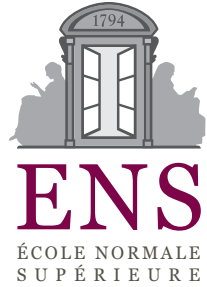


IBENS

Institut de Biologie de l'École Normale Supérieure



Inserm



Modelling of T cell co-inhibitory pathways to predict anti-tumour responses to checkpoint inhibitors

Céline Hernandez
Computational Systems Biology (D. Thieffry) – IBENS, Paris, France

Biological context : immunotherapies

Immunotherapies using **checkpoint inhibitors**

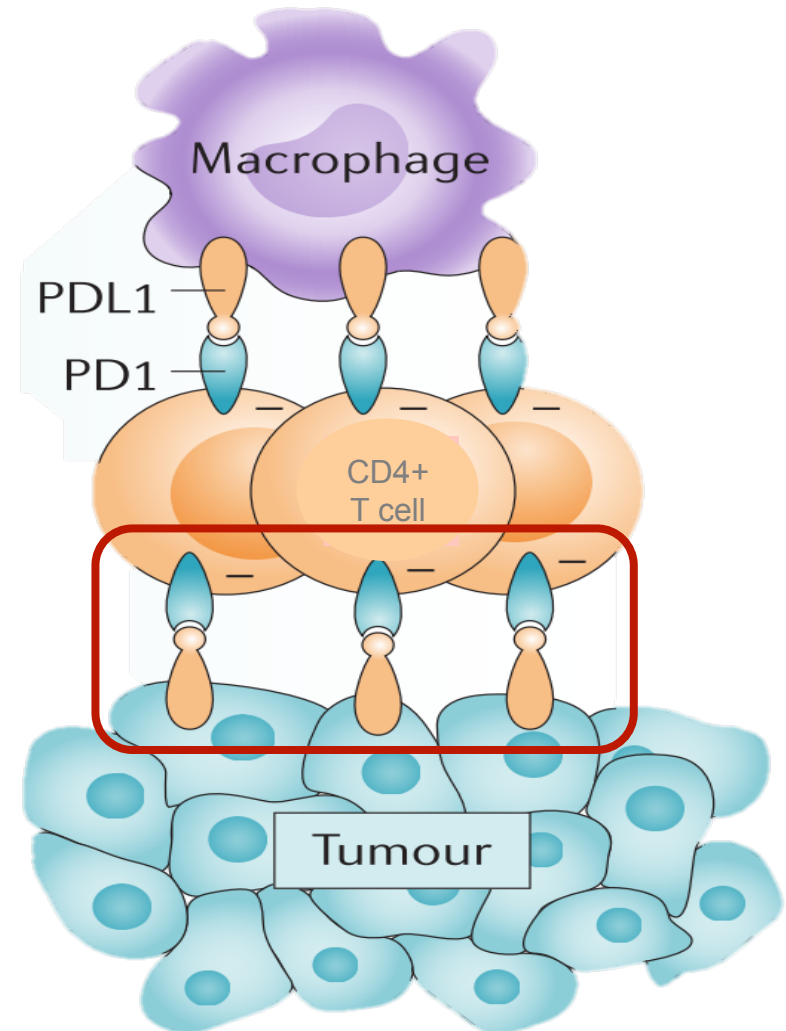
Cancer can be cured
by our own immune system

Biological context : immunotherapies

Tumour cells

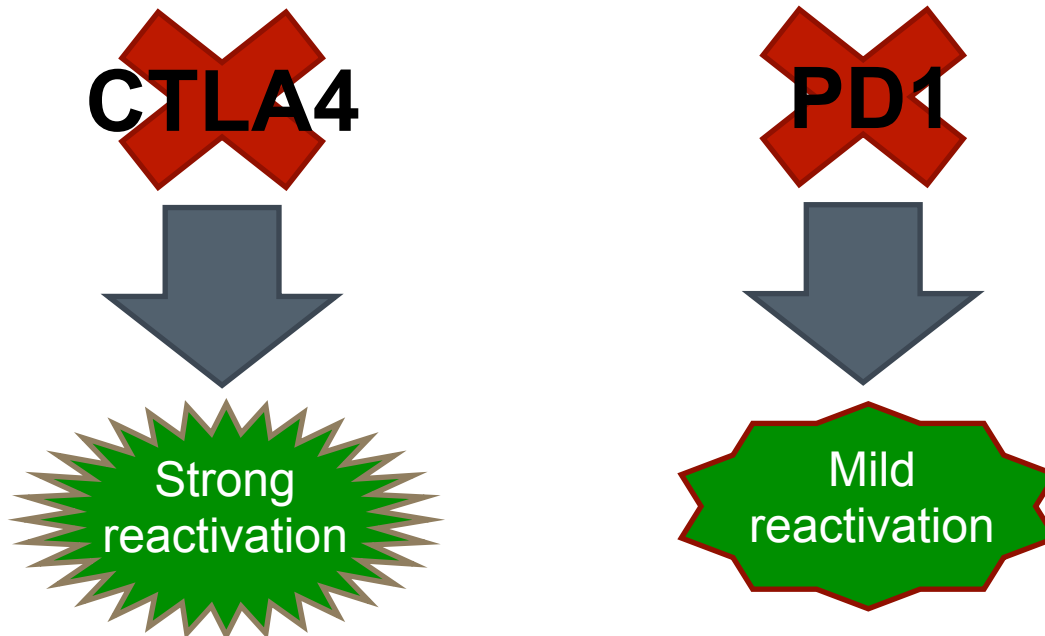
- ◆ Are detected by the immune system (e.g. T cells)
- ◆ But present ligands to inhibitory receptors, i.e. **checkpoints** (CTLA4, PD1)
- ◆ Thus evade the immune system recognition, preventing the response

