

Quantitative Abstractions For Collective Adaptive Systems

Mirco Tribastone and Andrea Vandin

SFM'16 – Bertinoro

24 June 2016

- ERODE: A tool for the Evaluation and Reduction of ODEs
 - Frontend
 - Core
 - Reduction Techniques
- Case Studies
 - Large case studies
 - Continuous time Markov chains from MRMC distribution
 - Chemical Reaction Networks from the literature
 - Spatial dynamics
 - Crowd Dynamics
 - Multi-community Epidemiology

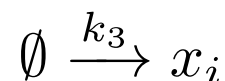
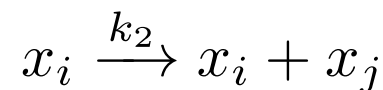
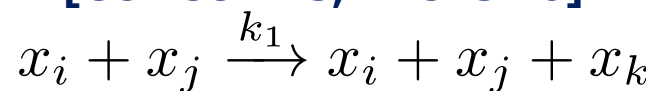
IDOL: Intermediate Drift-oriented Language [POPL'16]

$$p ::= \varepsilon \mid \dot{x}_i = f, p$$

$$f ::= n \mid x_i \mid f + f \mid f \cdot f \mid f^{\frac{1}{m}}$$

Reaction Networks

[CONCUR'15, TACAS'16]



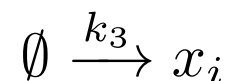
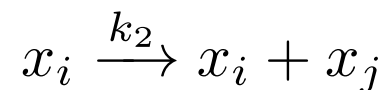
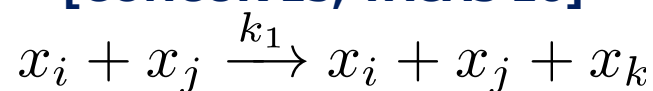
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$$p ::= \varepsilon \mid \dot{x}_i = f, p$$

$$f ::= n \mid x_i \mid f + f \mid f \cdot f \mid f^{\frac{1}{m}}$$

- Forward Equivalence
- Backward Equivalence

Reaction Networks [CONCUR'15, TACAS'16]



- Forward Bisimulation
- Backward Bisimulation

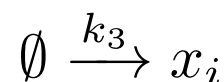
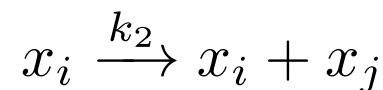
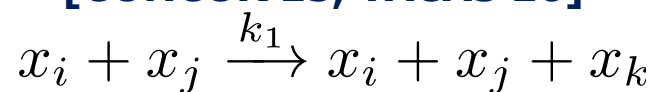
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Reaction Networks

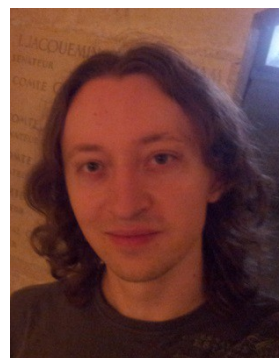
[CONCUR'15, TACAS'16]



