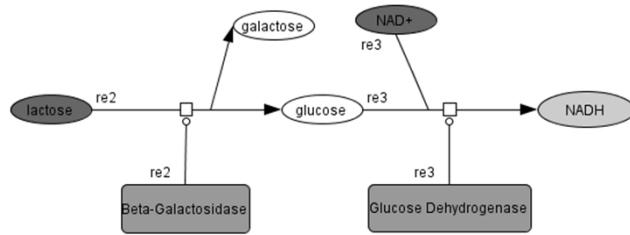
-Network formal description
 -EFM-like algorithm and Graph analyzes
 -Selected set of pattern satisfying RULES

Method for large scale bio-logic gate design (thousands!) > composability
 Compliant with Medical use (> safety)
 Study of Bio-logic gates distribution in natural network

Set of validated bio-logic gates





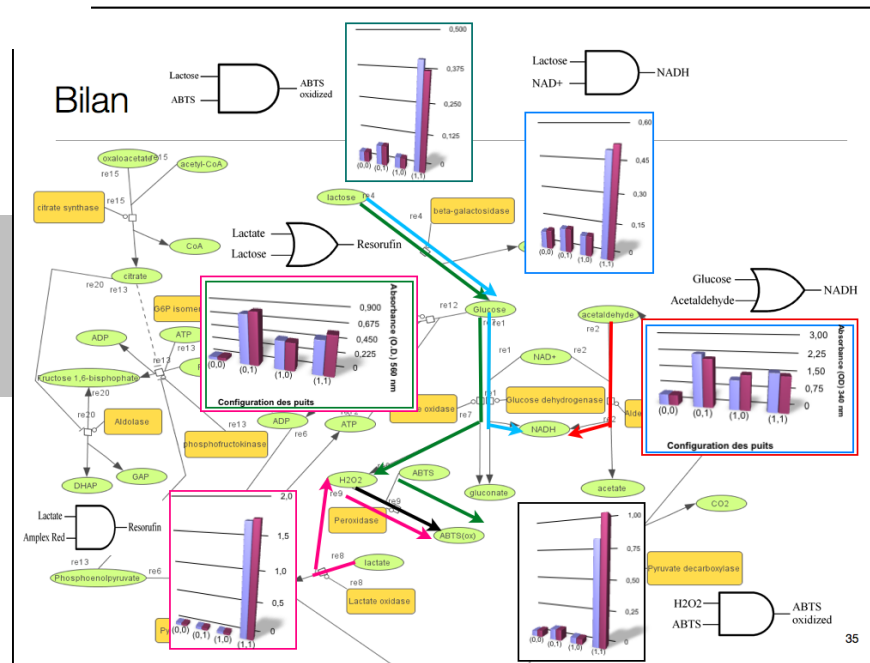




Bio-Logic gates for diagnosis

- Rules, method and tool to bio-logic gates identification in existing networks
- Study of logic properties of natural networks
- Experimental Validation

Felicori et al in preparation

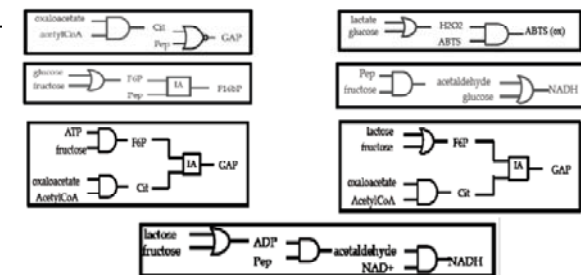


Set of experimentally validated Bio-logic gates. Robustness studies

Bio-logic gates Concatenation

Set of rules for bio-logic gates suitable for diagnostic use

Systematic approach to identify Bio-logic gates in biological networks EFM, graph





General principles for a simple biomarker sensing systems

Conditional sensing

Receptors,
biosensors etc.

Signal transduction

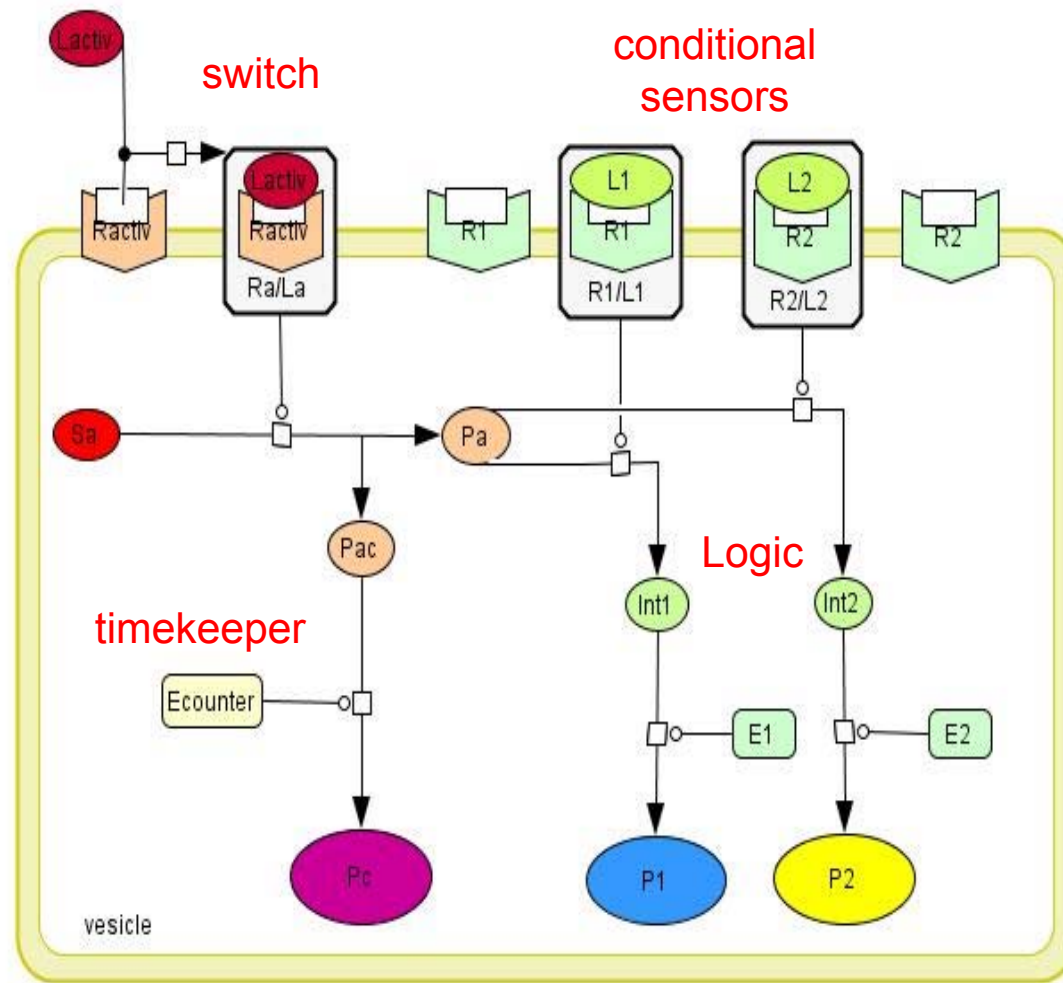
Integration/Control

Logical processing
Algorithms
Amplification,
normalization,
etc.