**FDA-approved
anti-CTLA4 and anti-PD1
immunotherapies**



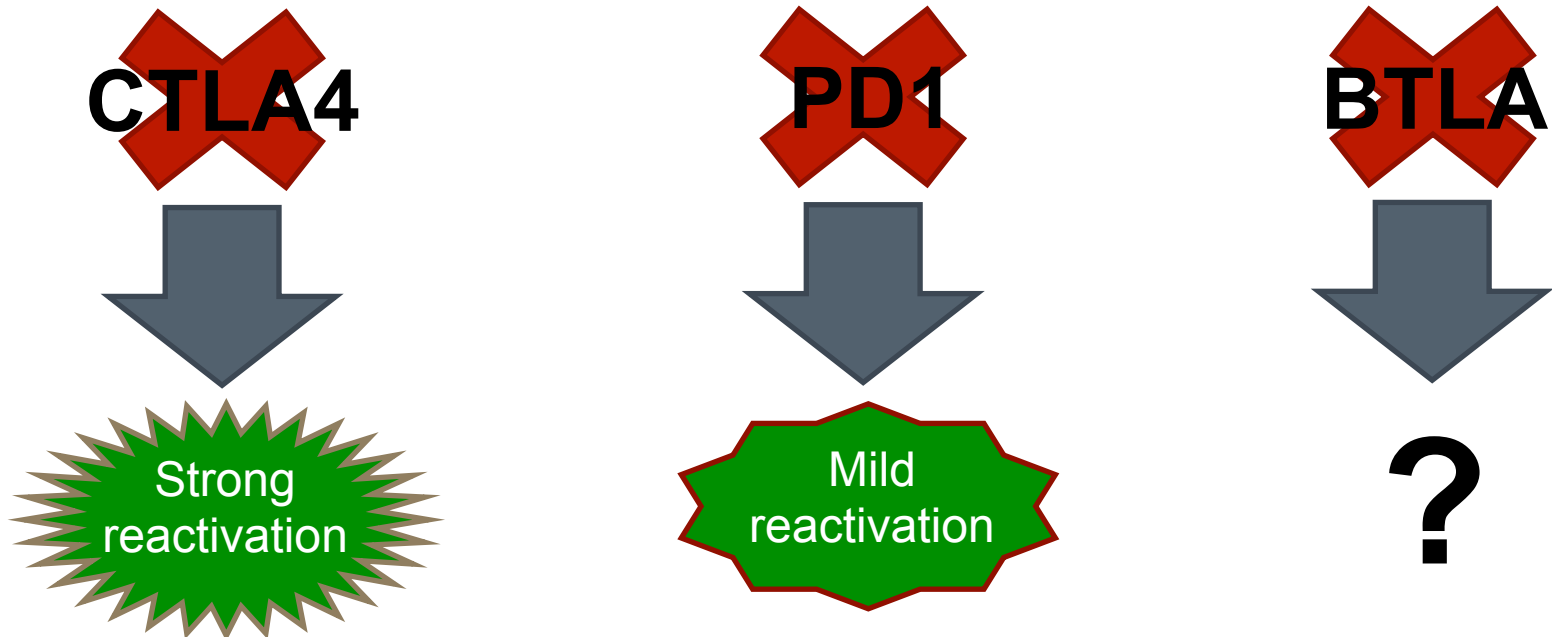
Biological context : immunotherapies

- ◆ Checkpoint inhibitors target specific ligand/receptor communication to re-activate T cells
- ◆ Downside: serious adverse effects, different in intensities



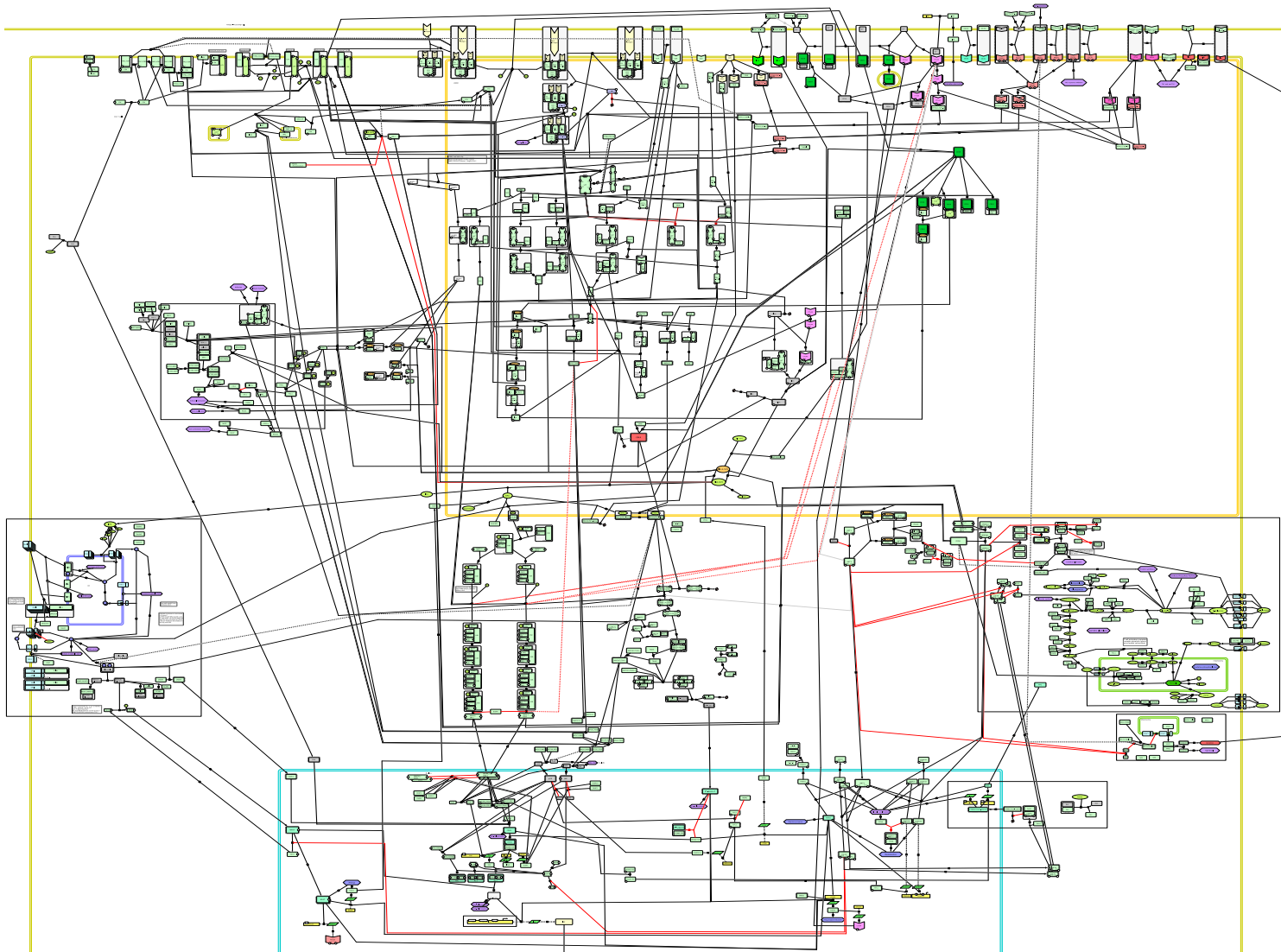
Biological context : immunotherapies

- ◆ Checkpoint inhibitors target specific ligand/receptor communication to re-activate T cells
- ◆ Downside: serious adverse effects, different in intensities



Why?
What's the difference?

