Modélisation booléenne probabiliste avec MaBoSS, modélisation en population de cellules avec UpPMaBoSS

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Outline

- Population interpretation of probabilistic Boolean space
- 2. MaBoSS for probabilistic Boolean modelling
- 3. UpPMaBoSS for dynamical and heterogeneous population modelling
- 4. Example with (TNF \rightarrow cell fate) model

1. Population interpretation of probabilistic Boolean space

Boolean node state: 0 or 1 for node *i*.

<u>Boolean network state</u>: vector \vec{S} of Boolean node state for a given network. For each node *i*, $S_i \mid \{0,1\}$.

<u>Network state space</u>: set of possible Boolean network state, $\Sigma = {\vec{S}}$ for a given network.

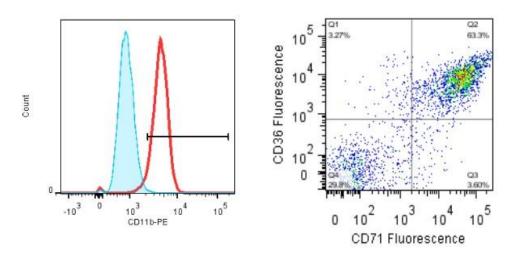
<u>Probability space over network state space (not on Boolean</u> node state): give a probability for each network space.

$$P(\vec{S}) \in [0,1], \sum_{\vec{S} \in \Sigma} P(\vec{S}) = 1$$

1. Population interpretation of probability space over network state space

Biological interpretation of network state probability: ratio of cells (among a population), that are in the given state (described by a Boolean network state)

→ Can be used to confront theoretical prediction to cytometry data.



Missing:

- Global size of cell population
- Paracrine interaction, ie population dependant interaction



2. MaBoSS for probabilistic Boolean modelling https://maboss.curie.fr/

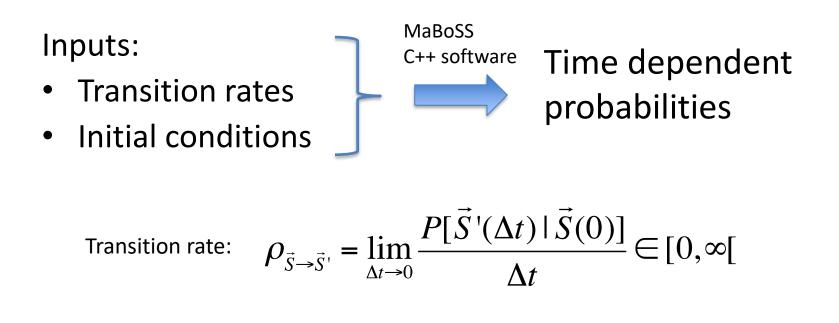
Motivation:

- Probabilistic interpretation of node/network states
- Model transient behavior
- Model asymptotic behavior
- Handle different time scale processes (transcription, phosphorylation, etc.)

→ Fill the gap between ODE and Boolean modeling: continuous Markov process on Boolean network state space

https://maboss.curie.fr/

MaBoSS: $P(\vec{S},t) \in [0,1], \sum_{\vec{S} \in \Sigma} P(\vec{S},t) = 1$ time dependent probabilities (over network states) within <u>Markov hypothesis</u>, with <u>asynchronous transitions</u> (single node transition)



Two files that define the model: a **bnd file** (*b*oolean *n*etwork *d*escription) and a **cfg file** (*c*on*f*iguration file).

1) <u>Transition rates</u>: define rate value (time⁻¹) and condition for activation and inhibition: logical operators (AND,OR, XOR, NOT) with real operators (+, -, *, /) and ternary conditional (? :), (condition ? value_if_cond_true : value_if_cond_false).