Signal transduction

Revelation

Coloration,
fluorescence, contrast,
etc.

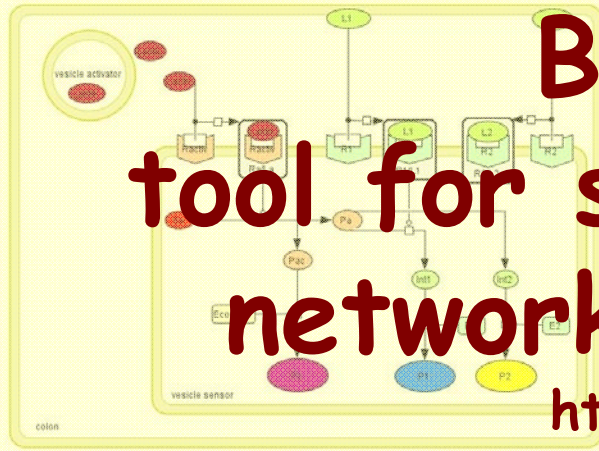


Reading



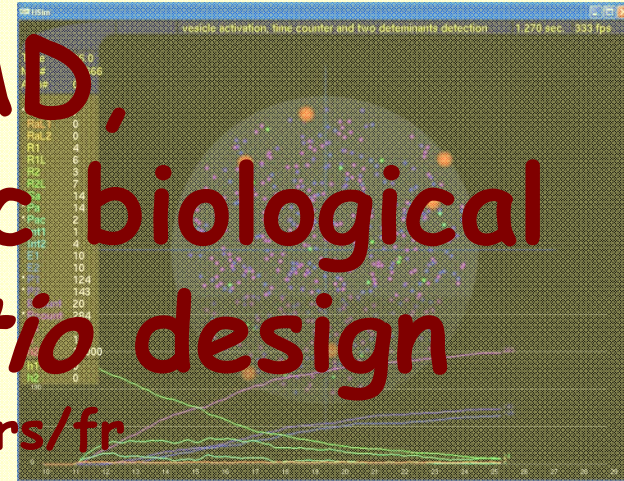
ab initio Artificial network design

from abstract network
to implemented network



BioNetCAD, tool for synthetic biological network *ab initio* design

<http://sysdiag.cnrs.fr>



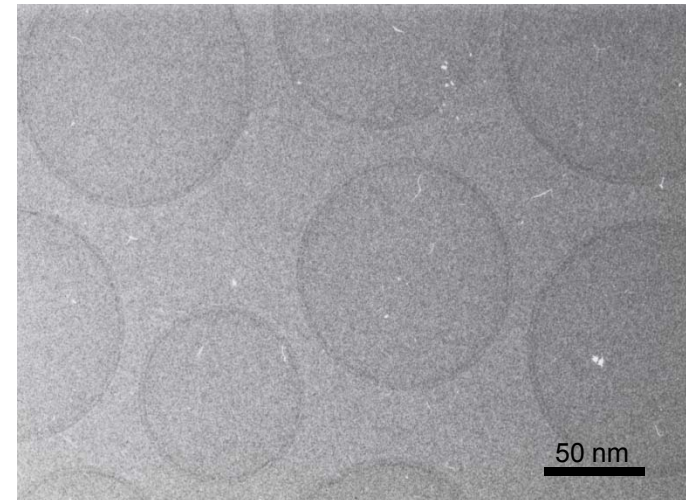
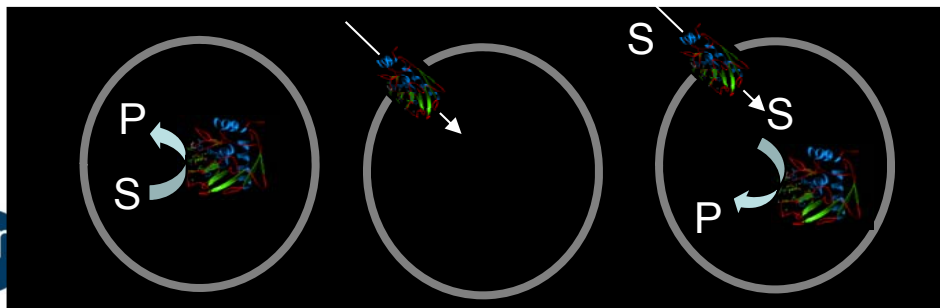
System design using our compound catalog
SBGN and *CellDesigner*

