

# Validation des modèles de simulation en géographie: nouvelles pratiques et perspectives

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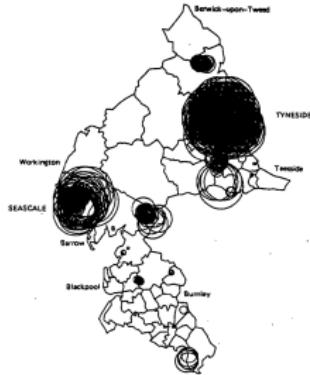
<sup>2</sup>UMR CNRS 8504 Géographie-cités

Séminaire EpisteMod

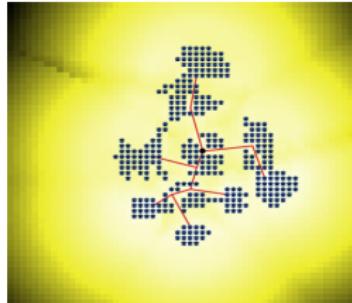
IHPHST

October 9th 2018

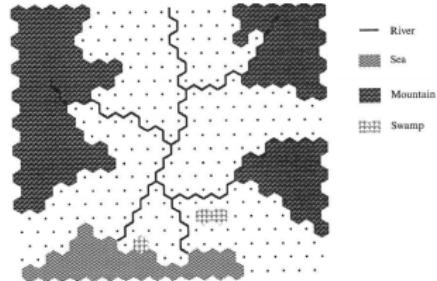
# Une longue histoire de modélisation et simulation en TQG



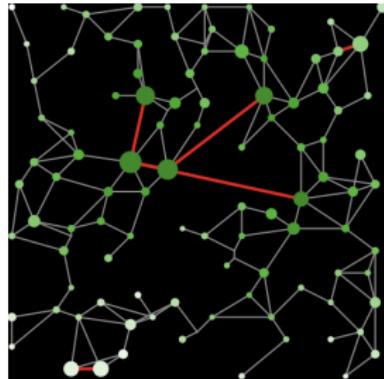
Machine d'analyse géographique  
[Openshaw et al., 1987]



Morphogenèse urbaine hybride  
[Raimbault et al., 2014]



Modèle Simpop 1 [Sanders et al., 1997]



Modèle SimpopNet [Schmitt, 2014]