In **bnd file**:

[A,B,C,D].istate = 0.7 [0,0,0,0], 0.2 [1,0,0,0], 0.1 [1,1,1,1];

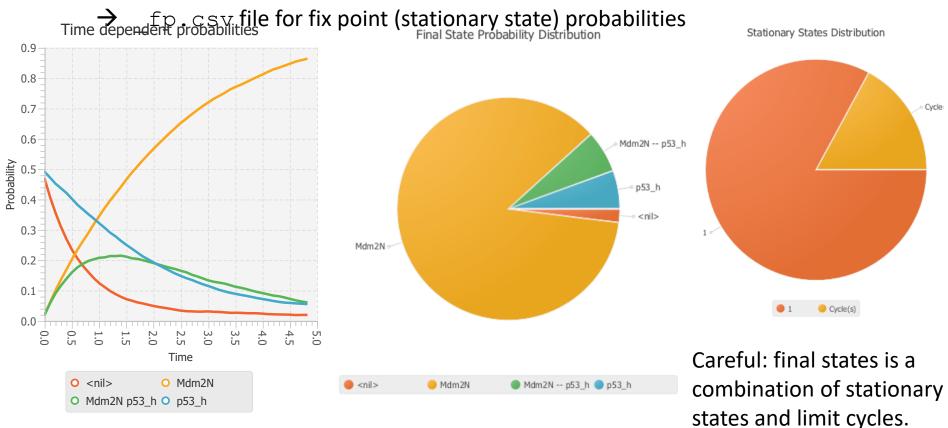
Modeling parameters in .cfg file: number of trajectories, maximum time, etc.

Core MaboSS c++ executable file (unix):

`MaBoSS -c configuration.cfg -o out_prefix network.bnd'

→ _traj.csv file for transient probabilities

 \rightarrow _statdist.csv file for stationary distribution decomposition (not used a lot)



Tool Box (MaBoSS-env, download from https://maboss.curie.fr):

- MBSS_FormatTable.pl for producing tables for spreadsheet
- MBSS_PieChart.py for pie chart of final distribution
- MBSS_TrajectoryFig.py for plot of probability trajectories
- MBSS_MutBndCfg.pl for handling mutation: add external variable in .bnd file that are controlled in .cfg file.
- MBSS_DrugSim.py for modeling drug effect

Tutorials and examples at https://maboss.curie.fr

3. UpPMaBoSS for population modeling

A Boolean model is initially based on signalling pathways inside each cell. Nodes representing cell death and cell division are added and inter-cell communications are specified.

Population dynamics using MaBoSS include:

- 1. Cell death,
- 2. Cell division,
- 3. Inter-cell communication (ligand \rightarrow receptor for instance).

Possibility to model mutant effect on population dynamics for instance, because it is based on a signalling pathway.

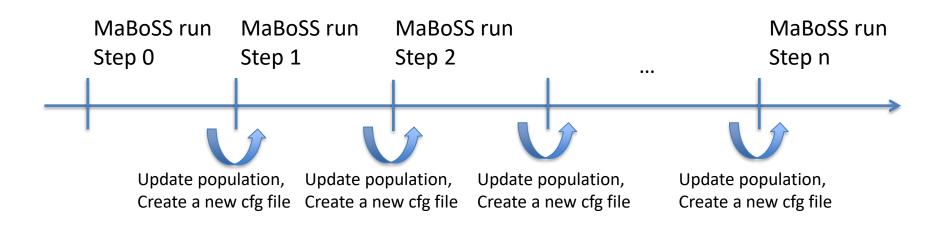
Outputs:

- size of cell population over time,
- probabilities of "network states" over time

UpPMaBoSS, "Update Population MaBoSS"

Perl script that runs MaBoSS several times ("steps"). For each MaBoSS run:

- 1. Extract the final network state probability distribution.
- 2. Create a new **cfg file** (configuration file) for the new step.



- New cfg file: take initial condition from final probability distribution of previous step, remove cell death state, double division state, renomarlize probability distribution, update external variables that represent inter-cell interaction
- Total time = (max_time of a MaBoSS run) * (number of steps).