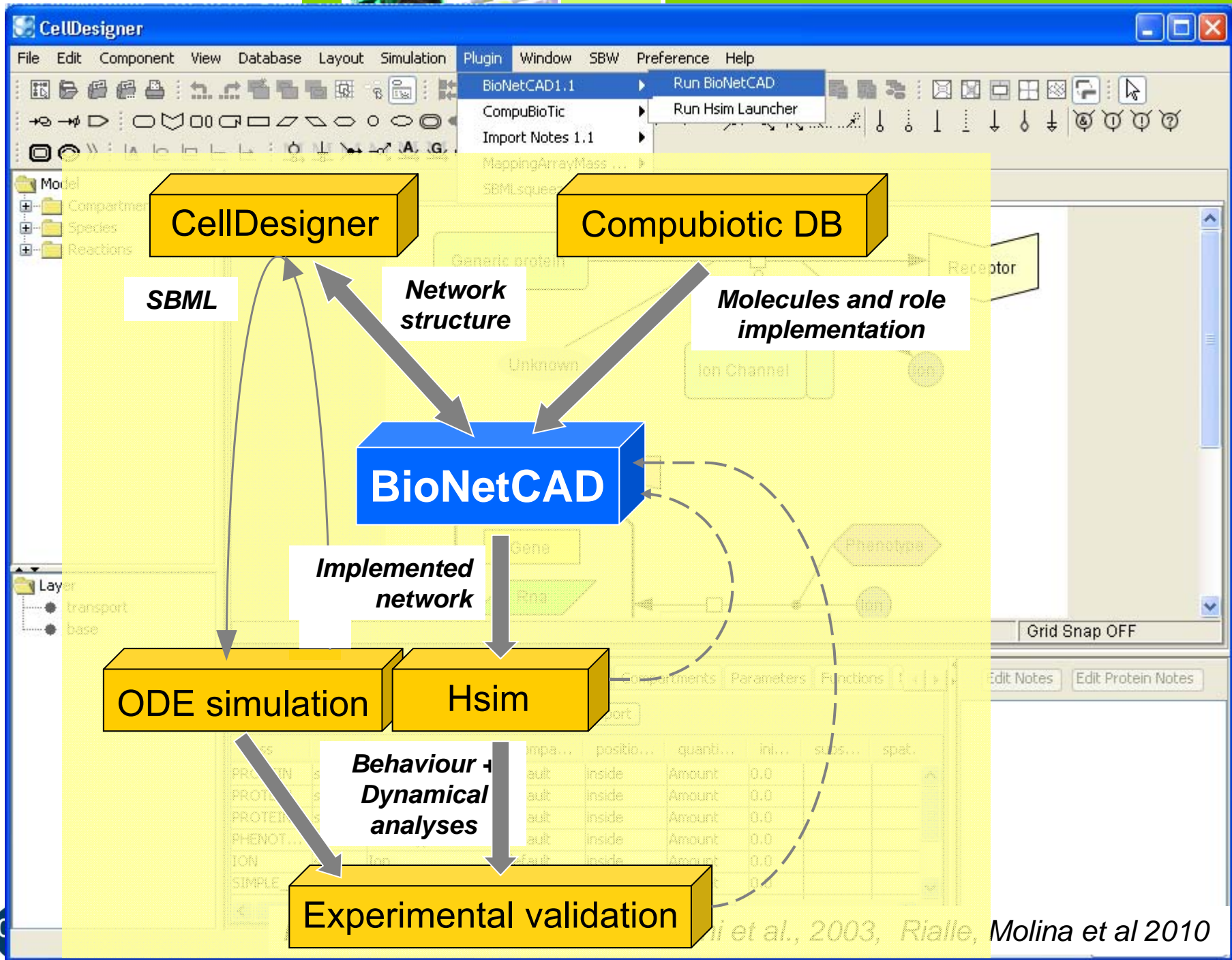
Bio Ψ modelling
Formal description

Simulation : Stochastic Cell automaton
and multi-agent

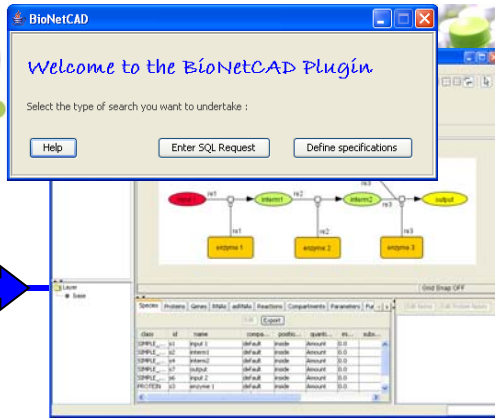
Experimental validation

- Stable Vesicles construction (liposomes) ~100nm
- Introduction of chosen functional compounds
- Opérational assays of full synthetic system (*in vitro*)