# Système géographique et complexité

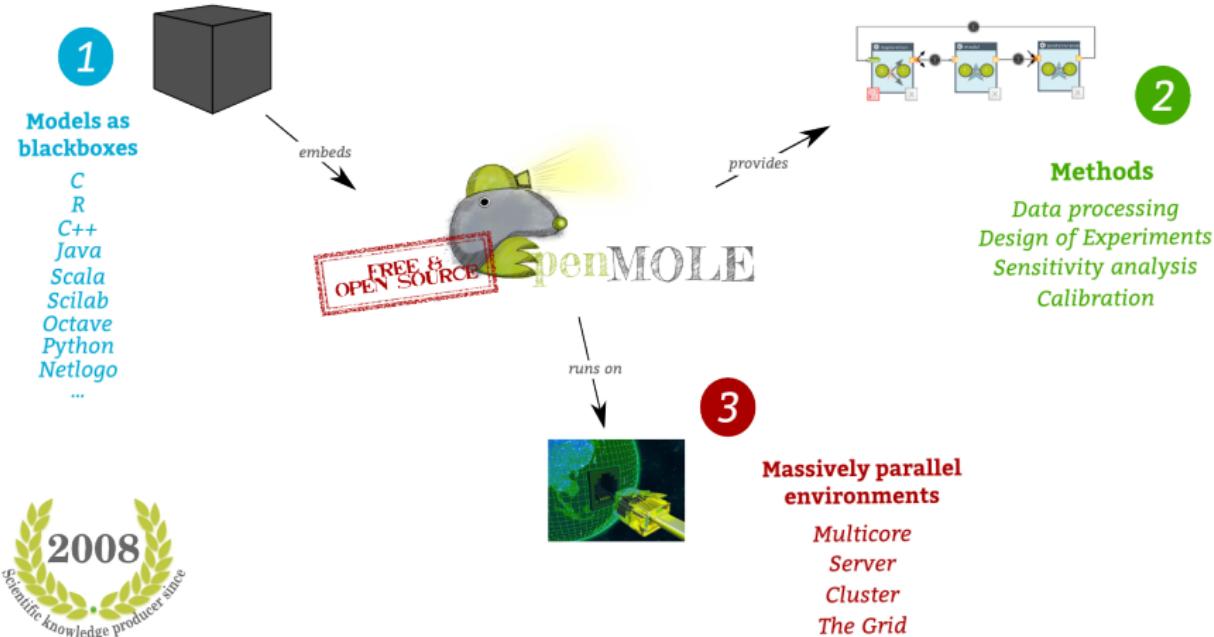
# Validation des modèles de simulation

# Vers de nouvelles pratiques: ERC Geodivercity

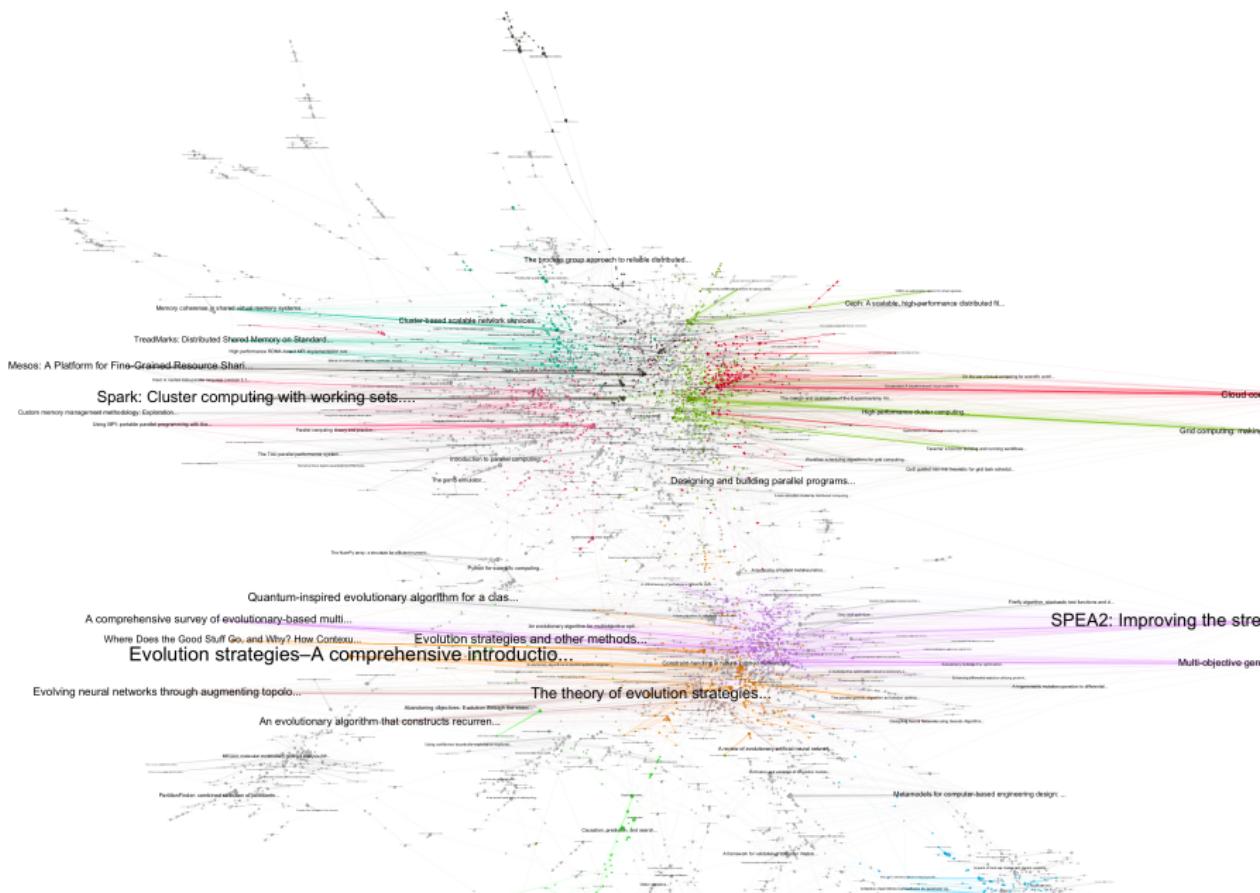
# Méthodes et outils: OpenMOLE

# Manifeste d'OpenMOLE

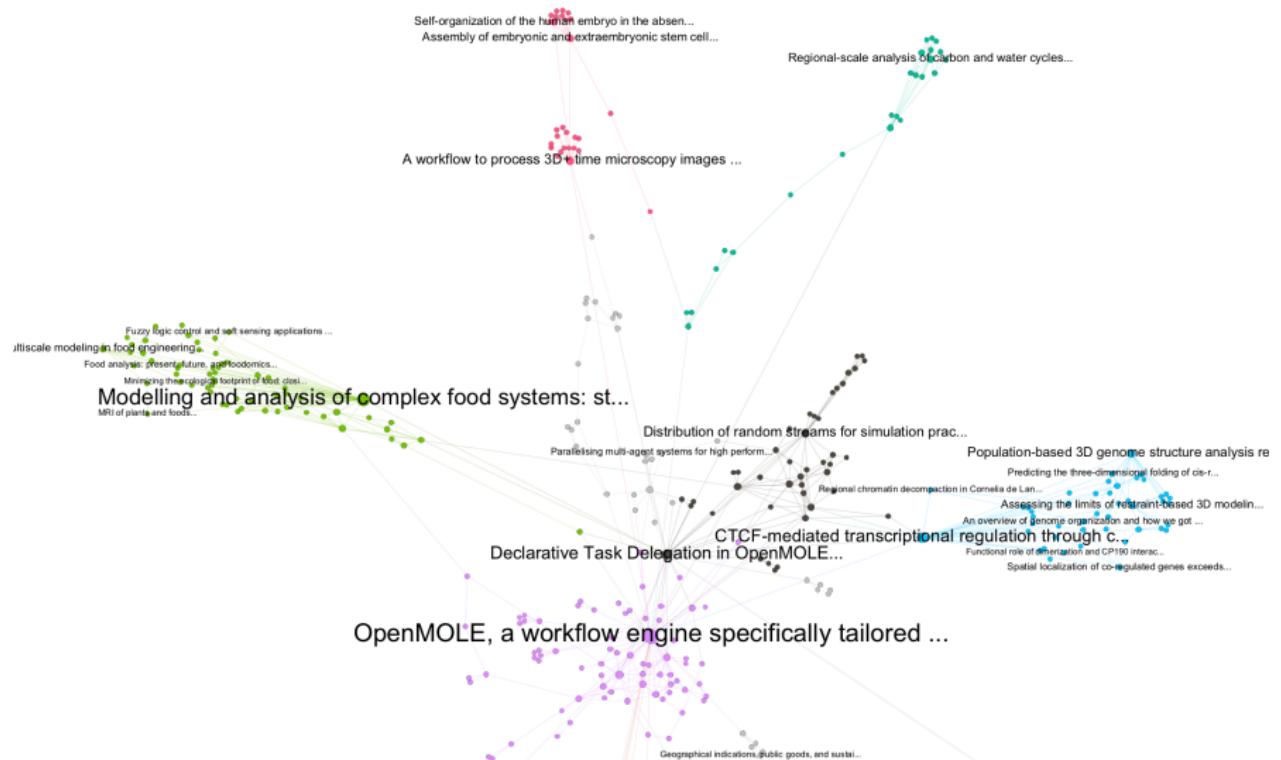
(i) Méthodes d'exploration innovantes; (ii) Passage à l'échelle par calcul distribué; (iii) Pas d'interférence avec le modèle.



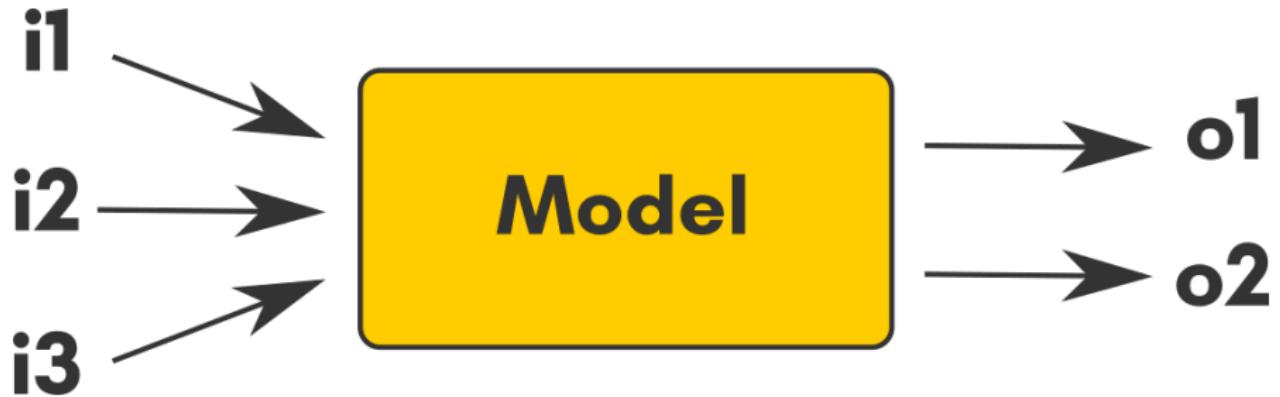
## Citation network analysis (full)



# Citation network analysis (OpenMOLE)



# Modèles de simulation



# Une interface web ergonomique

Mozilla Firefox

localhost:46857/app +

localhost:46857/app Rechercher

New project Run

Model Exploration Tutorial | calibrate

results 4.00E+0

Calibrate.oms 0.94E+0

Objective.oms 0.29E+0

fitnessVSEval.png 15.17E+0

Calibrate.oms

```
4
5 // Execute the workflow
6 // Define the population (10) and the number of generations (100).
7 // Define the inputs and their respective variation bounds.
8 // Define the objectives to minimize.
9 // Assign 1 percent of the computing time to reevaluating
10 // parameter settings to eliminate over-evaluated individuals.
11 val nsga2 =
12   NSGA2(
13     mu = 50,
14     genome = Seq(
15       diffusion in (0.0, 99.0),
16       evaporation in (0.0, 99.0)),
17     objectives = Seq(deltaFood),
18     stochastic = Stochastic(seed = seed, aggregation = Seq(median))
19   )
20
21 val evolution =
22   SteadyStateEvolution(
23     algorithm = nsga2,
24     evaluation = ants -- objective,
25     parallelism = 10,
26     termination = 100
27   )
28
29 // Define a hook to save the Pareto frontier
30 val savePopulationHook = SavePopulationHook(evolution, workDirectory / "results")
31
32 // Plug everything together to create the workflow
33 (evolution hook savePopulationHook)
34
35
36
37
```

built the 01/05/2018 18:51:28

C  
R  
C++  
Java  
Scala  
Scilab  
Octave  
Python  
Netlogo  
...



# Modèle NetLogo

```
val model =  
  NetLogo6Task(  
    workDirectory / "Fire.nlogo",  
    List("setup", "while [any? turtles] [go]")) set (  
      inputs += seed,  
      outputs += (seed, density),  
      inputs += density mapped "density",  
      outputs += burned mapped "burned-trees"  
    )
```

## Code R

```
val i = Val[Int]

val rTask =
  RTask("""
    source("function.R")
    function(i)""") set (
      resources += workDirectory / "function.R",
      inputs += i
    )
```

# Code Python en package

## Package

```
care -o python.tgz.bin python matrix.py data.csv 1 out.csv
```

## Run

```
val pyTask =  
    CARETask(  
        workDirectory / "../python.tgz.bin",  
        "python matrix.py data.csv 2 out.csv"  
    )
```

# Méthodes

- Reconstruction de données
- Estimation de paramètres
- Analyse de sensibilité
- Étude de robustesse
- Optimisation

*Conçues pour passer à l'échelle, prendre en compte la stochasticité, être utilisable sur n'importe quel modèle et environnement de calcul.*

# Méthodes: scripting

```
DirectSampling(  
    evaluation = myModel,  
    sampling =  
        LHS(  
            500,  
            diffusion in (10.0, 100.0),  
            evaporation in (10.0, 100.0)  
        )  
)
```

# Environnements de calcul

# Passage à l'échelle

```
val cluster = SLURMEnvironment("login", "cluster.domain.org")

DirectSampling(
    evaluation = myModel on cluster,
    sampling =
        LHS(
            500,
            diffusion in (10.0, 100.0),
            evaporation in (10.0, 100.0)
        )
)
```

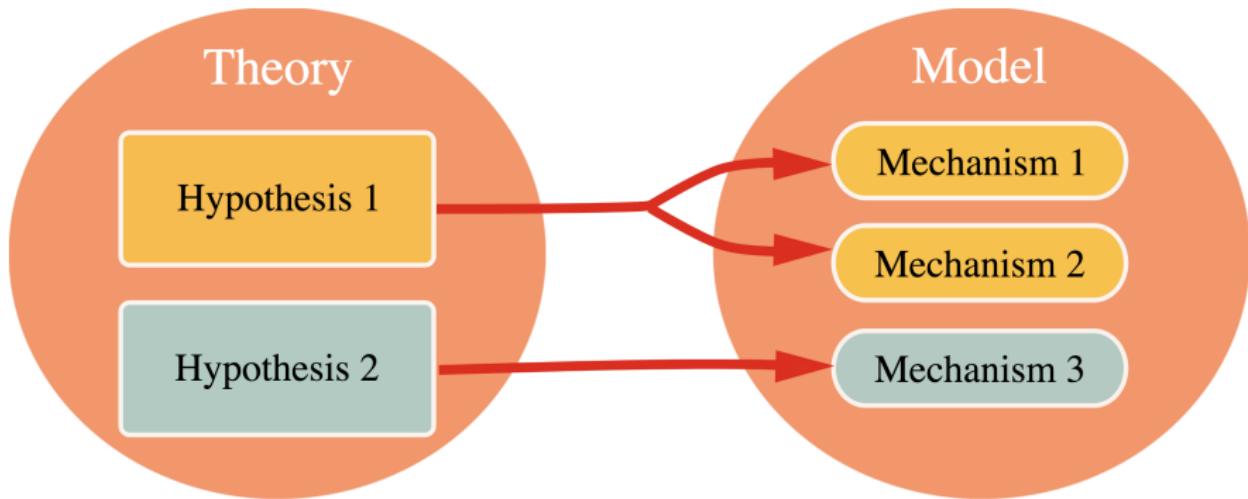
# Environnements supportés

Supported environments today  
Multi-thread Delegation through SSH  
PBS SLURM Condor SGE OAR EGI Grid

Computer assisted modeling Build a general framework and algorithms to guide the modeling process.

Make the modeling process:

traceable: understand what choices were made, reproducible: recompute the reason why they were made, reusable: study alternative choices.