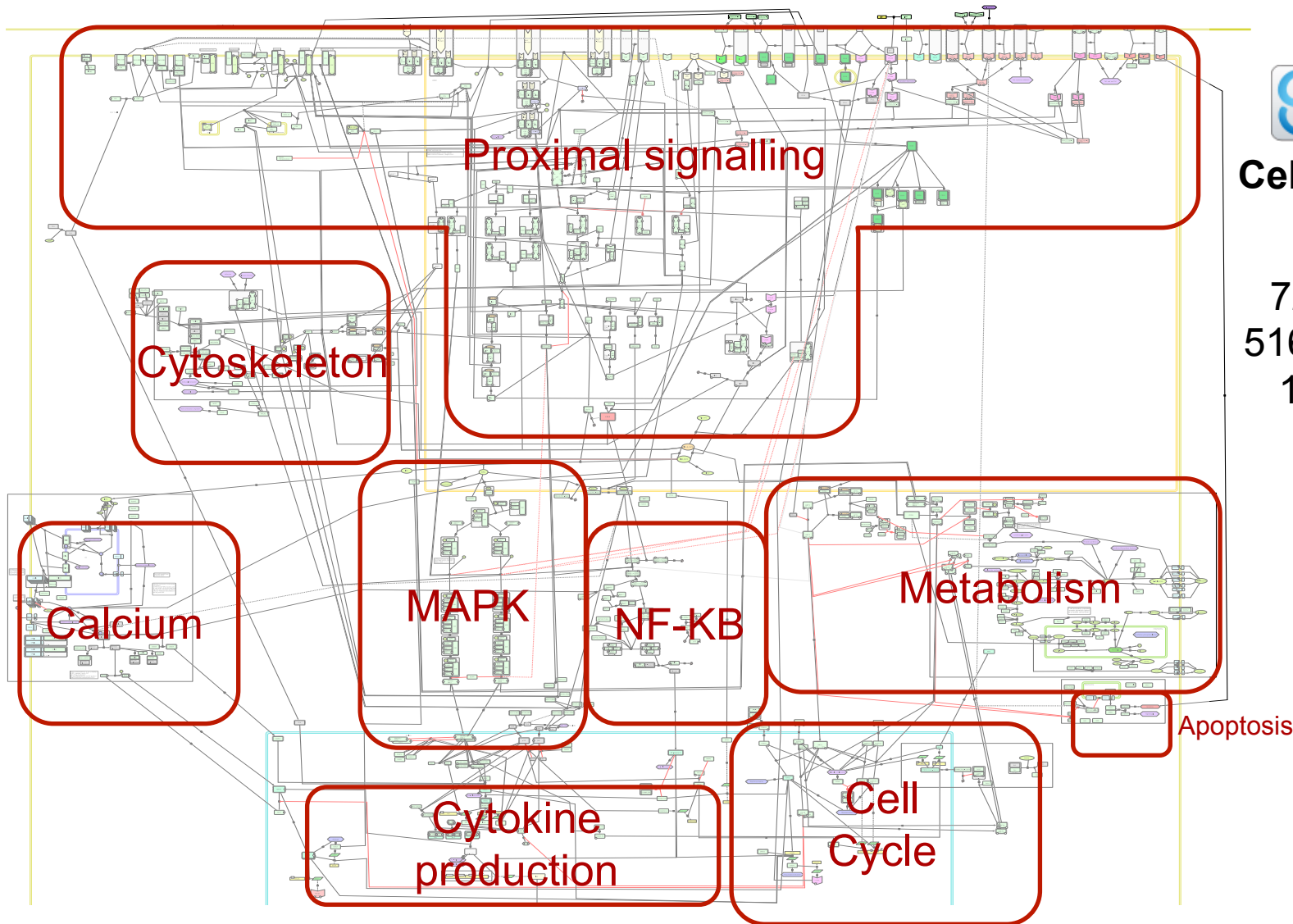
Molecular mapping of CD4+ T cell activation



CellDesigner

726 species
516 reactions
123 articles

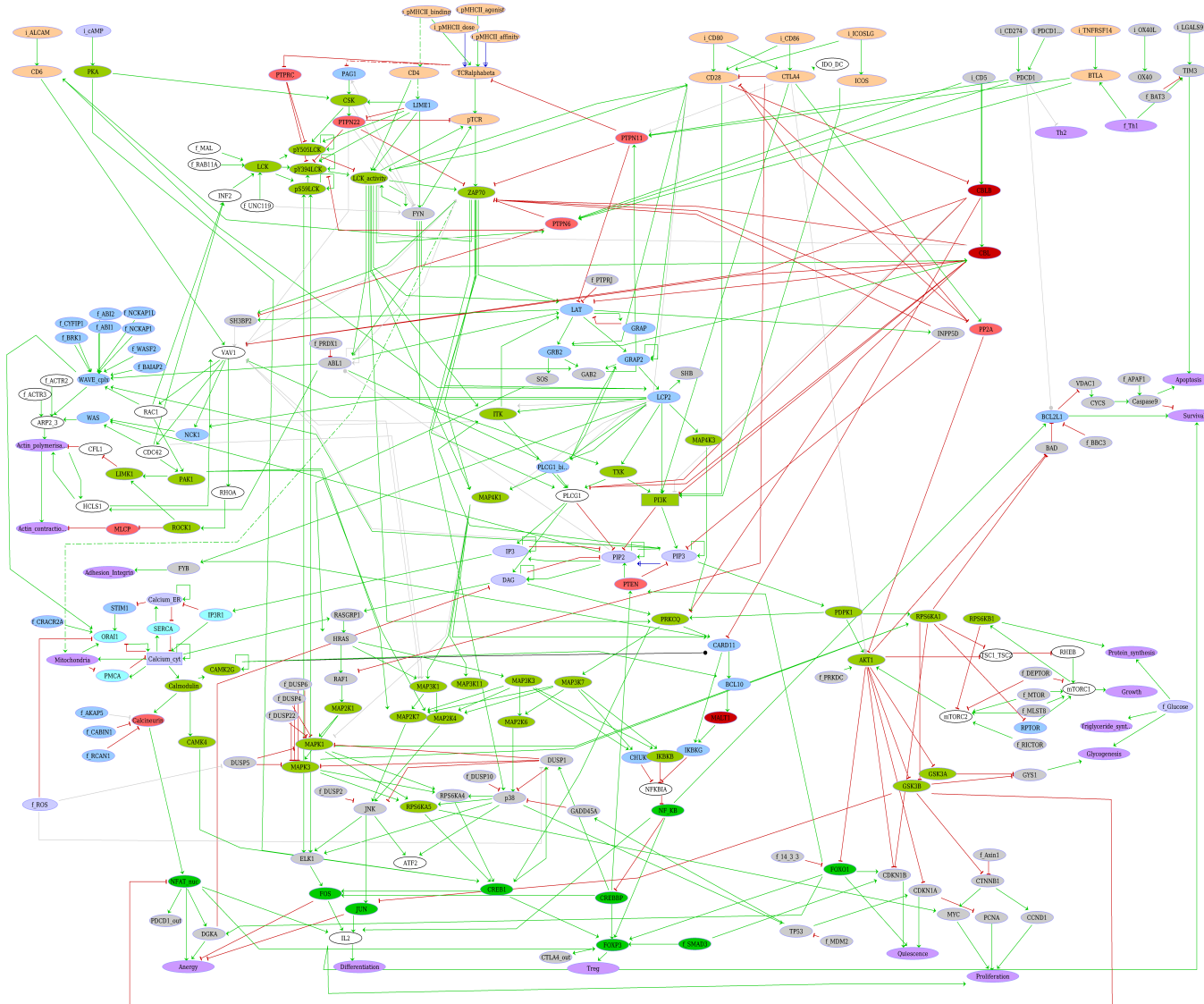
Molecular mapping of CD4+ T cell activation