Luca Cardelli¹



Mirco Tribastone²



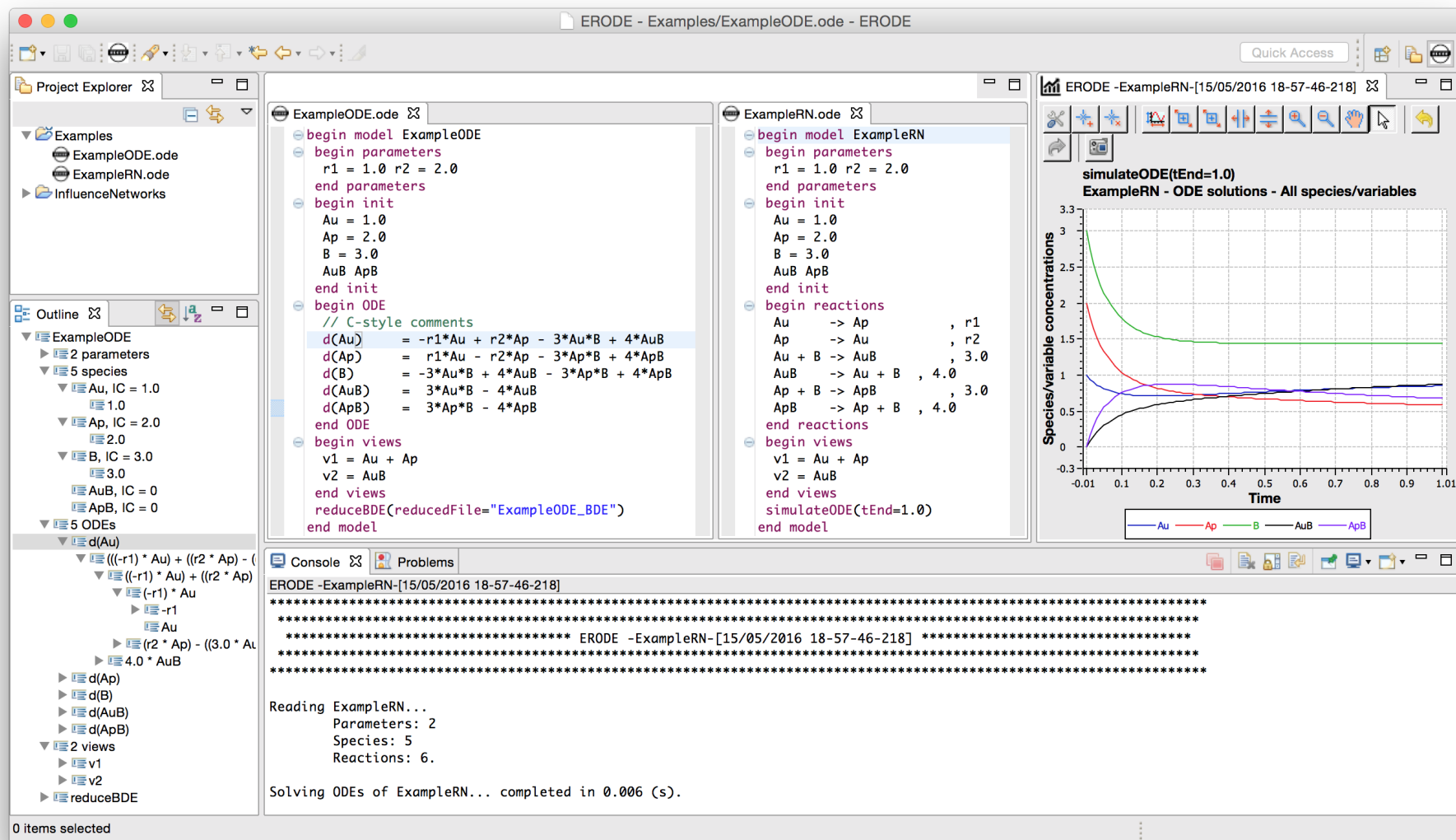
Max Tschaikowski²

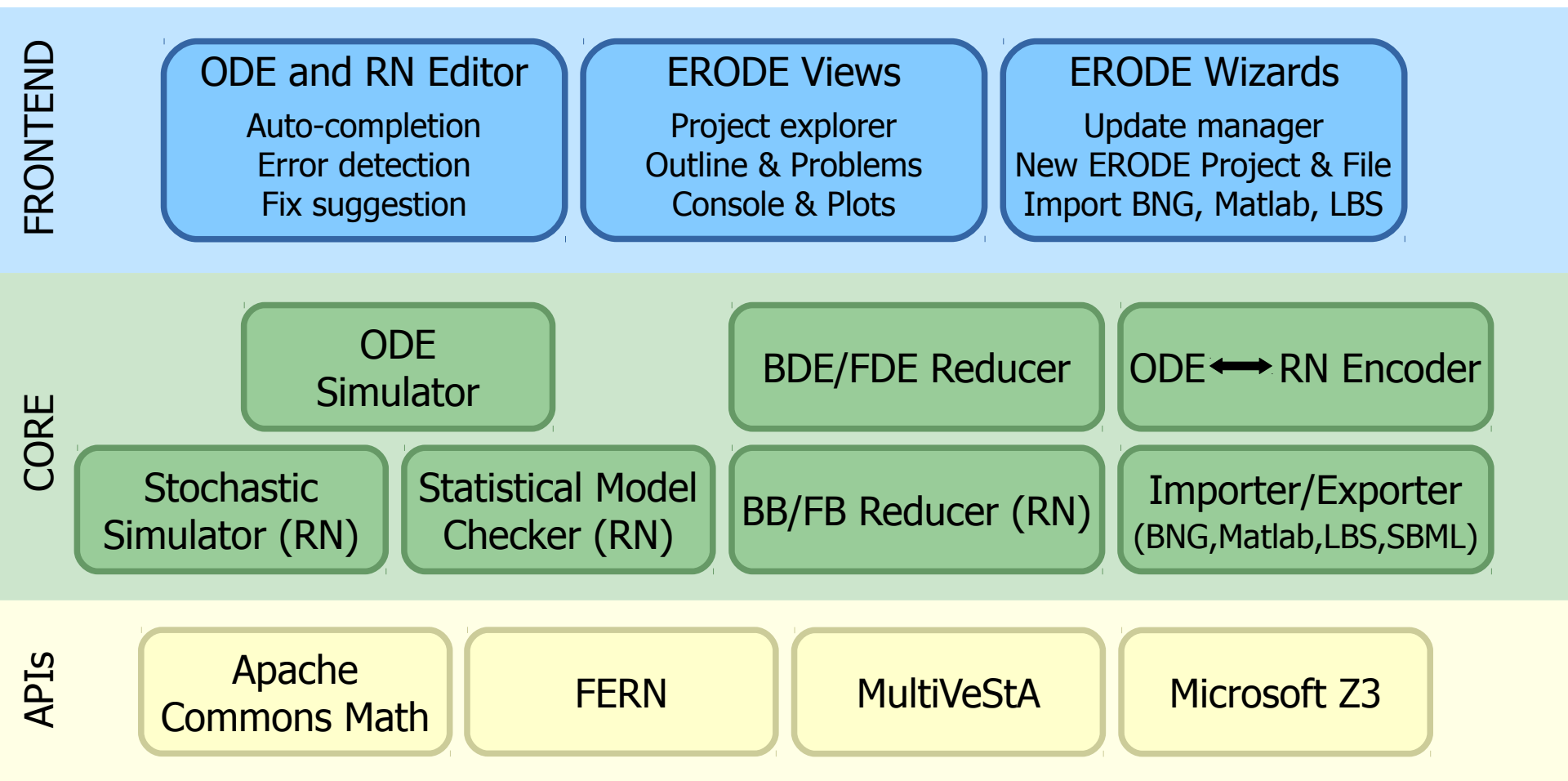


Andrea Vandin²

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² IMT School for Advanced Studies Lucca, Italy





FRONTEND

ODE and RN Editor

Auto-completion
Error detection
Fix suggestion

ERODE Views

Project explorer
Outline & Problems
Console & Plots

ERODE Wizards

Update manager
New ERODE Project & File
Import BNG, Matlab, LBS

CORE

ODE
Simulator

BDE/FDE Reducer

ODE \longleftrightarrow RN Encoder

Stochastic
Simulator (RN)

Statistical Model
Checker (RN)

BB/FB Reducer (RN)

Importer/Exporter
(BNG, Matlab, LBS, SBML)