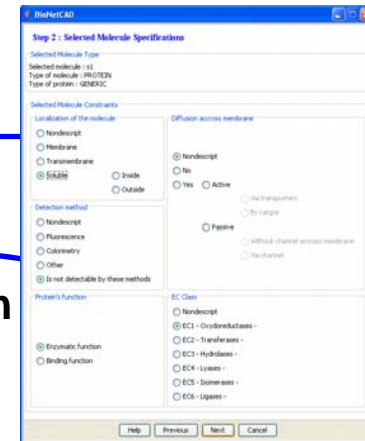


User needs



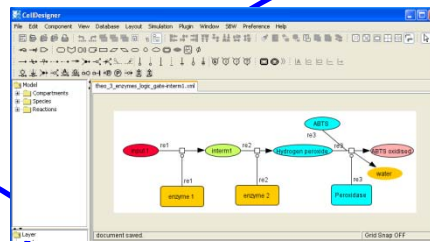
Abstract network

molecule
Sélection



I. Selection
molecule

Implemented
network



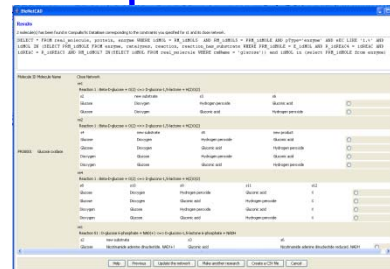
Intermediat
implemented network



Network constraints

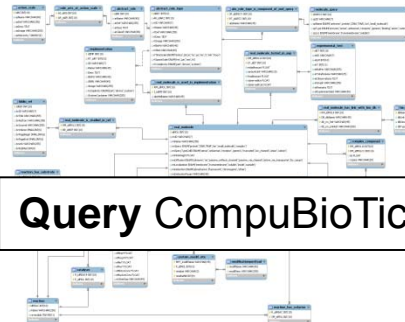


Simulation



Implementation
checking and validation

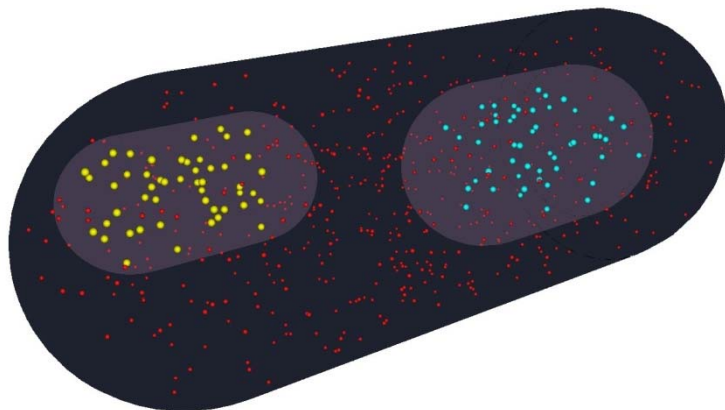
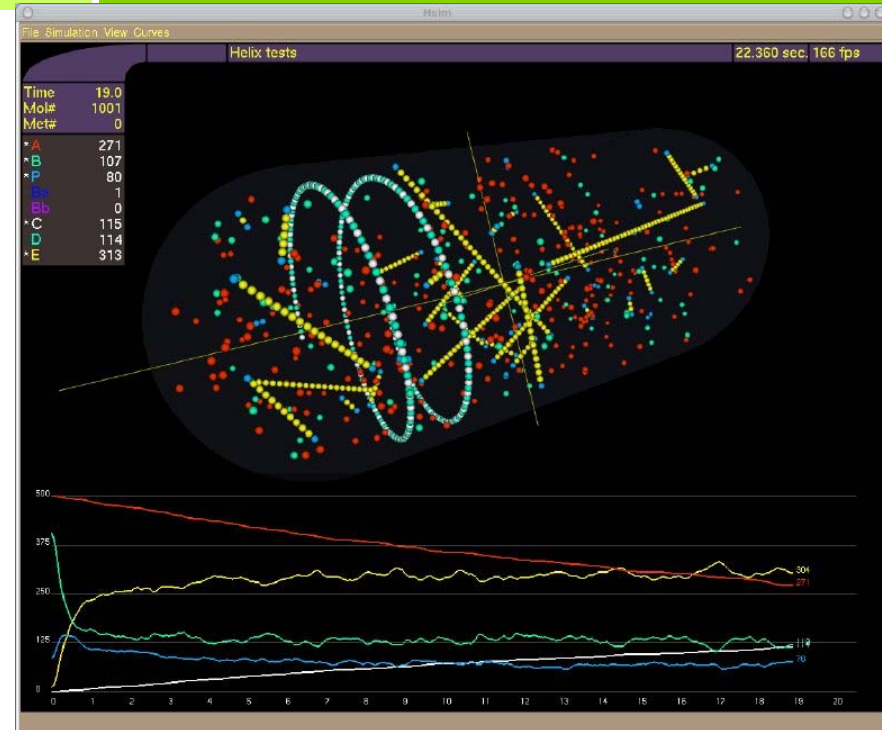
Query CompuBioTicDB





HSIM: Méthodes de calcul

1. Molécules individualisées
 - Localisation spatiale
 - Assemblages
 - Géométrie
2. Simulation Stochastique
 - Efficacité du calcul
3. Hybride: calcul individualisé/global
4. Equations différentielles ordinaires
 - Comportement moyen



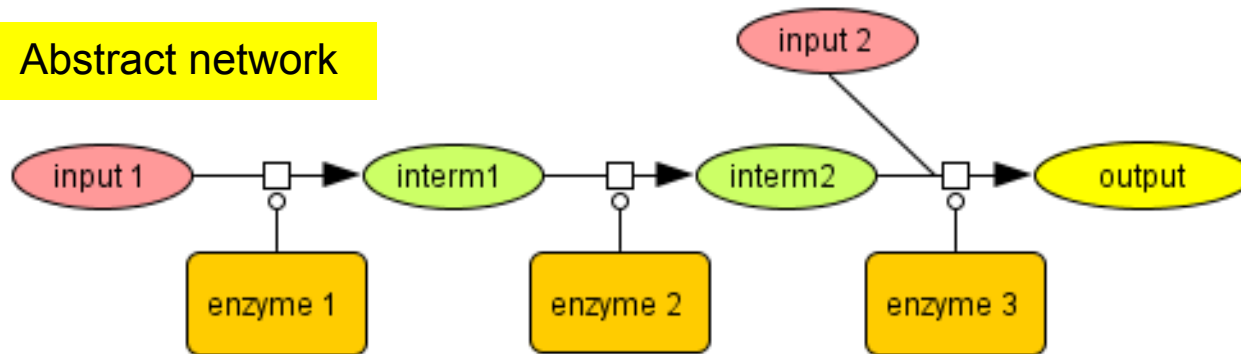
Caractéristiques

1. Multi compartiments
 - Diffusion passive a travers la membrane
2. Interactif : affichage graphique temps réel
3. Batch : sortie dans un fichier CSV



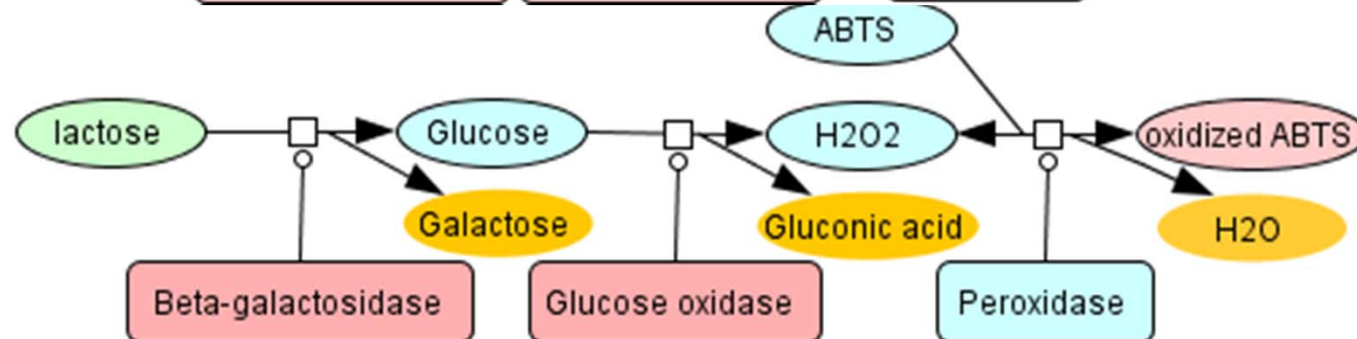
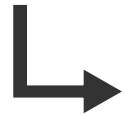
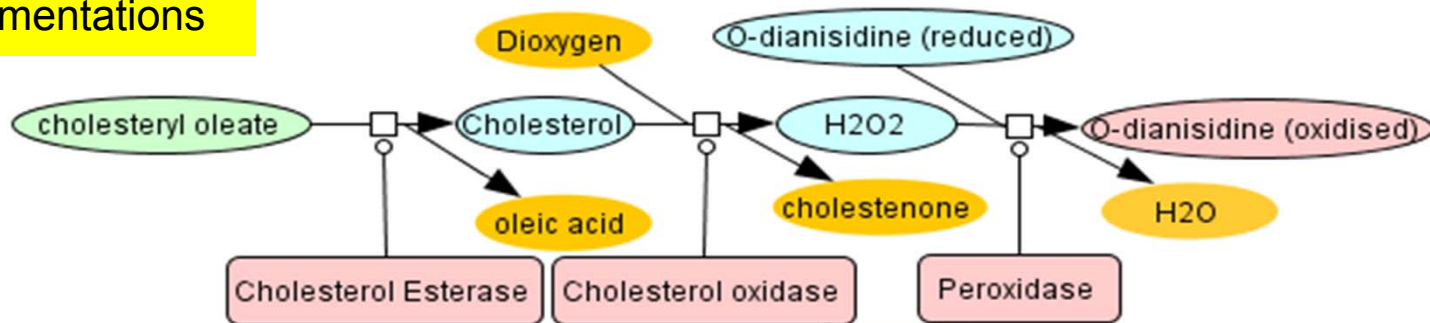
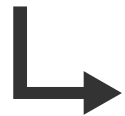
Bio-Logic gate design