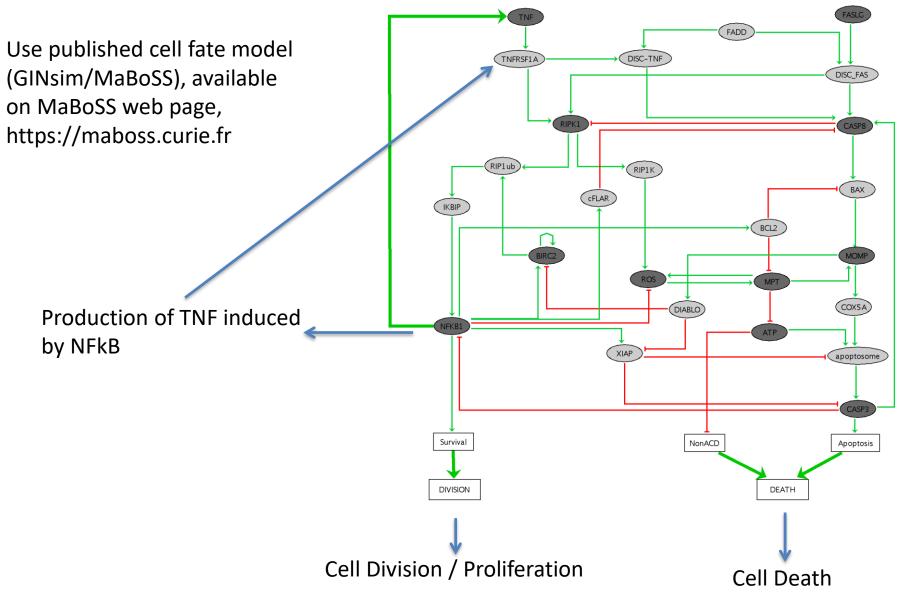
Outputs

- 1) MaBoSS output csv files (probabilities over network states over time).
- 2) Final probability distribution over network state for each step/MaBoSS run.
- 3) Population ratio for each step/MaBoSS run (updated according to probability renormalization factor)

The "population ratio" is a number that represents the relative size of the cell population compared to the initial population size. Therefore:

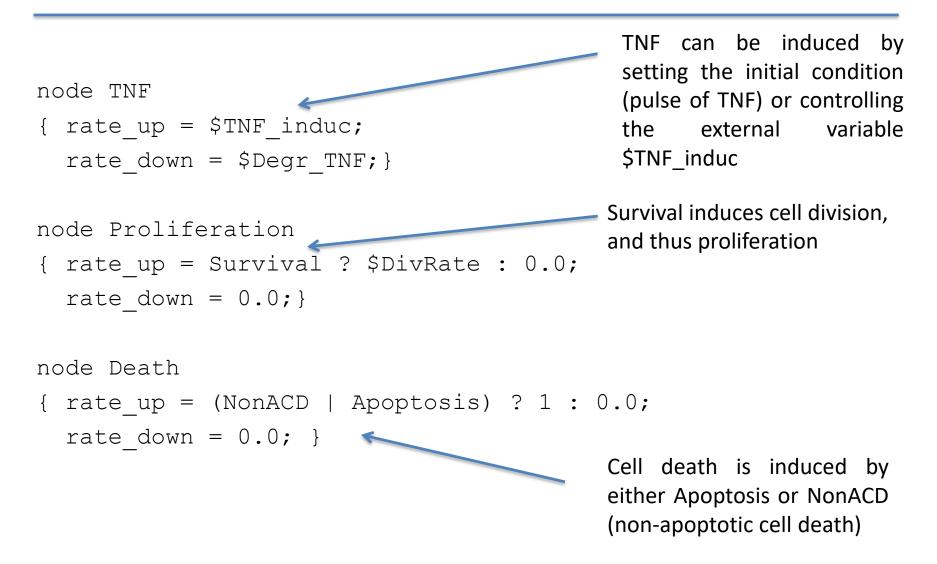
- Population ratio = $1 \rightarrow$ Stable size
- Population ratio > 1 \rightarrow Growing population
- Population ratio < 1 \rightarrow Decreasing population

Example with TNF \rightarrow cell fate



In the **bnd file**:

- add a node for cell death, a node for cell division
- modify the TNF node, in order that its activation is controlled by an external variable (\$TNF_induc).