APIs

Apache
Commons Math

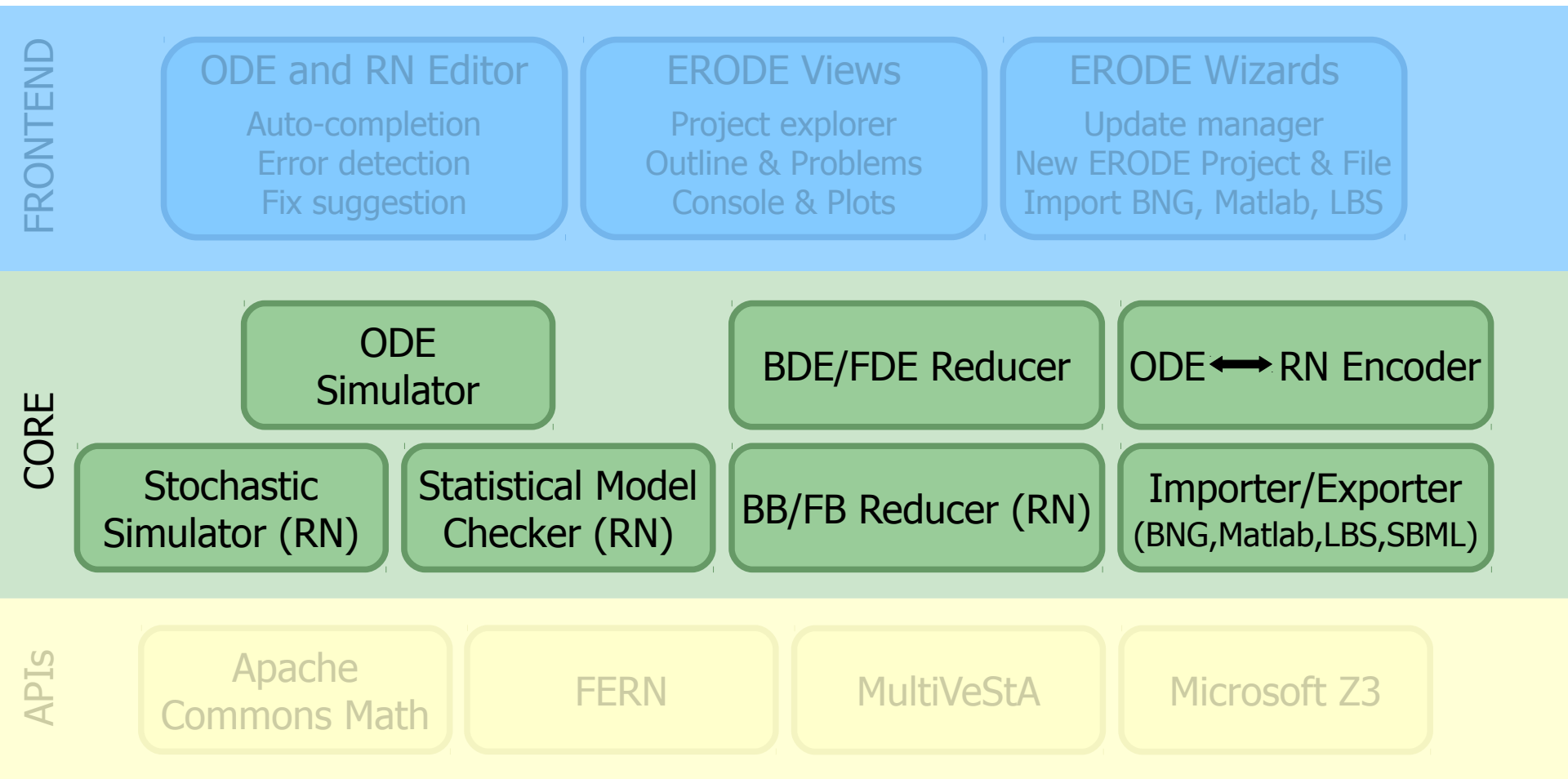
FERN

MultiVeStA

Microsoft Z3

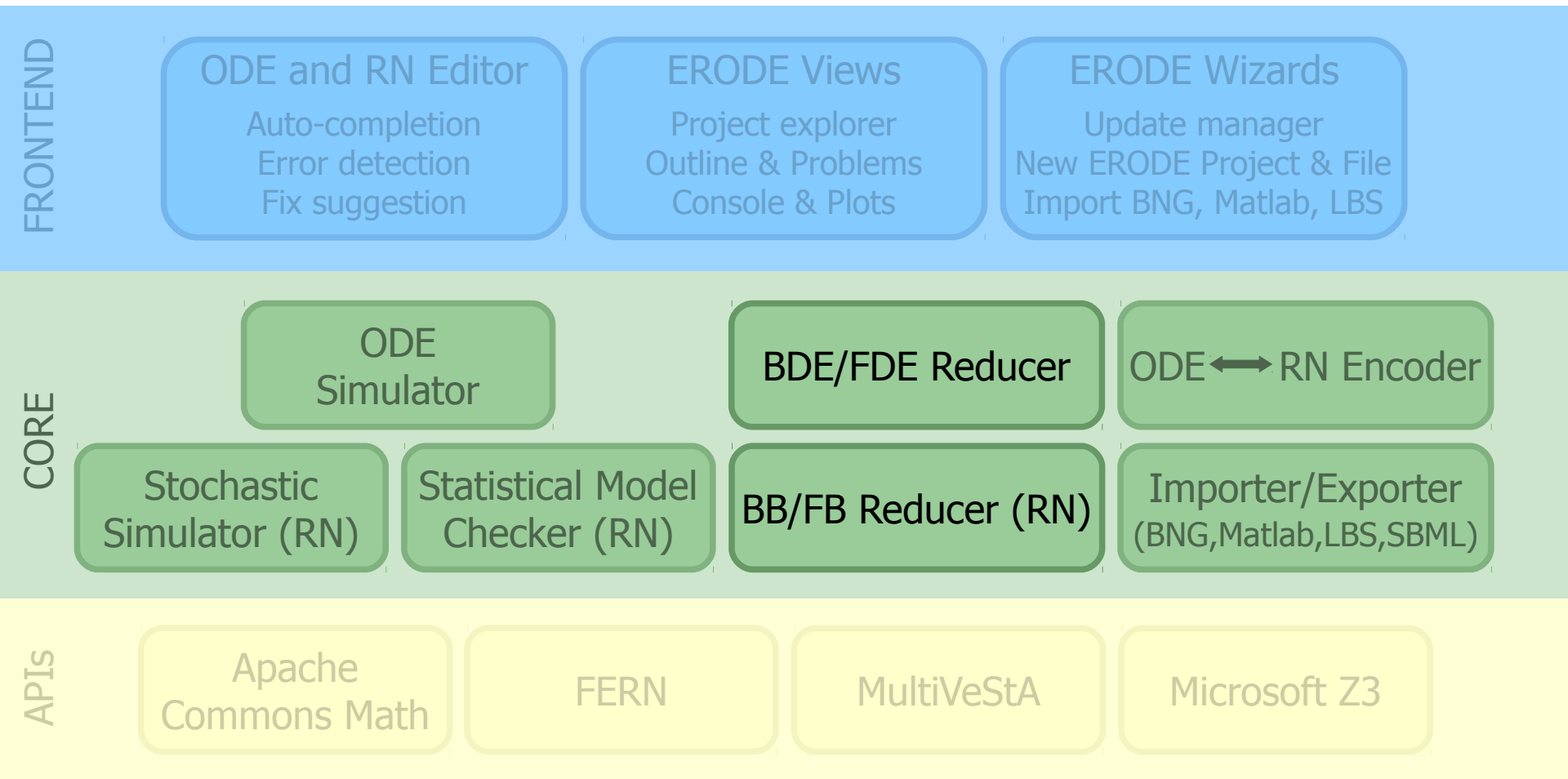
Highlights:

- The Project explorer
- Project and file creation
- Parts composing an ERODE file
- The Outline
- Auto completion, error detection and fix suggestion
- Importer wizards
- Updates manager

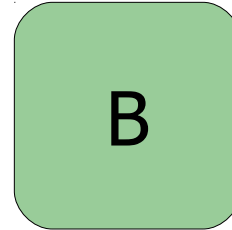