Abstract network



input 1	input 2	output
0	0	0
0	1	0
1	0	0
1	1	1

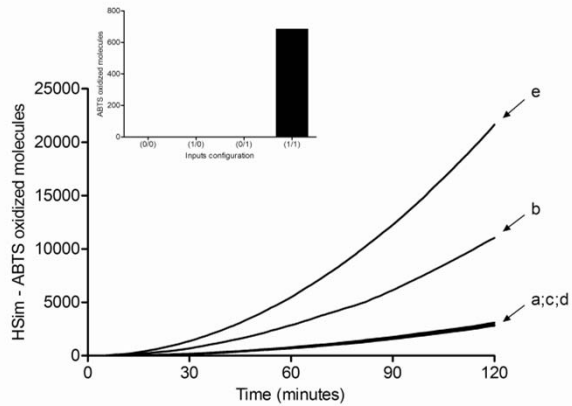
Network Implementations



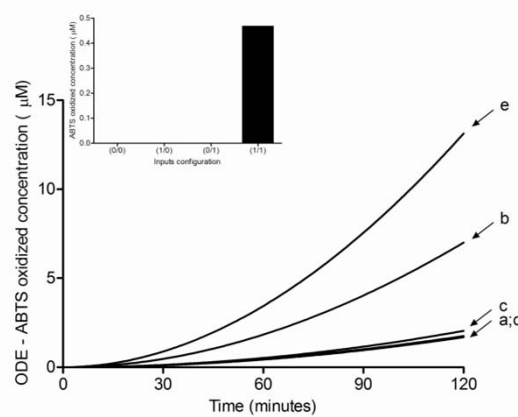


Simulation vs experimentation

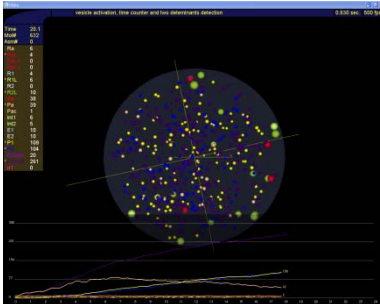
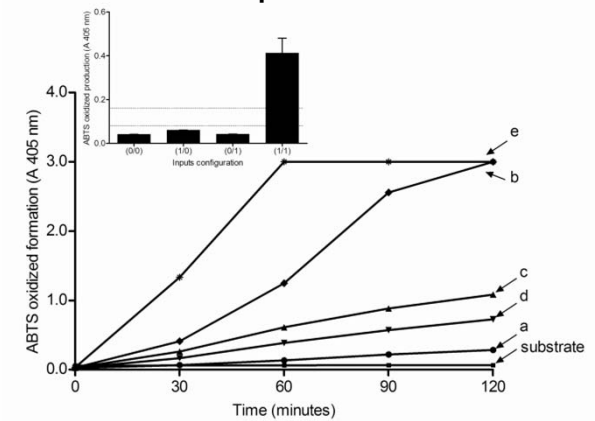
Multi-agent-like



ODE



Experiment



$$V_{\beta\text{-gal}} = kcat_{\beta\text{-gal}} \cdot [\beta\text{-gal}]_0 \cdot [\text{lactose}] / (K_{m_{\text{lactose}}} + [\text{lactose}])$$

$$V_{\text{god}} = kcat_{\text{god}} \cdot [\text{god}]_0 \cdot [\text{glucose}] / (K_{m_{\text{glucose}}} + [\text{glucose}])$$

$$V_{\text{pod}} = kcat_{\text{pod}} \cdot [\text{pod}]_0 \cdot [\text{H}_2\text{O}_2] \cdot [\text{ABTS}] / (K_{m_{\text{ABTS}}}[\text{H}_2\text{O}_2] + K_{m_{\text{H}_2\text{O}_2}} \cdot [\text{ABTS}] + [\text{H}_2\text{O}_2] \cdot [\text{ABTS}])$$