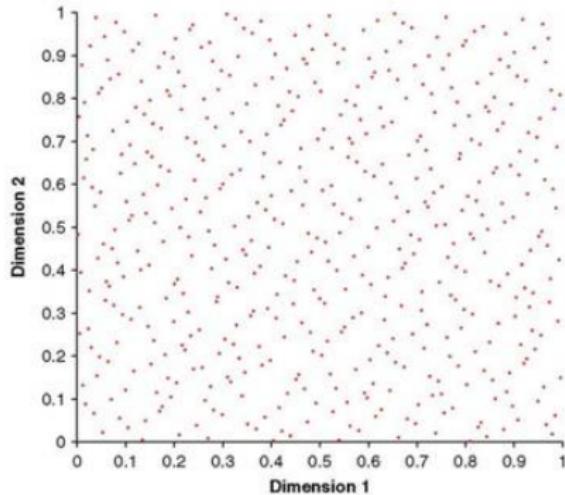


Descriptive models Design/evaluate a theory involving causal effects through its capacity to (re-)produce some patterns/data.

Model evaluation How to assess:  
the sufficiency of the mechanisms ? the necessity of the mechanisms ?  
the uniqueness of the mechanisms ?

# Sufficiency

The usual approach Use a classical design of experiment.

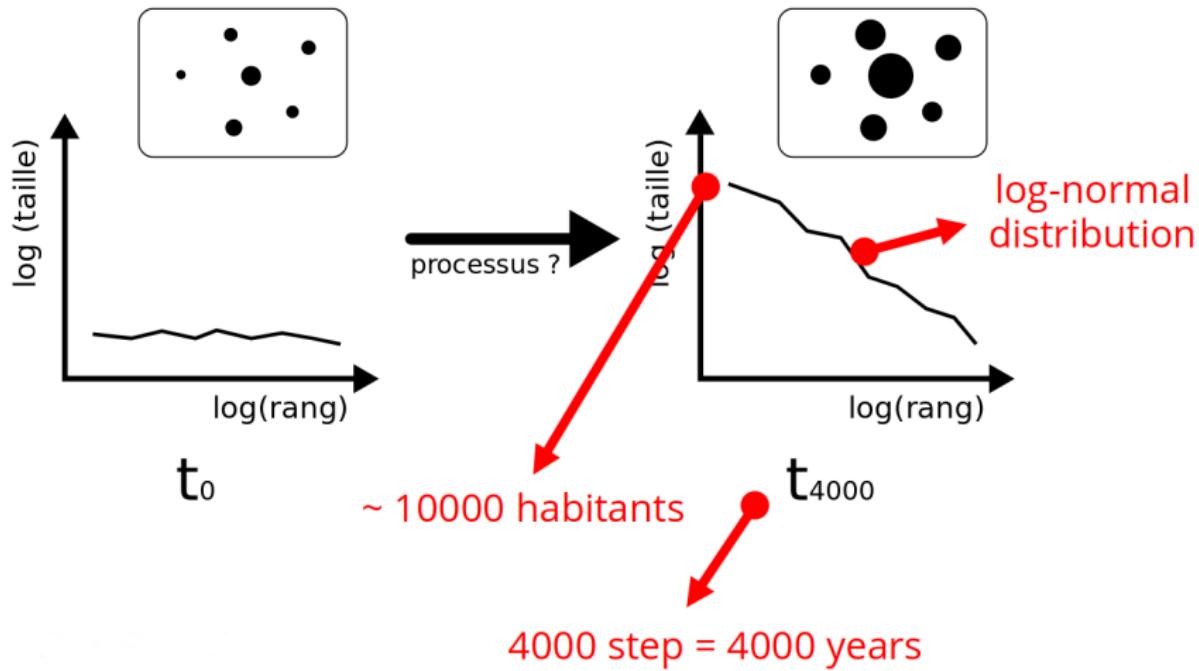


Limits It produces a huge quantity of data. It transforms a problem into a data-mining problem. The parameter space is mostly left unexplored due to the curse of dimensionality.

# Proposed method

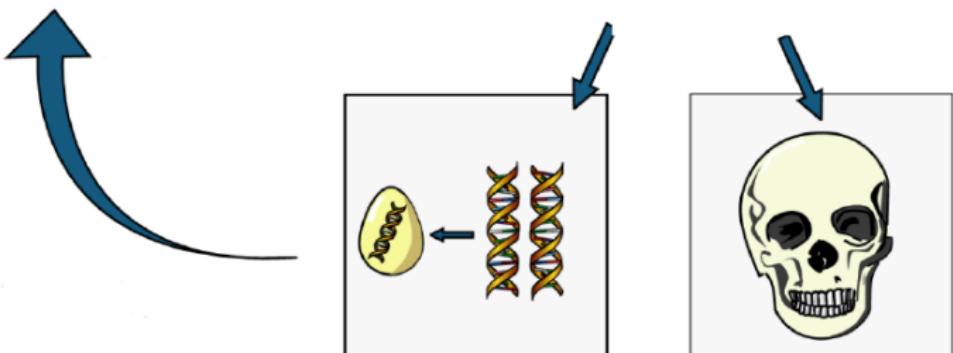
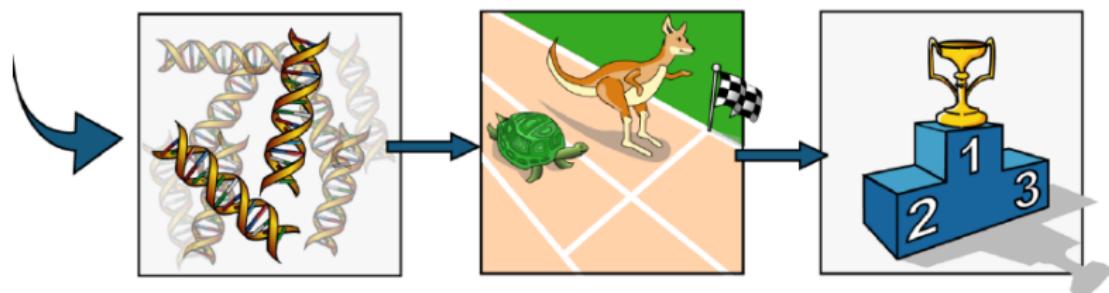
Reverse approach: From output values to parameter values.

Formalising the expectations



# Calibration algorithm

Genetic algorithm as a calibration algorithm



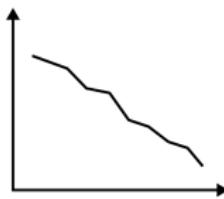
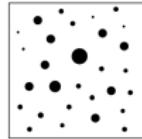
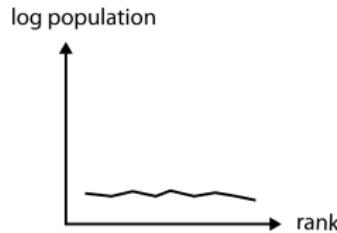
Calibration method Genome: 6 parameters of the model

Fitness: 3 descriptors of the quality of the simulated dynamics

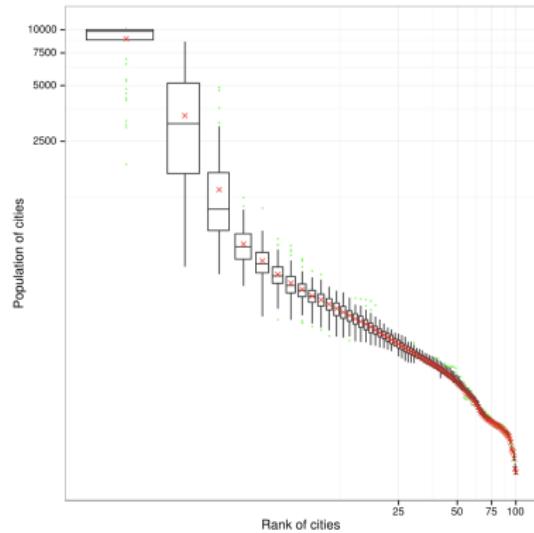
# Calibration results

No compromise between the 3 objectives.

Searched pattern



Produced pattern



Evaluation approach Would not have been found using a direct method.

The method is tractable (even for ABMs):

Handles stochasticity: 100x gain. Support for distributed computing:

1000x gain.

# Necessity

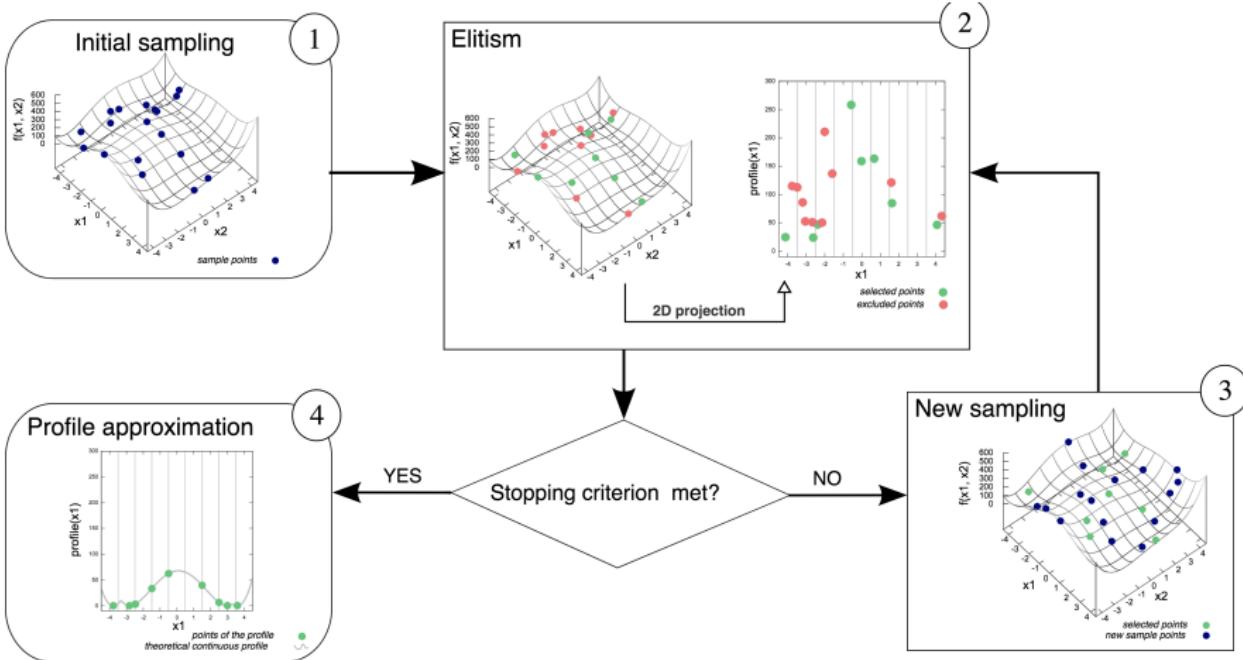
A new algorithm To detect if a parameter is useful: it impacts the capacity of the model to produce plausible outcomes.