CellDesigner

726 species
516 reactions
123 articles

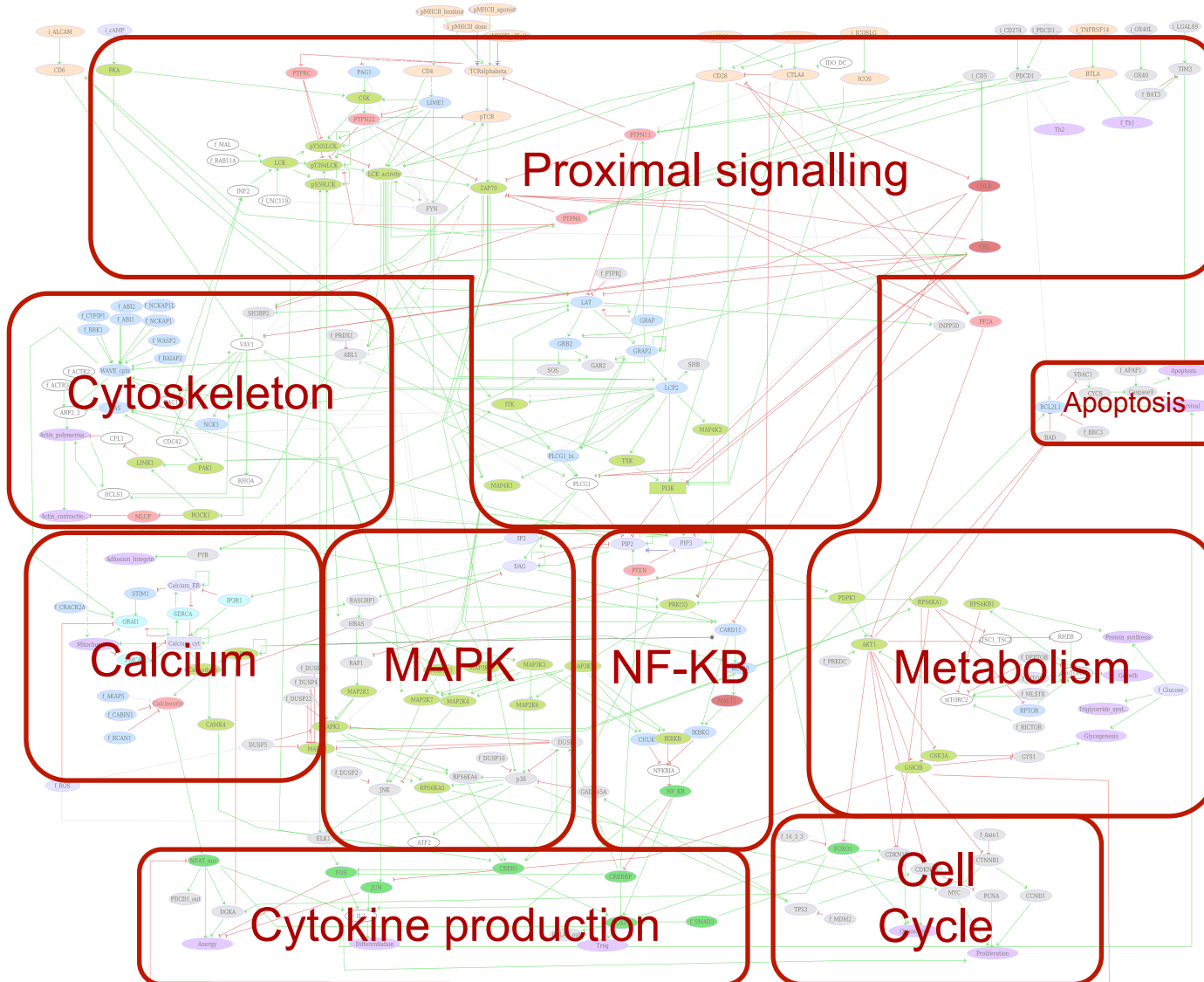
Dynamical modelling: regulatory graph



216 nodes
15 inputs
40 fixed values

450 edges
322 activations
120 inhibitions

Dynamical modelling: regulatory graph



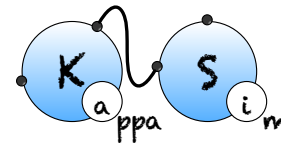
216 nodes
15 inputs
40 fixed values

450 edges
322 activations
120 inhibitions

Dynamical modelling: coping with complexity

- ◆ Dynamical modelling using a Rule-based approach

Collaboration with J. Ferret (ENS Paris)

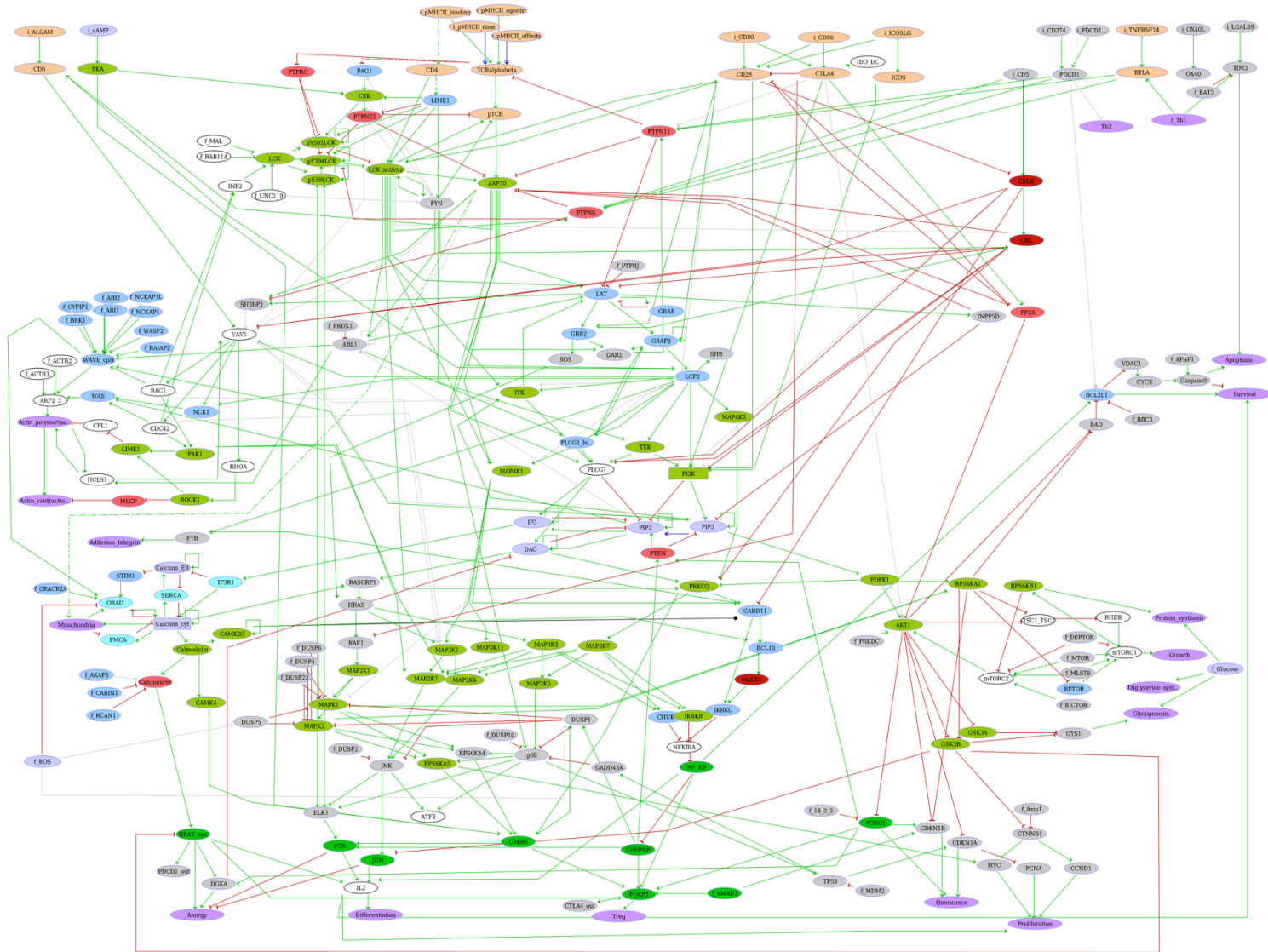


- ◆ Logical modelling with a combination of **unit testing** and **sub-model extraction**