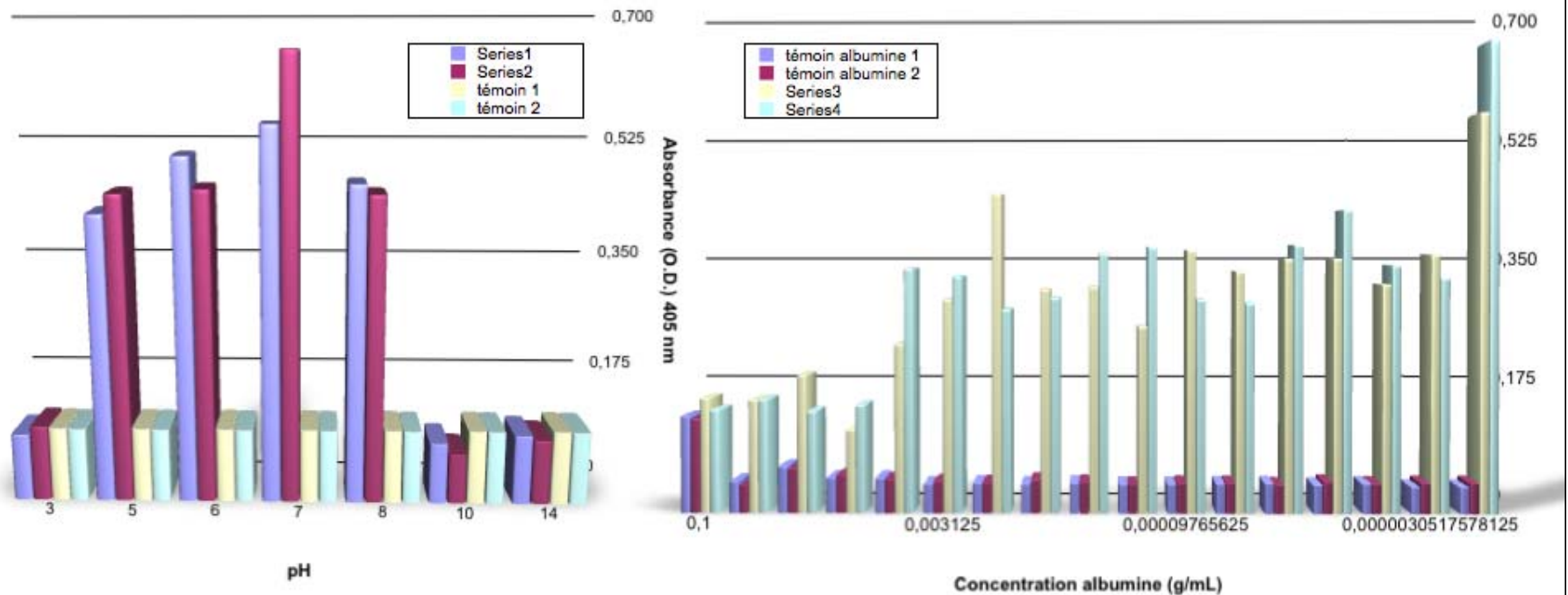
Rialle, S., Felicori, L., Dias-Lopes, C., Peres, S., El Atia, S., Thierry, A. R., Molina F. (2010). BioNetCAD: Design, simulation and experimental validation of synthetic biochemical networks. *Bioinformatics*, 26(18):2298-304

Etude de la porte bGal-GOD-POD (2)

Robustesse

- Etude de la résistance de la porte face à la perturbation : ajout d'albumine et variation du pH

Après 1h de réaction





New methods for biomarker detection (Synthetic biology)

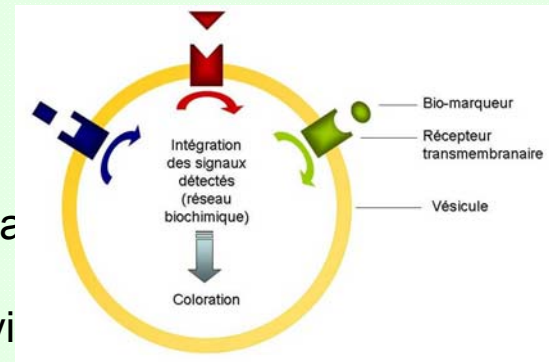
Implementation form ND/diabetis biomarker detection

« Close to the patient » simple assay

Multi-parametric measure

Sophisticated signal integration
(qualitative, quantitative, temporal, spacia

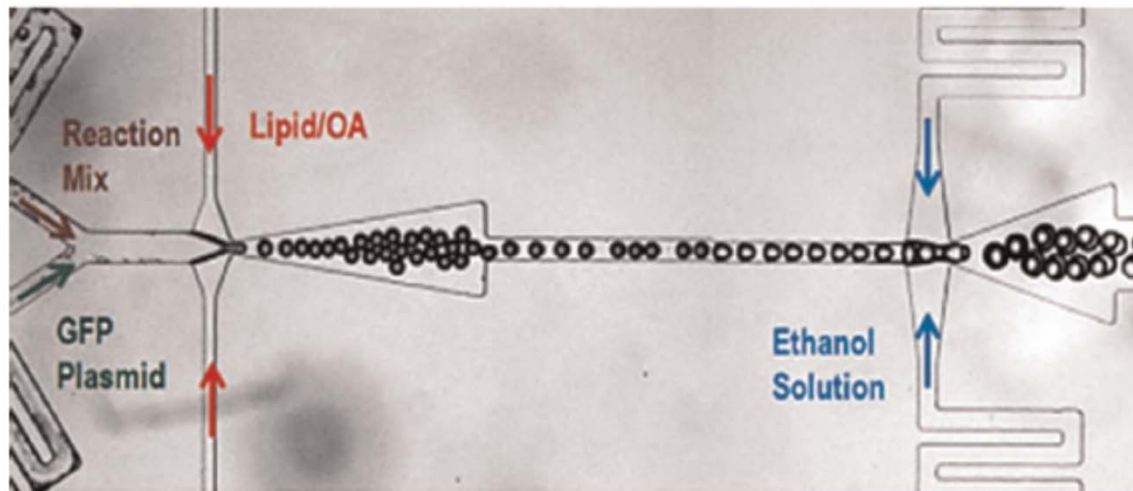
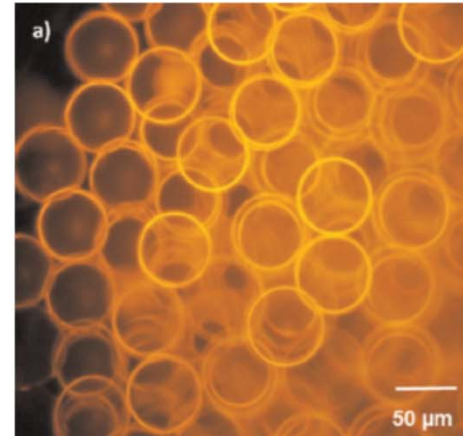
Result return in a simple way (local dyi



*Application to diabetic nephropathy
and colo-rectal cancer*



Vesicle building, film rehydratation, Microfluics etc.