In the initial **cfg file** (used for step 0):

- set cell division (\$DivRate) (cell death rate is 1 hour in the bnd file)
- define an external variable that control the strength of TNF production by NFkB (\$ProdTNF_NFkB)
- set carefully the max_time (tricky)
- set the initial condition for step 0

[TNF].istate = 1 [1] , 0 [0];\$ProdTNF NFkB = 1; \$TransRate = 1/24; DivRate = 1/24;Pressure Pressure PTNF induc = 0;max time = 3;

Treat cells with TNF \rightarrow initial state is 1 (shot of TNF)

Production of TNF by NFkB is switched on

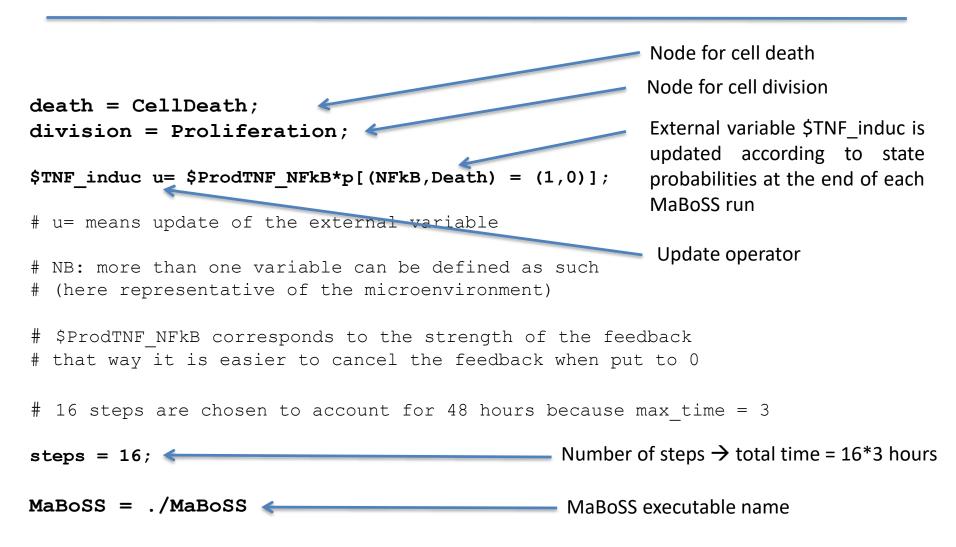
Mean transcription time = 24 hours Mean Division time = 24 hours Mean TNF degradation time = 6 hours

TNF is induced by initial condition, but this external variable will be updated (TNF permanent)

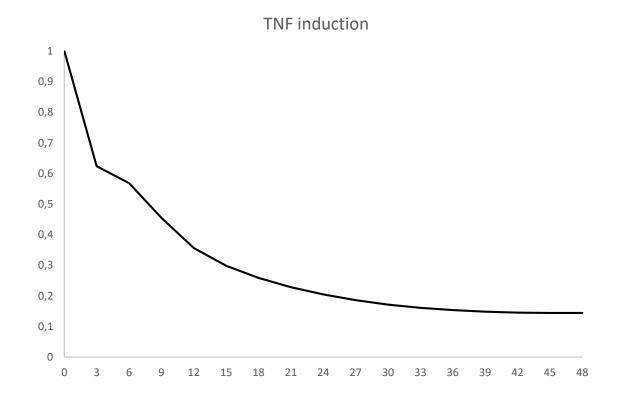
Time for population update = 3 hours (tricky to set up) In the **upp file**:

- Define the cell death node, the division node, the number of steps.

- Define how the external variable that control TNF is updated (according to NFkB status).