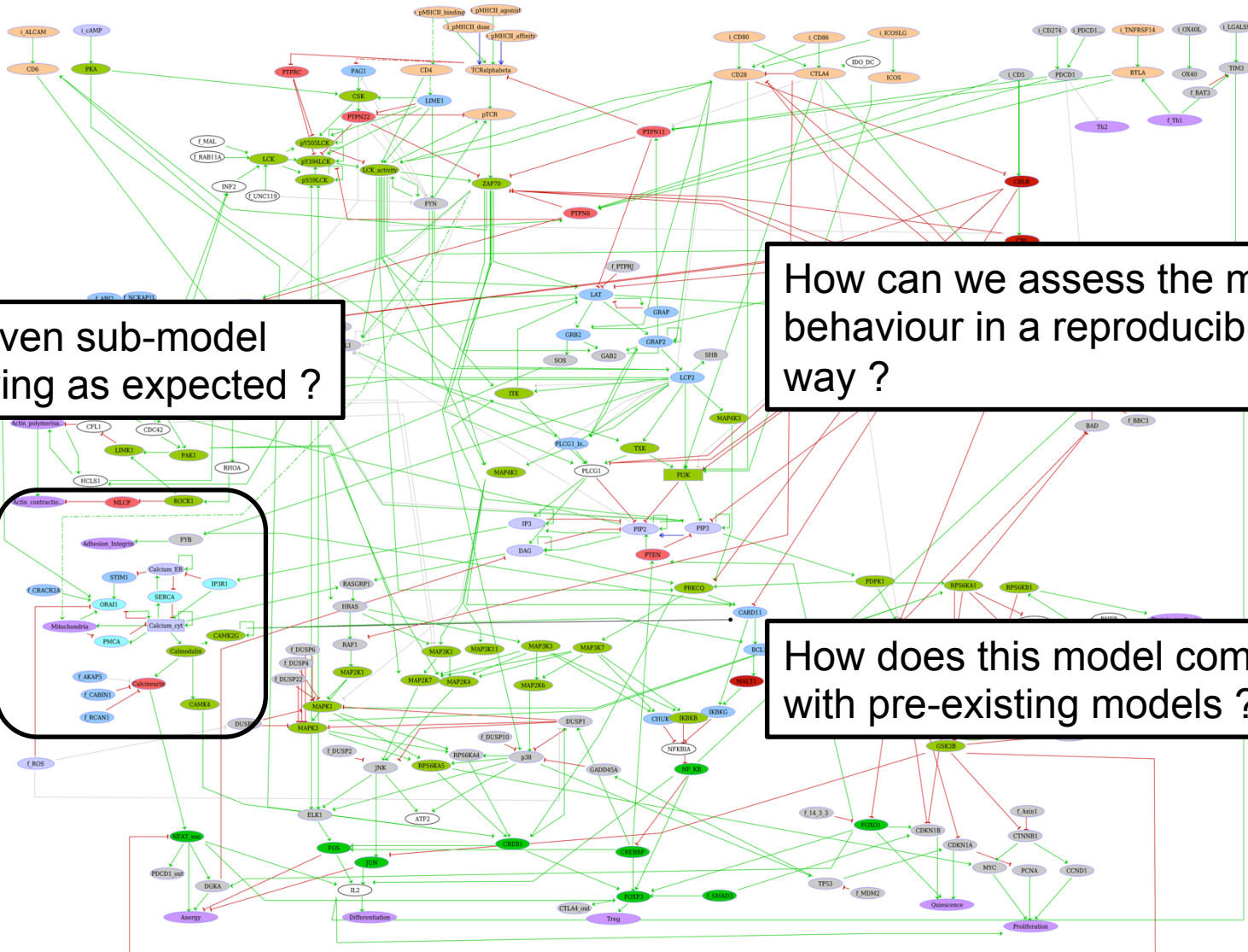
CoLoMoTo Docker image



Logical modelling : questions and challenges



Logical modelling : questions and challenges



Is a given sub-model behaving as expected ?

How can we assess the model behaviour in a reproducible way ?

How does this model compare with pre-existing models ?

Logical modelling with the CoLoMoTo Docker image

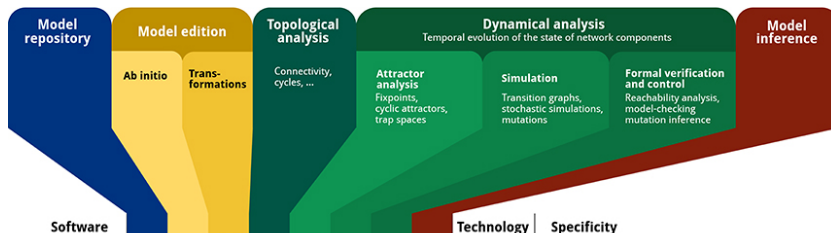
How can we assess the model behaviour in a reproducible way ?



CoLoMoTo Docker image

+

Unit testing (Python)



Software	Model repository	Model edition	Topological analysis	Dynamical analysis	Model inference	Technology	Specificity
CellCollective	●	●	●	●		Javascript	Web application
GINsim	●	●	●	●		Java	Graphical and script interface
bioLQM			●	●		Java, ASP	Command line and script interface
MaBoSS		●		●		C++	Stochastic simulations
Pint		●	●	●		OCaml, ASP	Large-scale analysis; formal inference of mutations
NuSMV				●		C	General purpose symbolic model-checker
BoolSim			●			C++	Scalable cyclic attractors identification
BooleanNet		●	●	●		Python	Discrete and hybrid semantics
pyBoolNet		●	●	●		Python, ASP	Boolean network toolbox
BoolNet			●	●		R	Simulation, attractor analysis, reverse-engineering
CellNOptR						R	Network optimisation, series
CaspoTS						ASP	Exhaustive inference, series

```
In [1]: 1 import unittest
        2 import biolqm
        3
        4 class ExampleTest(unittest.TestCase):
        5     def test_fixedpoints(self):
        6         # Load model to be tested
        7         lqm = biolqm.load("http://ginsim.org/sites/default/f
        8         # Compute fixed points using bioLQM
        9         fixpoints = biolqm.fixedpoints(lqm)
        10        # Test case: there should be only one fixed point
        11        self.assertEqual(len(fixpoints), 1)
        12
        13 runner = unittest.TextTestRunner(verbosity=2)
        14 runner.run(unittest.makeSuite(ExampleTest))
```

This notebook has been executed using the docker image `colomoto/colomoto-dc`

```
test_fixedpoints (__main__.ExampleTest) ...
Downloading 'http://ginsim.org/sites/default/files/phageLambda4.zginml'
ok
-----
Ran 1 test in 0.314s
OK
```

Out[1]: <unittest.runner.TextTestResult run=1 errors=0 failures=0>



Adapted from Naldi et al. (2018) Front. Physiol.

Logical modelling with the CoLoMoTo Docker image

How does a model compare with pre-existing models ?