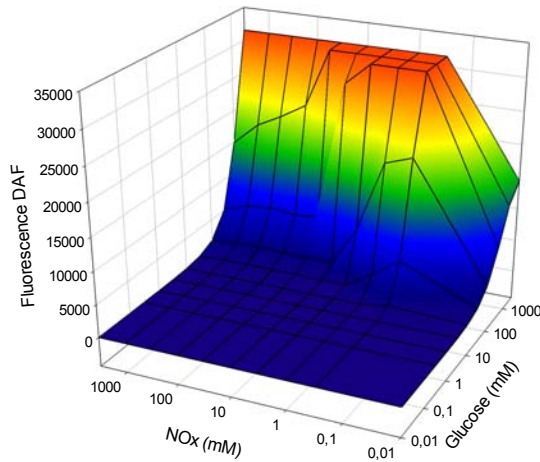


Ex Vivo synthetic biochemical Network : clinical relevant AND gate

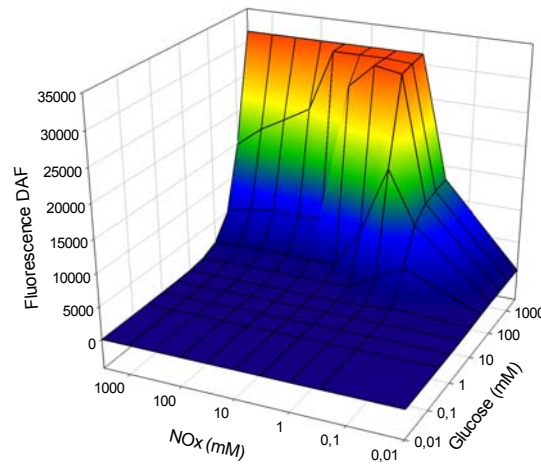
Transfer functions

Sensitivity to enzyme (G1DH/NR) concentration ratio?

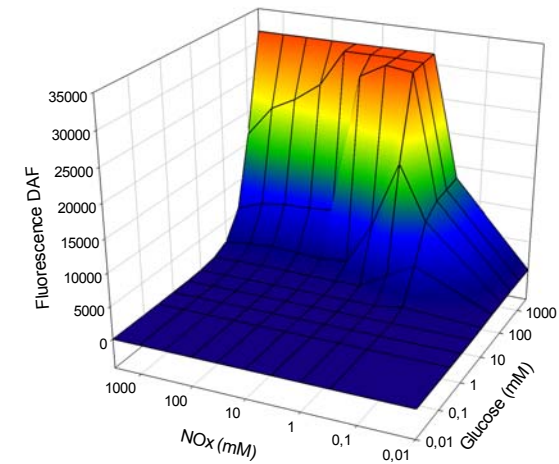
AND gate G1DH/NR=0.01



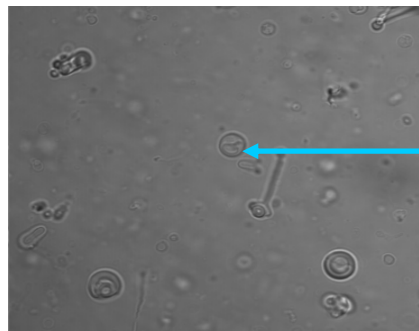
AND gate G1DH/NR=1



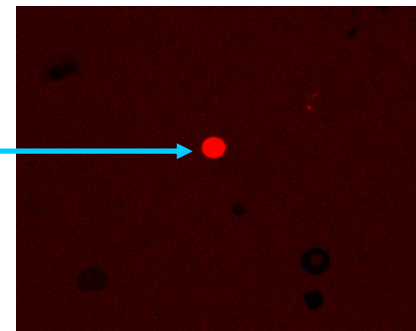
AND gate G1DH/NR=100



Simulation

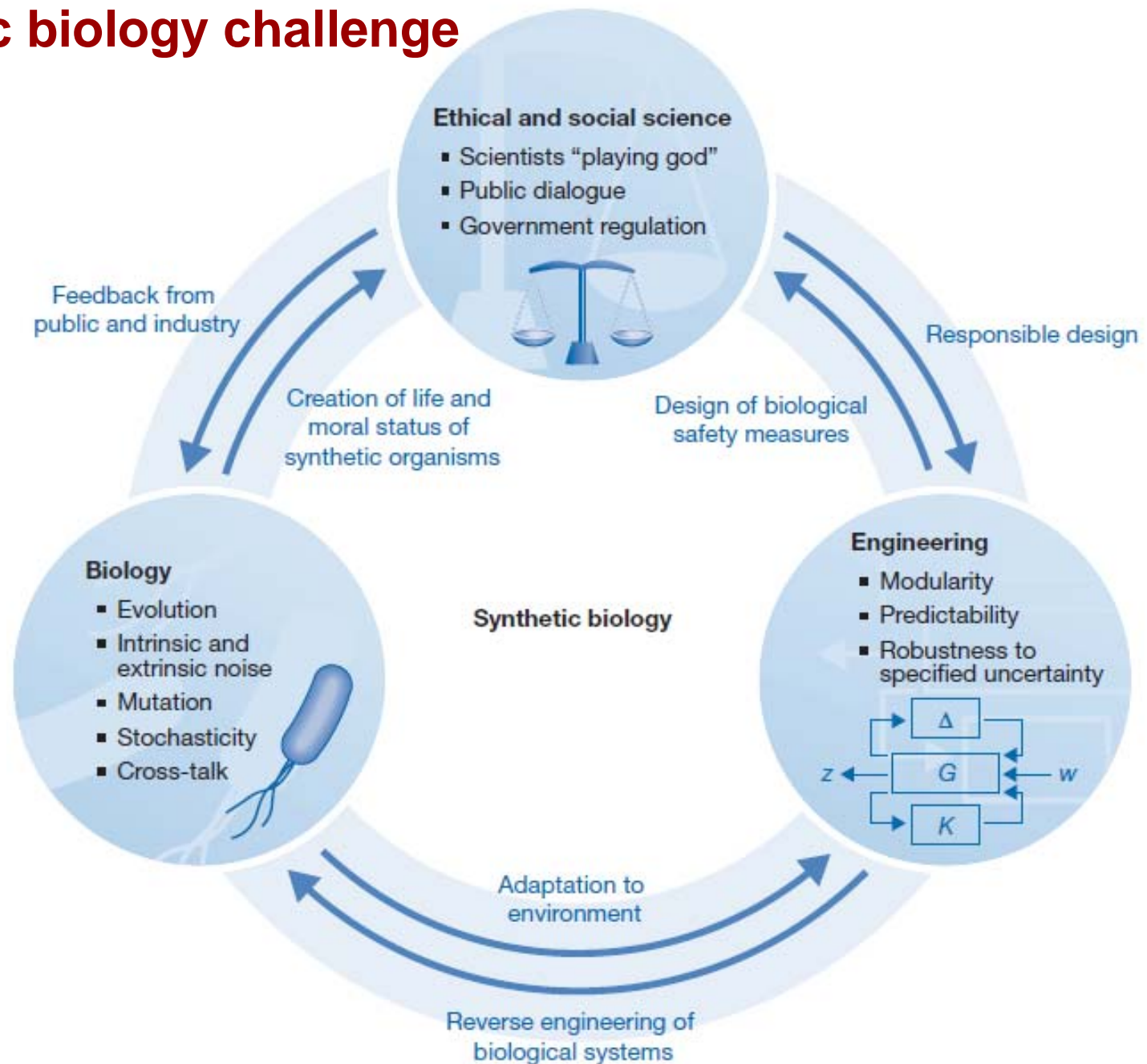


Transmission



Fluorescence (Cy3)