To better constrain the parameter ranges.

As an indirect way to detect if some of the mechanisms are expandable

Objective Hundreds of calibrations

# Algorithme des profils

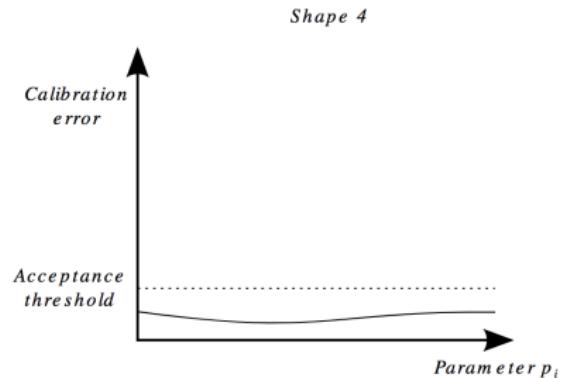
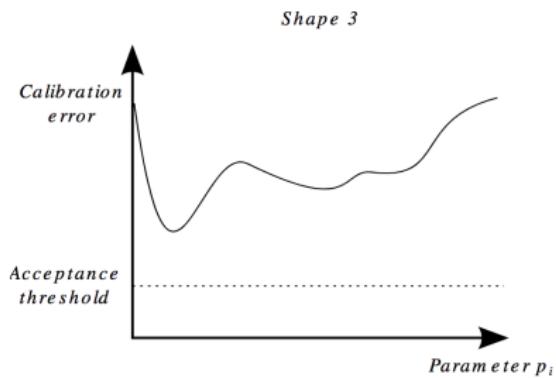
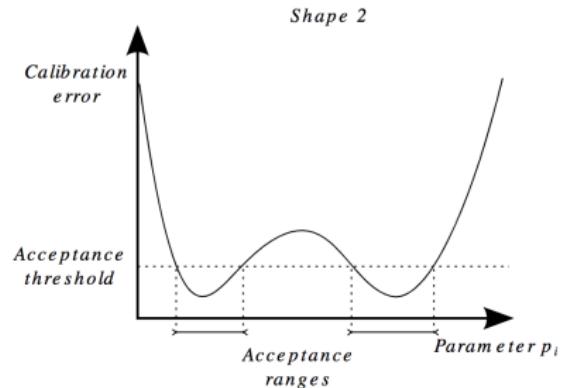
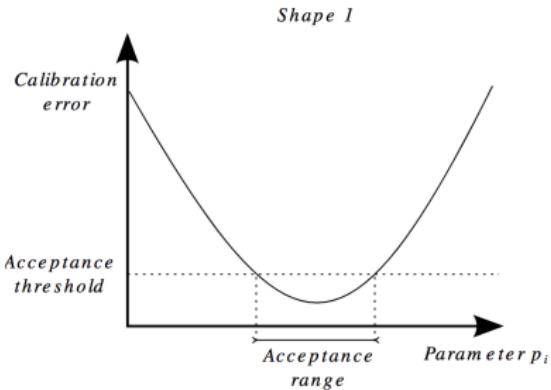


The profile algorithm Compute the best of calibration for hundreds of values along the definition domain of a parameter.

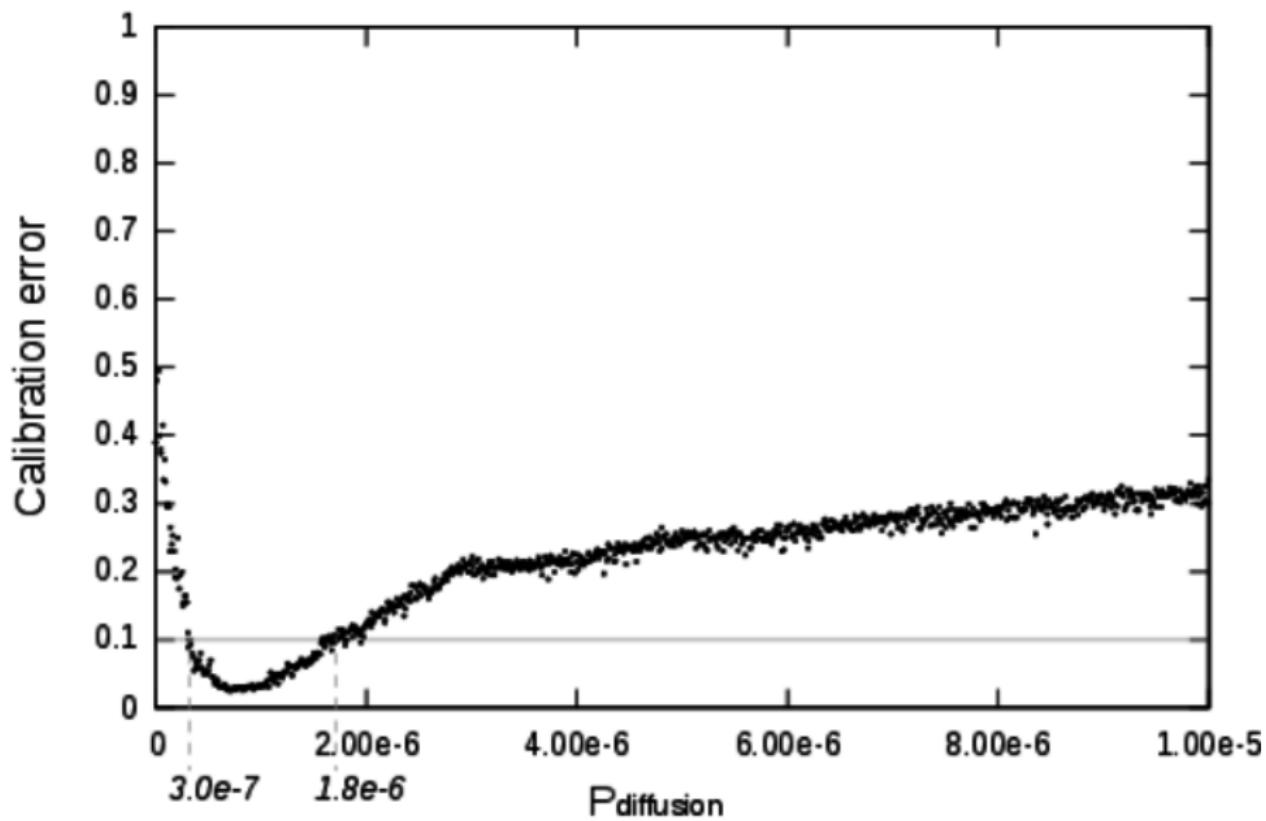
The method is tractable (even for ABMs):

Converges almost as fast a single calibration. Handles stochasticity: 100x gain. Support for distributed computing: 1000x gain.