Create dedicated notebooks for already published models



cellcollective



BioModels

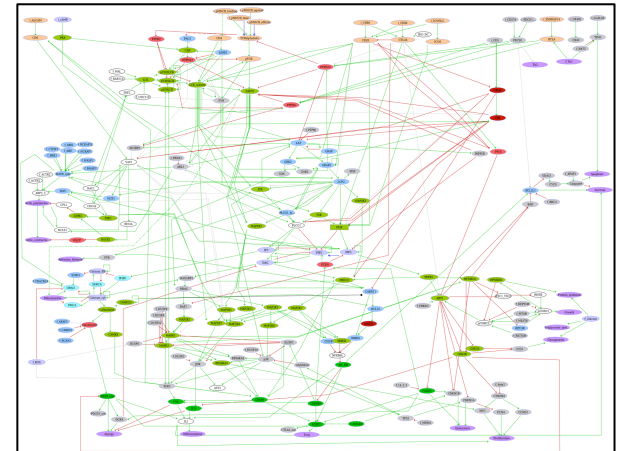


```
In [1]: 1 import unittest
2 import biolqm
3
4 class ExampleTest(unittest.TestCase):
5     def test_fixedpoints(self):
6         # Load model to be tested
7         lqm = biolqm.load('http://ginim.org/sites/default/files/phase4_ambd44.zqmml')
8         # Compute fixed points using biolqm
9         fixpoints = biolqm.fixpoints(lqm)
10        # Test case: there should be only one fixed point
11        self.assertEqual(len(fixpoints), 1)
12
13 runner = unittest.TextTestRunner(verbosity=2)
14 runner.run(unittest.makeSuite(ExampleTest))
```

This notebook has been executed using the docker image colomoto/colomoto-docker

```
test_fixedpoints (_main_.ExampleTest) ...
Downloading 'http://ginim.org/sites/default/files/phase4_ambd44.zqmml'
ok
Ran 1 test in 0.3144
OK
```

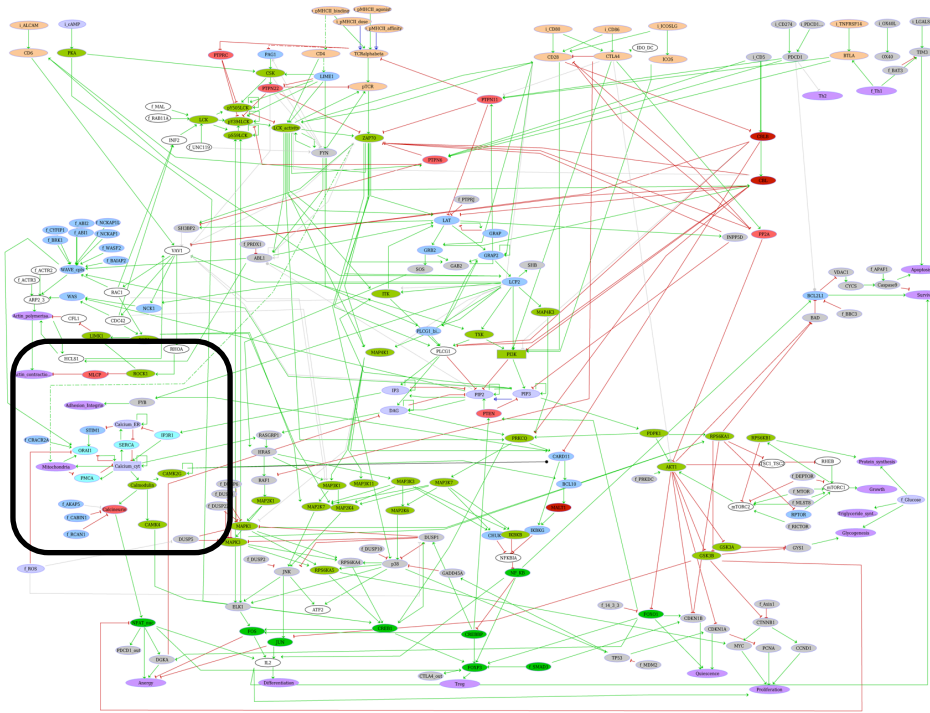
Transposable test cases



Unit testing
jupyter

Logical modelling with the CoLoMoTo Docker image

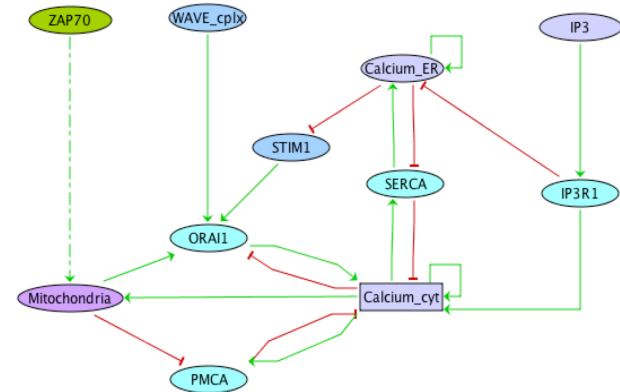
Is a given sub-model working as expected ?



on-the-fly extraction of a sub-model



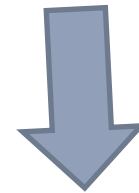
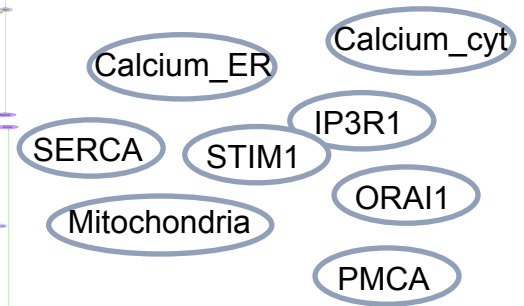
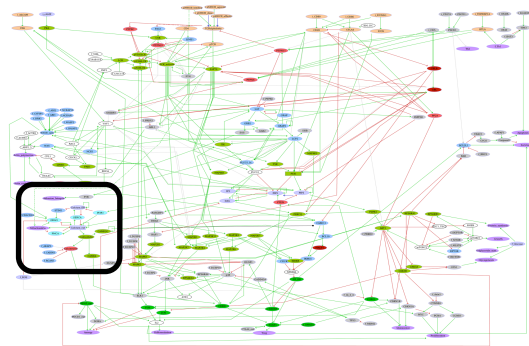
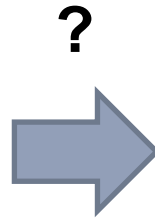
bioLQM
new “submodel” modifier
<http://www.colomoto.org/biolqm>



- Extract a core set of components
- Additional regulatory components considered as inputs

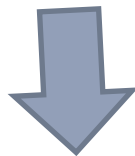
Logical modelling with the CoLoMoTo Docker image

"IP3 diffuses through the cytosol and binds to IP3 receptors located on the endoplasmic reticulum (ER) membrane, which results in a rapid release of intracellular calcium stores. This moderate and transient rise in the intracellular calcium concentration activates store-operated calcium entry (SOCE) channels in the plasma membrane to induce sustained elevations of intracellular calcium required for optimal TCR-induced signal transduction." Baine et al. 2009 Immun. Rev.

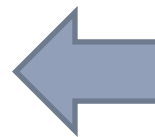


Test case

IP3	ZAP70	WAVE_cplx	IP3R1	Calcium_cyt	Calcium_ER	SERCA	STIM1	ORAI1	PMCA	Mitochondria
1	1	1	*	*	1	*	*	*	*	*
Expected :										
1	1	1	?	2	?	?	?	?	?	?