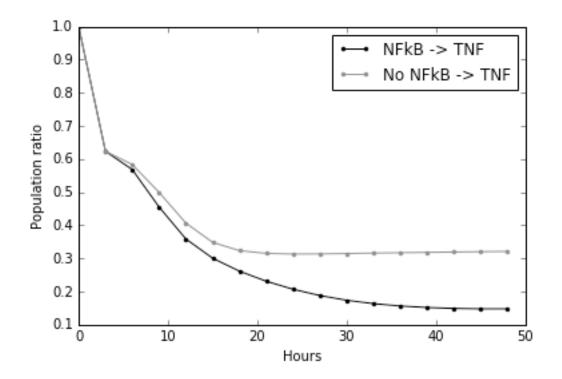


Population ratio of TNF induction model



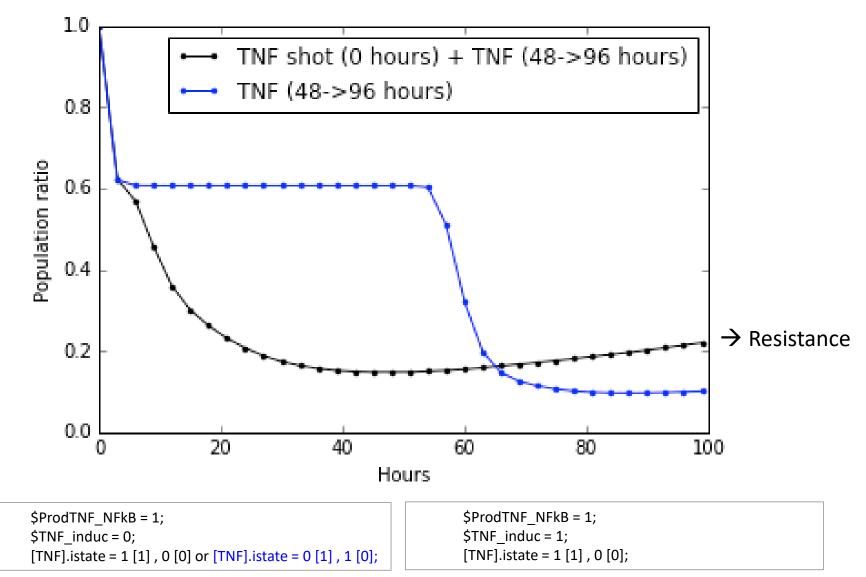
\$ProdTNF_NFkB = 1; \$TNF_induc = 0; [TNF].istate = 1 [1] , 0 [0];

Population ratio with/without paracrine loop

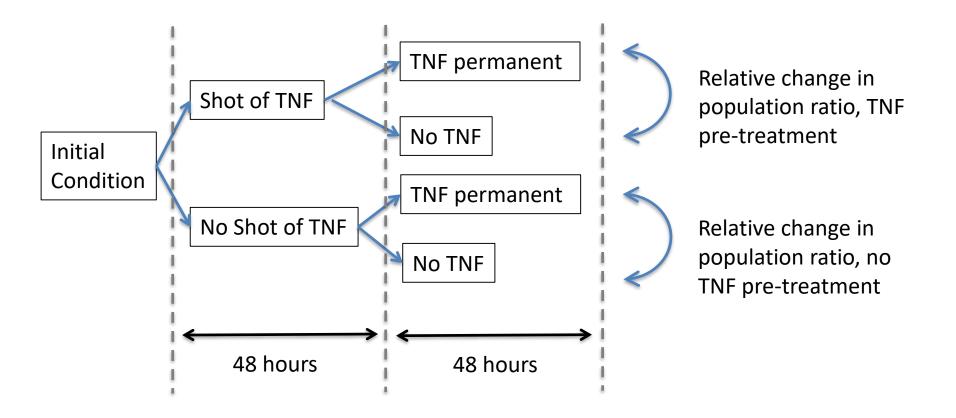


\$ProdTNF_NFkB = 1; or \$ProdTNF_NFkB = 0; \$TNF_induc = 0; [TNF].istate = 1 [1] , 0 [0];