The Synthetic biology challenge





Challenges for Synthetic biology clinical applications

-Technical issues:

Robustness and reliability, comparison with current performances of clinical application

Table 1 | Engineering, natural and synthetic solutions for designing complex systems

Design challenge	Engineering solution	Natural organism	Synthetic biology solution
Scalability	Modularity	Motifs Modularity?	Modularity Well-characterized modules
Retroactivity/ cross-talk	Insulation and feedback	Feedback	Orthogonal design
Robustness	Feedback	Feedback	Feedback
Complexity	Hierarchical design	?	Hierarchical design
Evolution/ mutation	?	Efficiency/robustness trade-off	?

-Production and Cost issues

-Regulatory and Safety issues (Objective demonstration of composition, behavior, etc.)

-Ethics and social issues

-Misuse use (terrorism, etc.)



Collaborators:

Emilio Parisini, Havard Medical school, US

Alfonzo Jaramillo Univ Evry,

P. Amar, LRI, Orsay

Vic Norris, University lesceiter/Rouen

D. Boturyn, CNRS Grenoble

Luigi Luizi, Rome, Italy

J. Stelling, UTHZ, Switzerland

European BaSysBio partners UE FP6

H. Mischak, Hanover, DE

H. Holtofer, Dublin, Ireland

A Wlahou, Athens, Greece

A. Hamadi, Sfax, Tunisia

Sowdonamini and Srinivasan, Bangalore, NCBS, India.

C. Chavez, Belo Horizonte, Brazil

C. Godin INRIA, Montpellier

François Fages, INRIA Roquencourt

Tony Wilkinson, Univ. York, UK

Jerome Bonnet, Drew Endy, Univ Stanford, US



Systems and synthetic biology team

Franck Molina

Alexis Courbet

Dung LeNguyen

Laurence Molina

Nicolas Salvetat

Patrick Amar (Visitor)

David Jean

Christophe Nguyen

Emmanuelle Sidot

Former members:

Claude Granier

Liza Felicori

Sabine Peres

Stéphanie Rialle

Randa Benhameur

Camila Lopes

Chams Kifagi

Francisco Schneider

Sylvaine Cordroch

Cécile Fleury

Sarah pariset



Grants:

BaSysBio UE FP6, Eurokup UE FP7 ; LIA CNRS

ANR MSDmind ; PICS CNRS ; Région LR

France-Univ Stanford Grant

Clinicians

E. Renard, CHRU Montpellier

F. Jarraya Hopital de Sfax, Tunisie

A. Argiles, CHU Sète

M. Ychou, CRLC, Montpellier

Corinne Fayolle, CHU, Montpellier

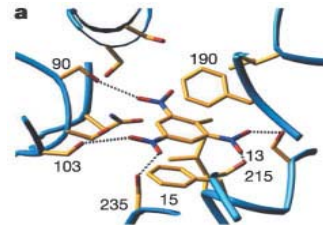




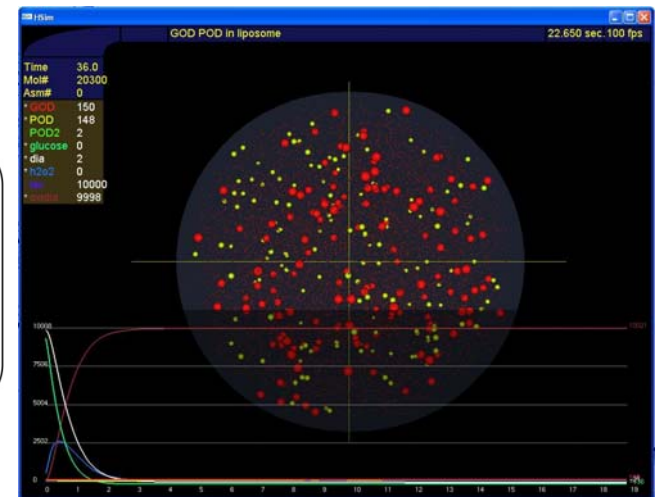
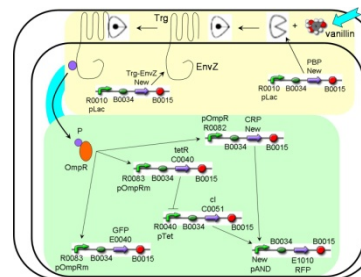
Complex System Modeling and Engineering for Diagnosis
UMR3145 CNRS/Bio-Rad, Montpellier



<http://www.sysdiag.cnrs.fr>



F. Molina, CNRS, Montpellier



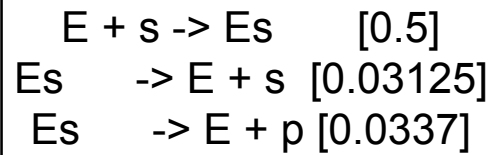


Modèle HSIM

E (s -> p) Km = 0.25 mM; Kcat=337;

init (E, 0.1 uM);
init (s, 100 uM);

Réactions



Export ODE

$$\frac{d[E]}{dt} = -k_1 \cdot [E] \cdot [s] + k_2 \cdot [Es] + k_3 \cdot [Es]$$

$$\frac{d[s]}{dt} = -k_1 \cdot [E] \cdot [s] + k_2 \cdot [Es]$$

$$\frac{d[p]}{dt} = k_3 \cdot [Es]$$

$$\frac{d[Es]}{dt} = k_1 \cdot [E] \cdot [s] - k_2 \cdot [Es] - k_3 \cdot [Es]$$

Paramètres

$$k_1 = 0.00847771$$

$$k_2 = 312.5$$

$$k_3 = 649.5.$$