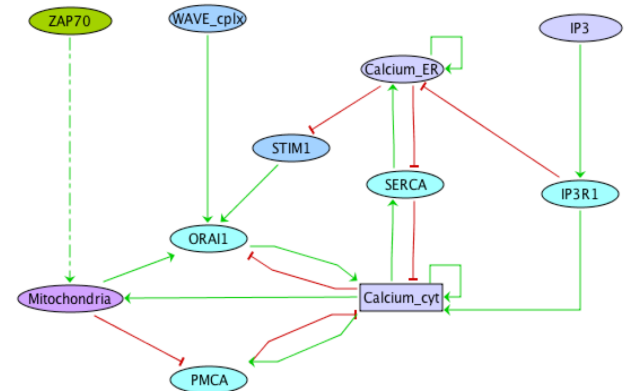


Ran 1 test in 0.051s
OK

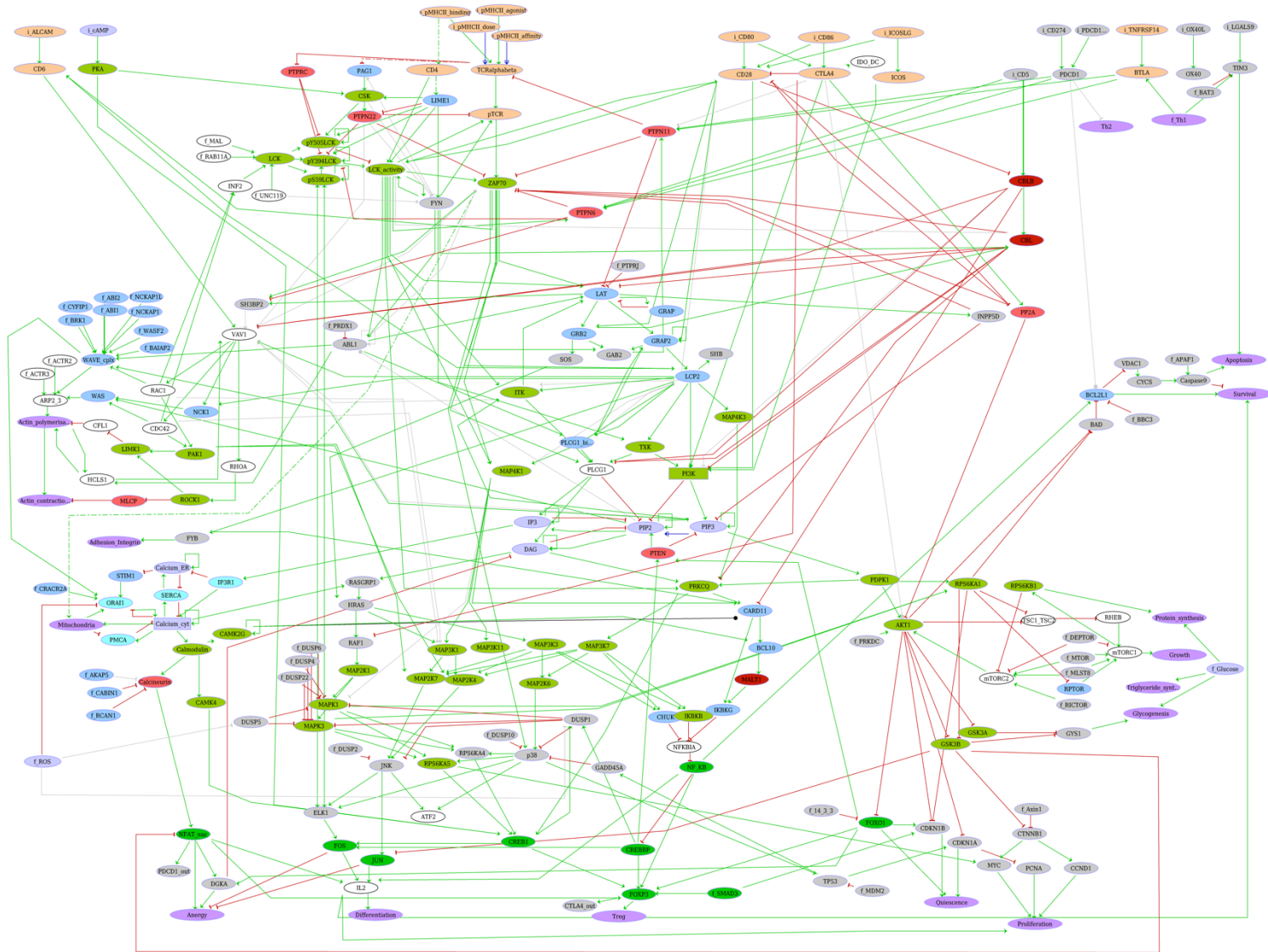


biolqm.fixpoints()

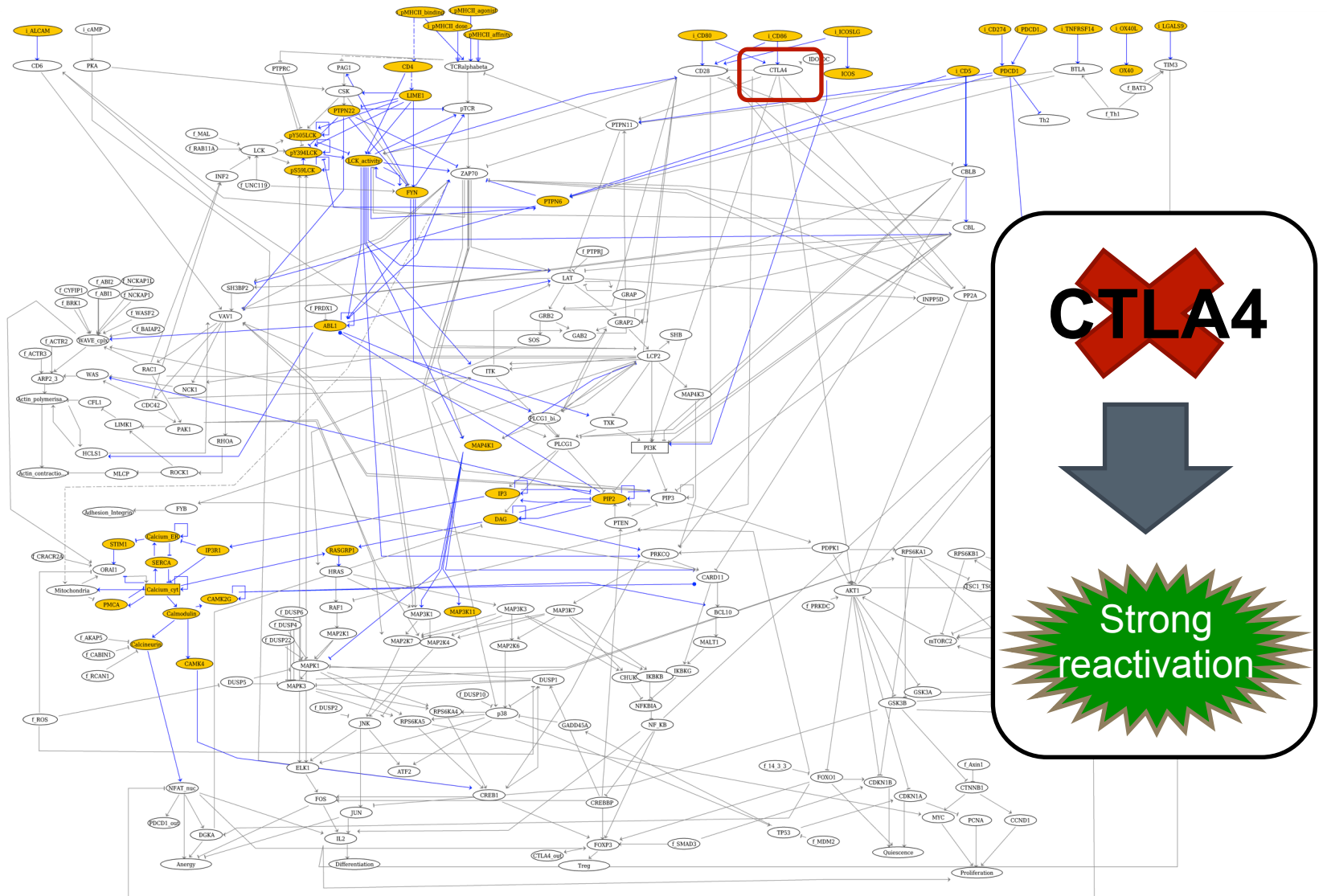
biolqm.submodel()