Population ratio upon different TNF treatment protocols

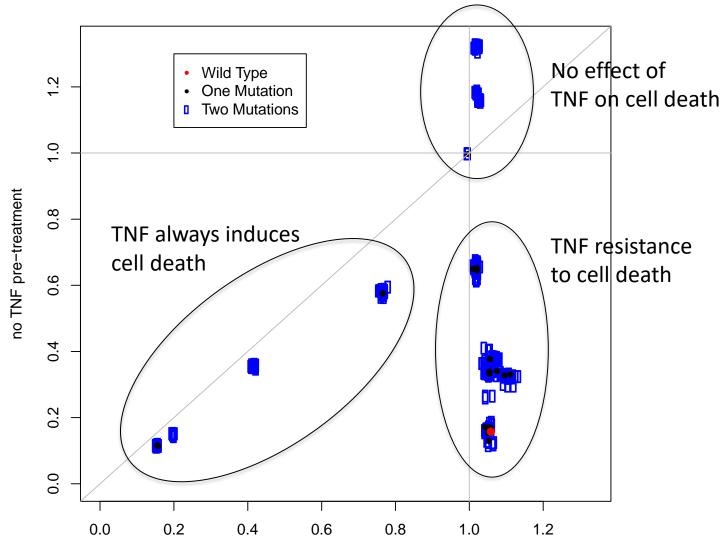


In silico "Experimental" protocol for testing TNF resistance



Simulation for every single and double mutations

Change in population ratio upon TNF treatment, single and double mutants.



TNF pre-treatment

Perspective

- Study time step size dependence
- Develop ligand \rightarrow receptor non-linear models
- Apply UpPMaBoSS.pl to Immunogenic Cell Death
- Asynchronous population update (PopMaBoSS)?

Acknoledgements

- D. Thieffry and C. Hernandez (ENS, Paris)
- E. Viara (MaBoSS developper)
- L. Calzone (Institut Curie Paris)

5. Example with T-cell activation/differentiation

Céline Hernandez (ENS, D. Thieffry group) model.

Observed phenotype (clonal expansion): - A transient activation of TCR (T-cell receptor, induced by infection) induces T-cells proliferation, that is amplified by IL2 (Interleukin-2).

IL2R

JAK3

STAT5

SOS

ERK1

FOS

JNK

AP1

CD25

NFKB

NFAT

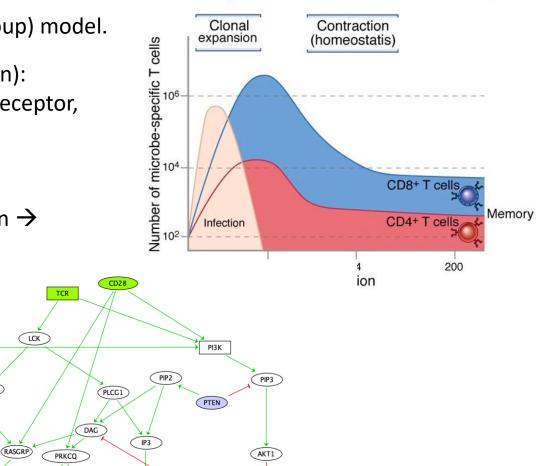
- TGFbeta induces T-cell differentiation ightarrow

TGFBR

SMAD3

FOX 23

~ 10% of Treg (FOXP3+)



DGKA

FOX01

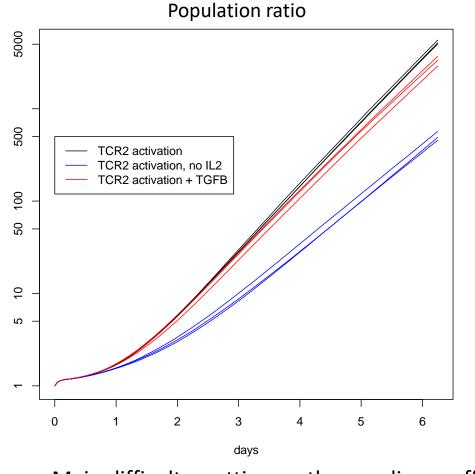
Model:

Three conditions of transient activation of TCR:

- TCR activation

population ratio

- TCR activation with no IL2 loop
- TCR activation with TFGB (\rightarrow Treg)



Main difficulty: setting up the non-linear effect ligand \rightarrow receptor

Cell populations (TCR activation + TGFB)