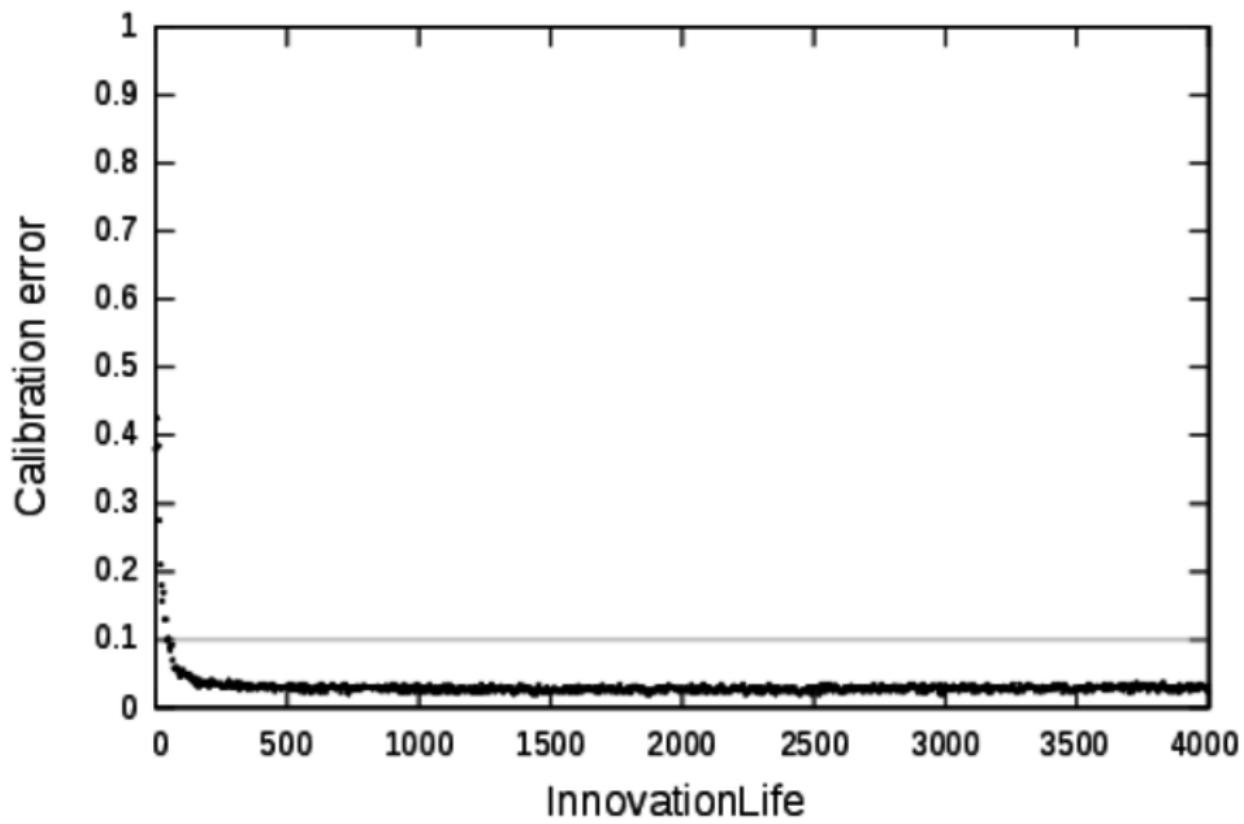
# Algorithme des profils



## Results



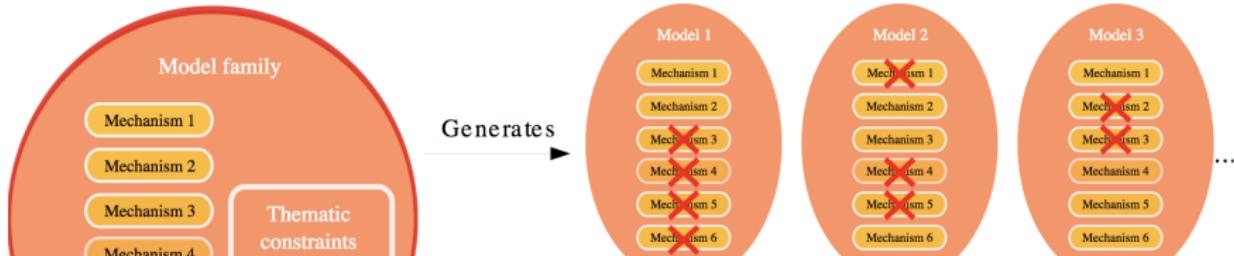
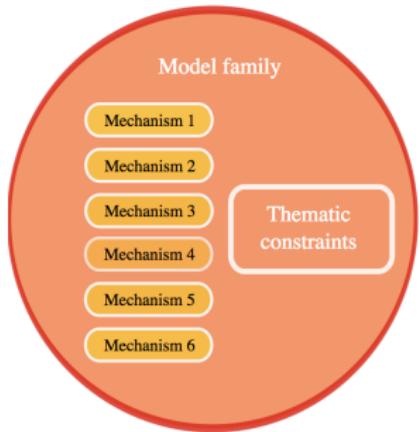
## Results



# Uniqueness

Automate the confrontation of alternative hypothesis / mechanisms.

*Thematic hypothesis*

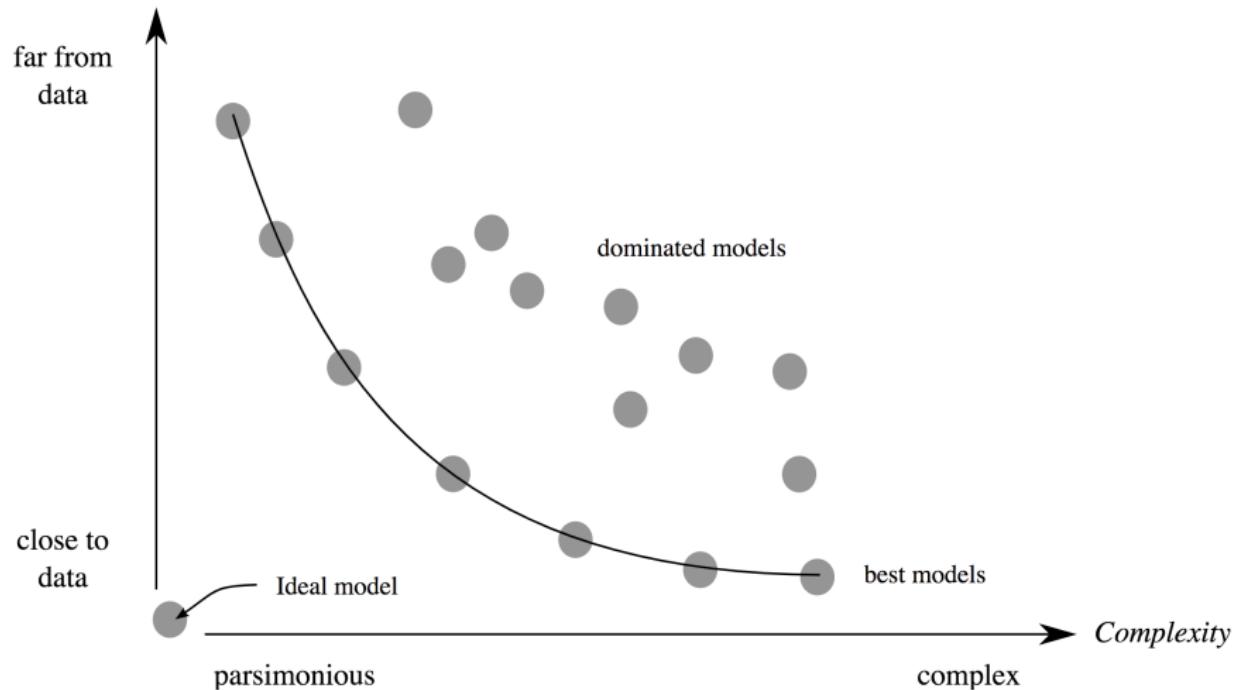


New calibration algorithm  
designed to calibrate  
a model family

# Objective

What we are running after?

*Distance to data*

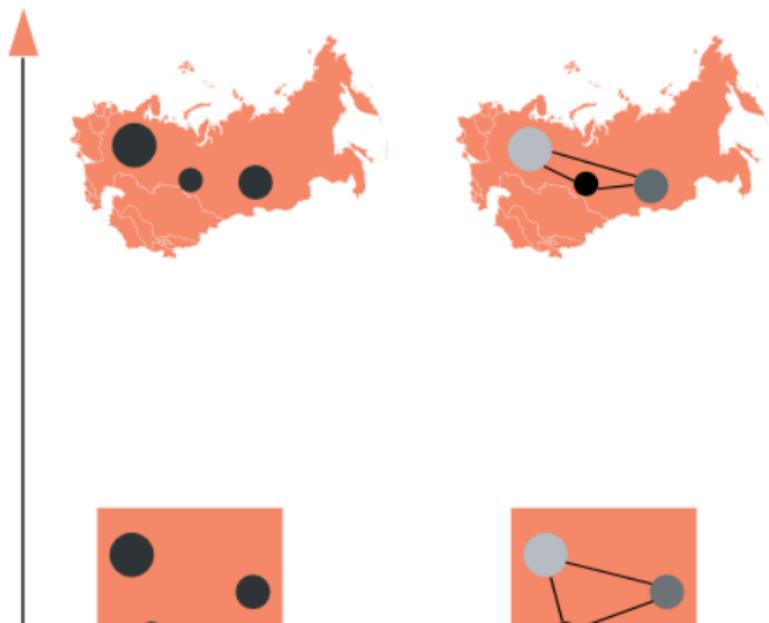


# Multi-modeling

64 models

## Axe 2

Interactions entre  
les agents et  
l'environnement



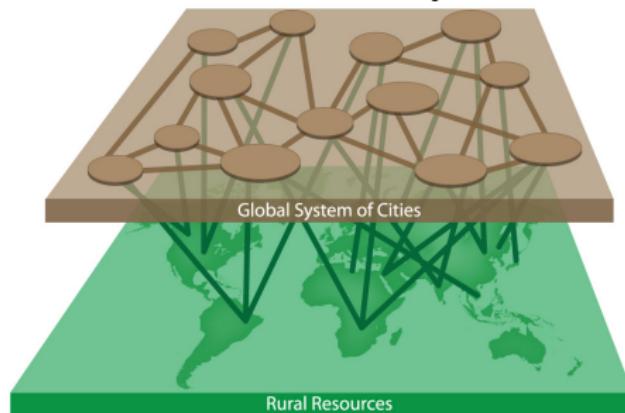
# Example of concurrent hypotheses

Exchange mechanism:

Market based Centrally planned

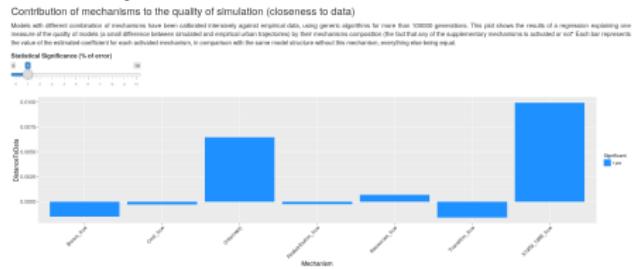
City growth mechanism:

Purely inter-urban interactions Influenced by environmental situation



# Model family calibration

Compute the best set of parameters for all 64 models.