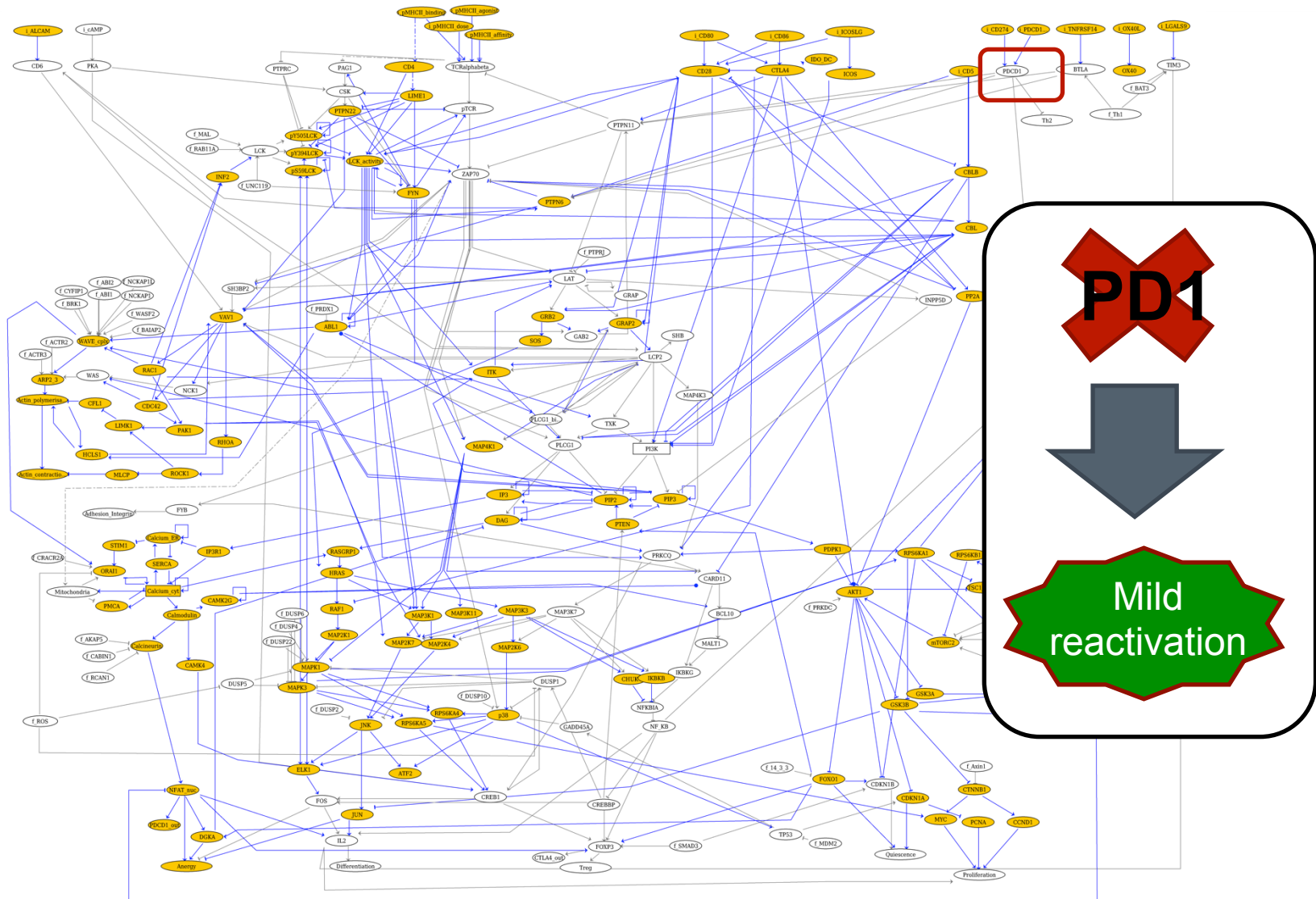
Logical modelling T cell co-inhibitory pathways



Percolation of CTLA4 activation



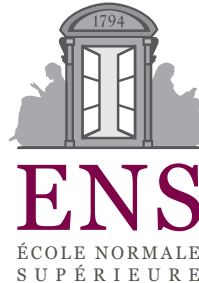
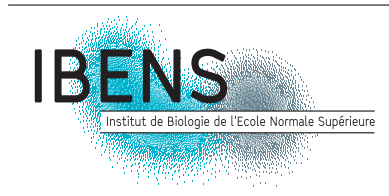
Percolation of PD-1 activation



Take home messages

- ◆ An automated framework based on unit testing
 - Reproducible (CoLoMoTo Docker image)
 - Allows comparison with previous models
- ◆ Iterative and modular method of model refinement
 - On-the-fly extraction of sub-models
 - Implemented in bioLQM (<http://colomoto.org/biolqm/>)
- ◆ Percolation analysis to understand the impact of checkpoint inhibitors
- ◆ Perspectives
 - Validation of other sub-models
 - Global simulation

Acknowledgements

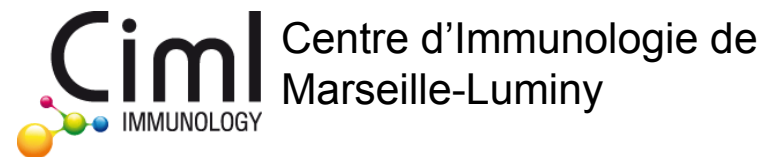


CSB team (IBENS)

- Denis Thieffry
- Geoffray Brelurut
- Olivier Collin
- Karla Corral
- Hatim El Jazouli
- Swann Floc'hlay
- Sylvie Hermann-Le Denmat
- Marika Kapsimali
- Nathalie Lehmann
- Aurélien Naldi
- Morgane Thomas-Chollier
- Pierre Vincens

Antique team (DIENS)

- Jérôme Feret



- Bernard Malissen
- Romain Roncagalli
- Guillaume Voisinne



This work is supported by grants from the Plan Cancer INSERM (SYSTAIM and CoMET projects).