# Calibration

Calibration method Algorithm principle: reuse the good parameters found on one model to calibrate on the others.

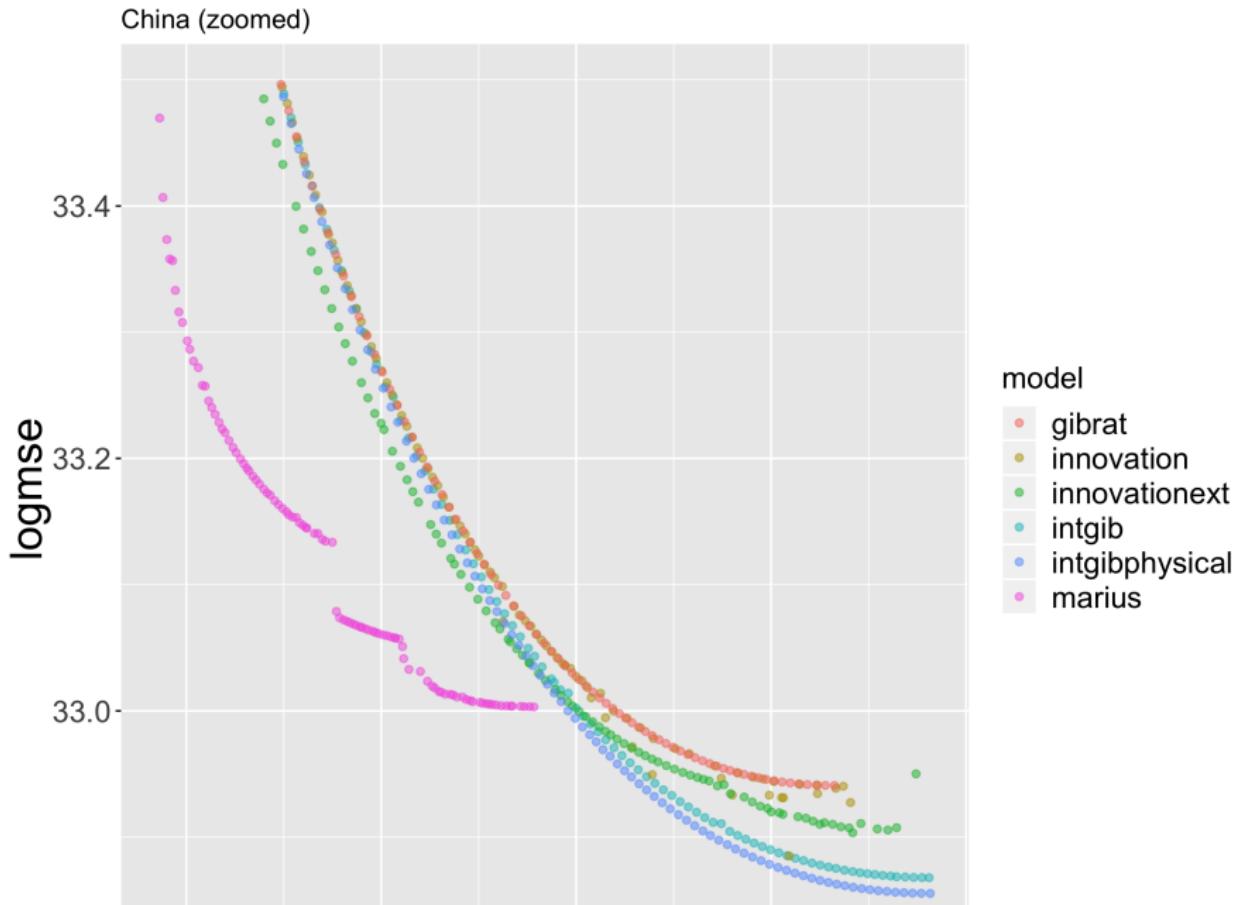
The method is tractable (even for ABMs):

as fast (faster ??) as calibrating a single model. Handles stochasticity: 100x gain. Support for distributed computing: 1000x gain.

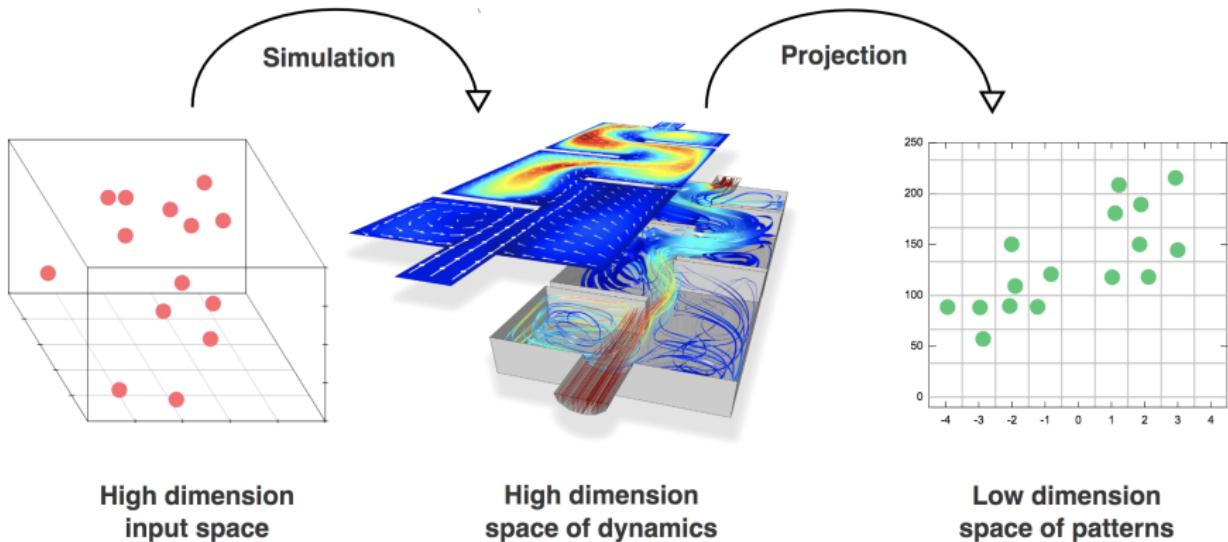
Alternative approach All these methods are based on an a priori measurement of the quality.

Can we propose an alternative approach to evaluation?

# Autre exemple de multimodélisation

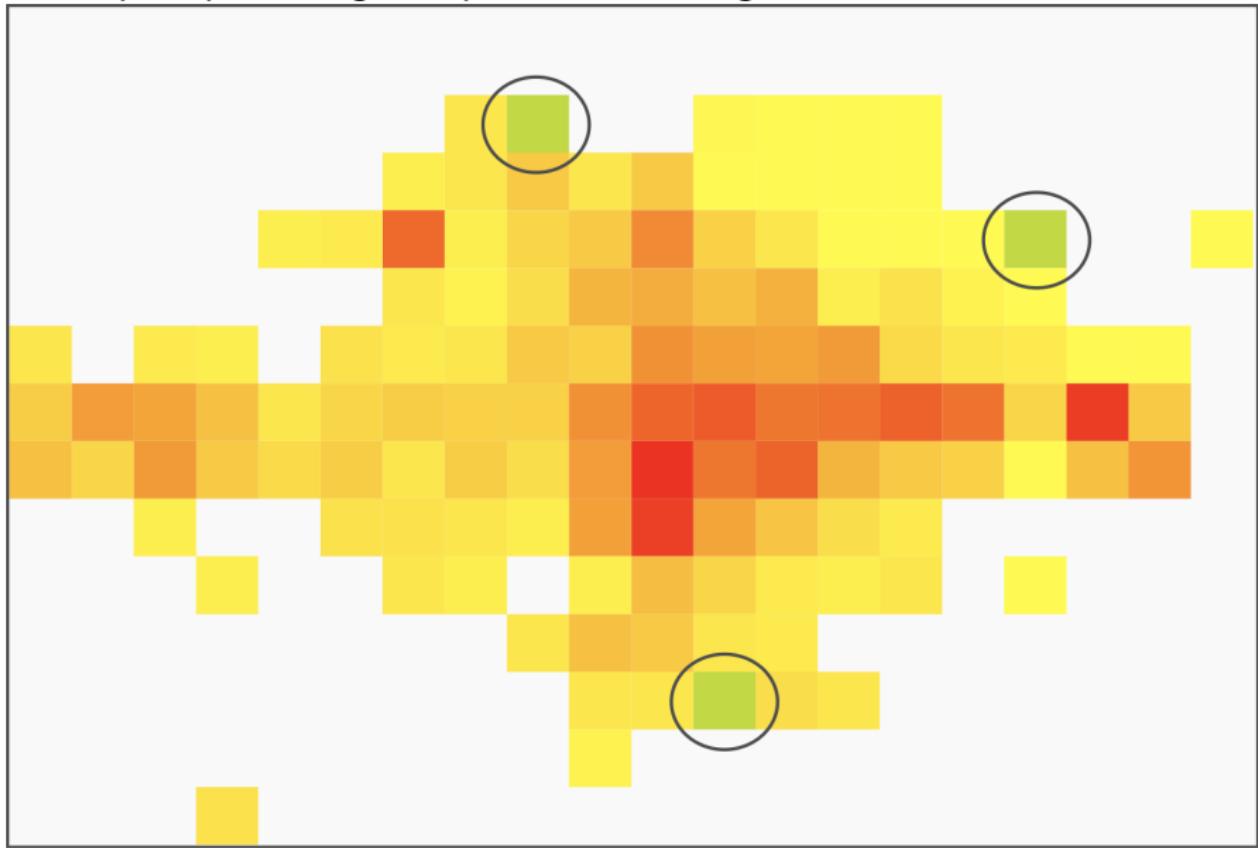


# Pattern search



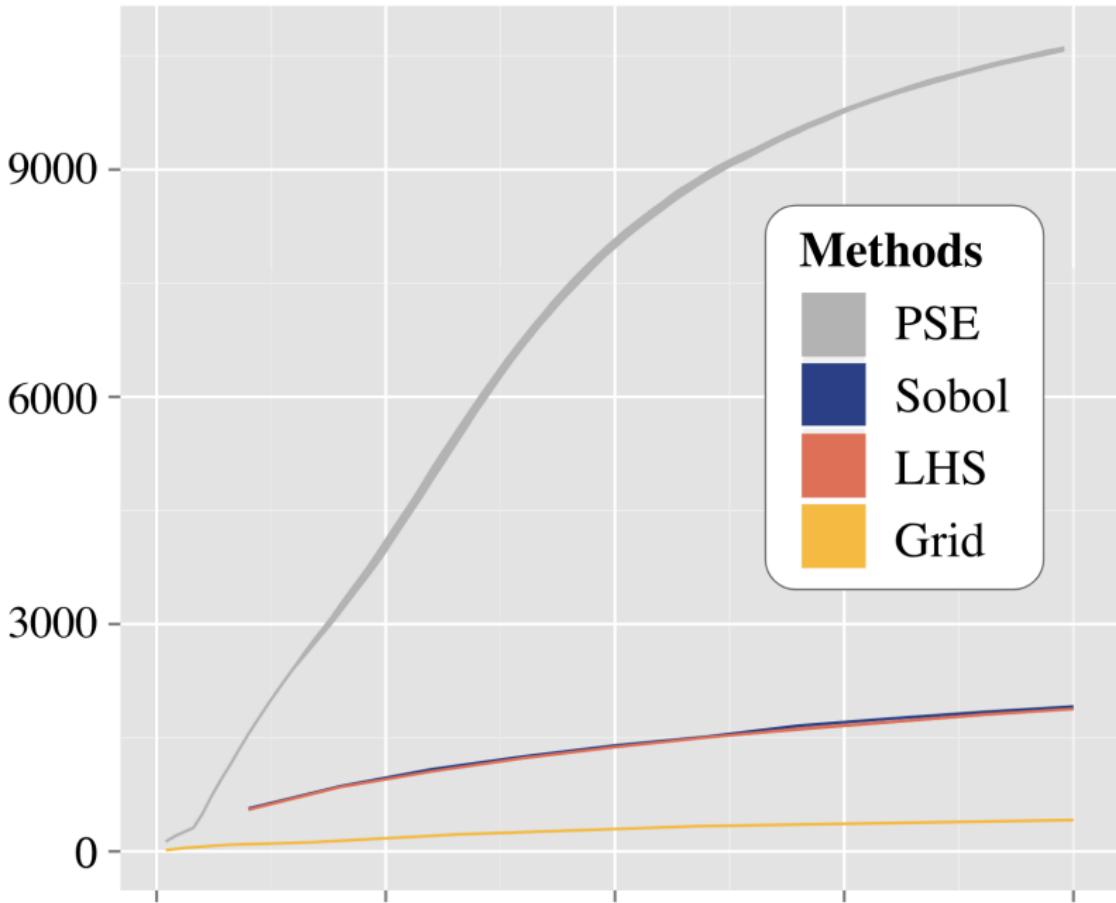
# Recherche de nouveauté

The inputs producing rare patterns have high fitness values.

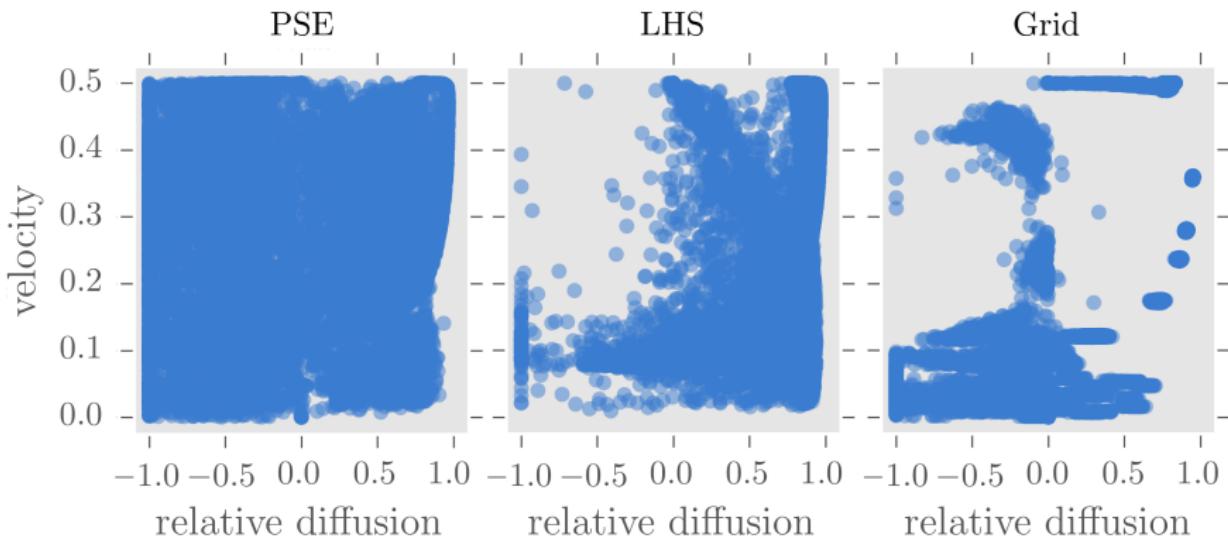


# Résultats

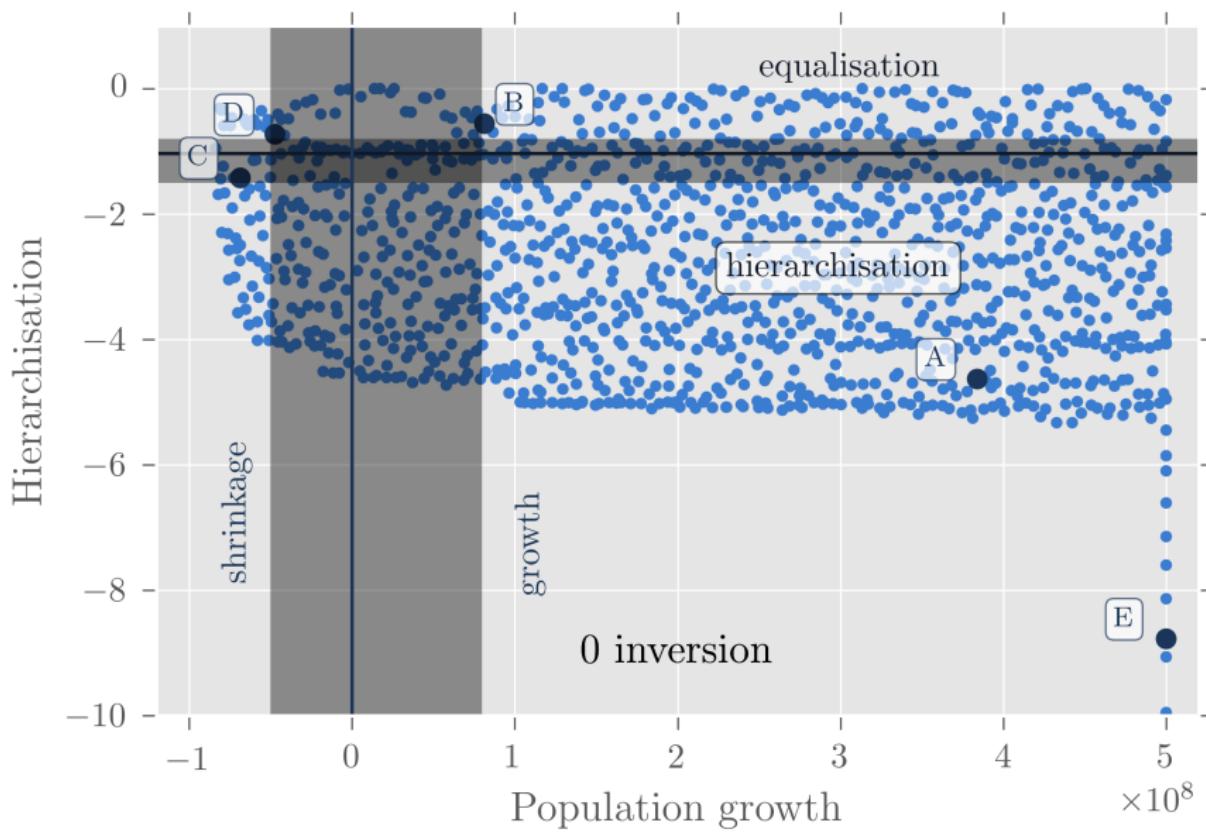
Total number of patterns discovered



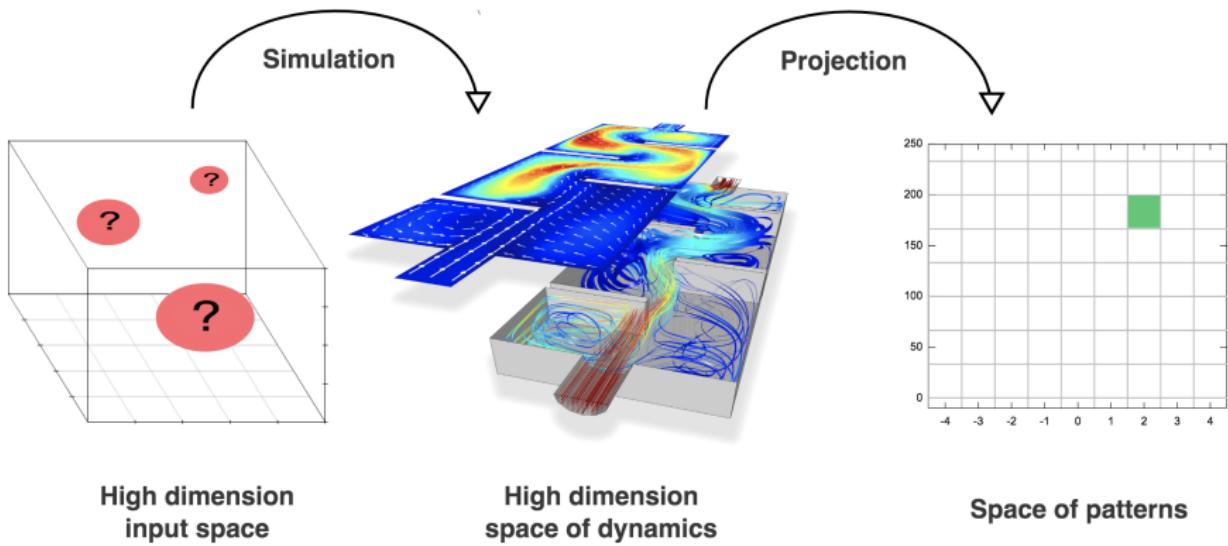
# Résultats



# Résultats



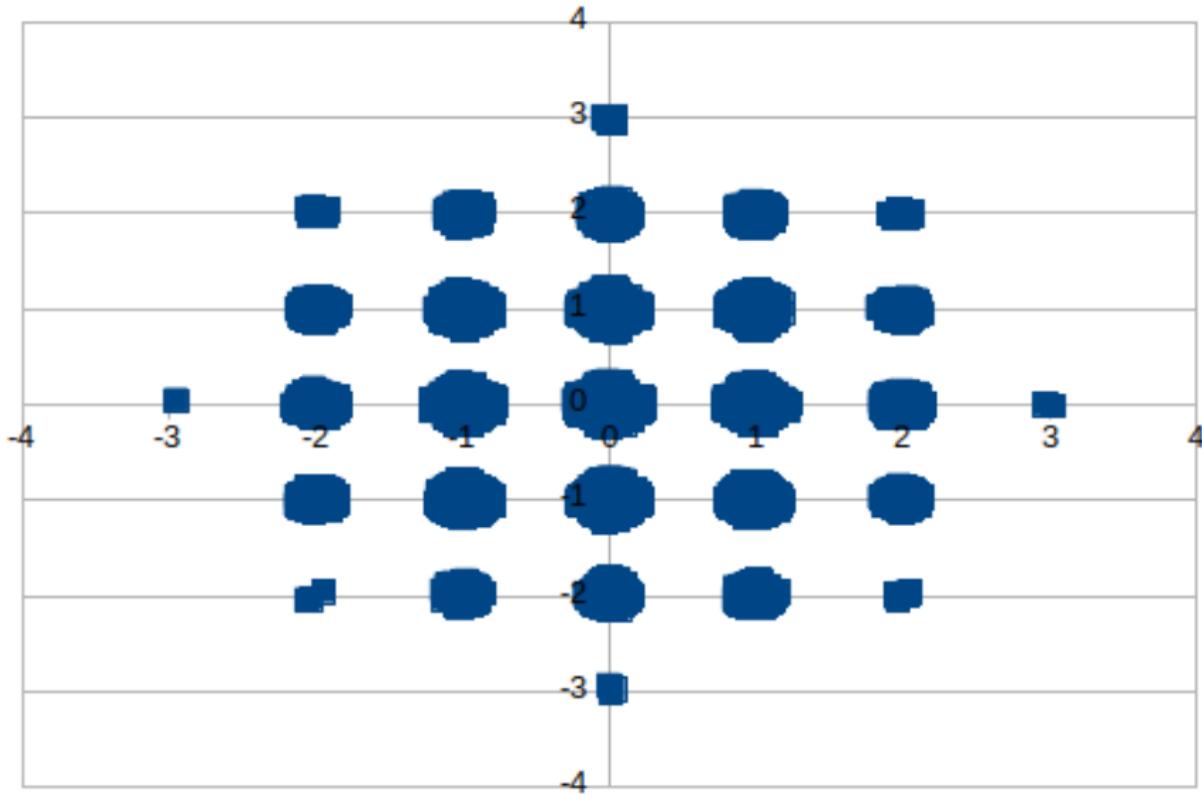
# Origin search



# First results

Formulation:  $\Delta$  pattern  $< \varepsilon$

$$f(x_1, x_2, \dots, x_n) < 10$$



# OpenMole's positioning

*A qualitative shift in knowledge that can be extracted from a simulation model with model exploration methods.*

Success stories: SimpopLocal [Schmitt et al., 2015], Marius [Chérel et al., 2015], Ecological modeling [Lavallée et al., 2018], epidemiology [Arduin, 2018], etc.

## Key features:

- Unique role of complementary axis of computation environment access, methods providing, and model embedding.
- Iterative and integrated construction of models and theories, using all dimensions of knowledge enhanced by simulation and computation (modeling, theory, empirical, data, methods, tools [Raimbault, 2017]).
- Coupling models and reproducibility at the core of the workflow approach [Passerat-Palmbach et al., 2017]

# Perspectives and questions ouvertes pour la simulation en TQG

## Question générales:

- Surajustement des modèles de simulation
- Couplage des modèles
- Direct and inverse mapping
- Stochasticité

## Spécifiques aux systèmes spatio-temporels:

- Non-stationnarité spatio-temporelle
- Données synthétiques spatio-temporelles

# Surajustement

# Un perspectivisme appliqué pour coupler les approches de modélisation

Perspectivisme de Giere [Giere, 2010]: a “third way” beyond constructivism-realism: *Any scientific knowledge construction process as a perspective by an agent to answer a purpose with a media (model).*

Applied knowledge framework proposed by [Raimbault, 2017] to study [complex] systems: *co-evolution* of cognitive agents and knowledge domains, through the intermediate of perspectives.

## Applied perspectivism principles:

- Foster consistence of perspectives and their communication (Banos' virtuous spiral between disciplinarity and interdisciplinarity)
- Importance of reflexivity to ease the coupling of perspectives
- New model exploration methods increase the integration between knowledge domains
- Coupling of models as a possible medium to couple perspectives (transfer hypothesis)

*Still to be formalized, specified as possible implementations, and experimented.*

# Conclusion

*Significant accomplishments beyond disciplines, construction of new research practices (see the satellite presentations) → still much to do ? (e.g. how to put into practice ? how to achieve true integration ? etc.)*

## **Related works on epistemological considerations**

Raimbault, J. (2017, December). An Applied Knowledge Framework to Study Complex Systems. In Complex Systems Design & Management (pp. 31-45). arXiv:1706.09244.

Raimbault, J. (2018). Caractérisation et modélisation de la co-évolution des réseaux de transport et des territoires (Doctoral dissertation, Université Paris 7 Denis Diderot). <https://halshs.archives-ouvertes.fr/tel-01857741>

# Reserve Slides

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*Modélisation mathématique des interactions entre pathogènes chez l'hôte humain: Application aux virus de la grippe et au pneumocoque.*  
PhD thesis, Université Paris-Saclay.
-  Chérel, G., Cottineau, C., and Reuillon, R. (2015).  
Beyond corroboration: Strengthening model validation by looking for unexpected patterns.  
*PLoS ONE*, 10(9):e0138212.
-  Giere, R. N. (2010).  
*Scientific perspectivism.*  
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-  Lavallée, F., Alvarez, I., Dommangeat, F., Martin, S., Reineking, B., and Smadi, C. (2018).  
A dynamical model for the growth of a stand of japanese knotweed including mowing as a management technique.  
In *Conference on Complex Systems 2018*.
-  Openshaw, S., Charlton, M., Wymer, C., and Craft, A. (1987).  
A mark 1 geographical analysis machine for the automated analysis of point data sets.  
*International Journal of Geographical Information System*, 1(4):335–358.
-  Passerat-Palmbach, J., Reuillon, R., Leclaire, M., Makropoulos, A., Robinson, E. C., Parisot, S., and Rueckert, D. (2017).  
Reproducible large-scale neuroimaging studies with the openmole workflow management system.  
*Frontiers in neuroinformatics*, 11:21.

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-  Rimbault, J. (2017).  
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In *Complex Systems Design & Management*, pages 31–45.
-  Rimbault, J., Banos, A., and Doursat, R. (2014).  
A hybrid network/grid model of urban morphogenesis and optimization.  
In *4th International Conference on Complex Systems and Applications*, pages 51–60.
-  Sanders, L., Pumain, D., Mathian, H., Guérin-Pace, F., and Bura, S. (1997).  
Simpop: a multiagent system for the study of urbanism.  
*Environment and Planning B*, 24:287–306.

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Schmitt, C. (2014).

*Modélisation de la dynamique des systèmes de peuplement: de SimpopLocal à SimpopNet.*

PhD thesis, Paris 1.



Schmitt, C., Rey-Coyrehourcq, S., Reuillon, R., and Pumain, D. (2015).

Half a billion simulations: Evolutionary algorithms and distributed computing for calibrating the simpoplocal geographical model.

*Environment and Planning B: Planning and Design*, 42(2):300–315.