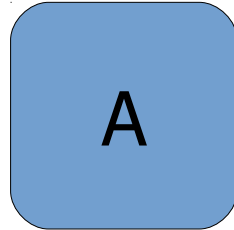


Highlights

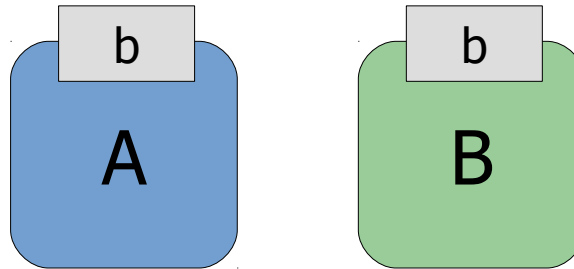
- SimulateODE
- setIC/setParam,
- Plots and CSV files
- Views in plots
- RN to ODE encoding
- Specific features of ODE editor
- Matlab export/import
- Stochastic simulation



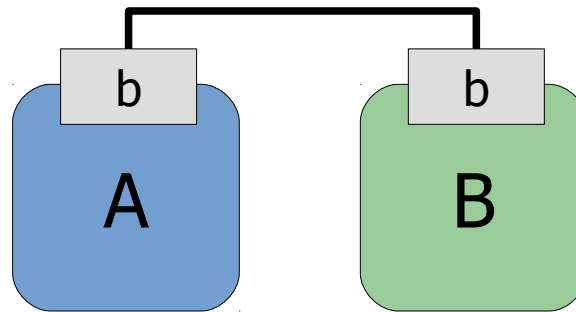
A simple binding/unbinding model



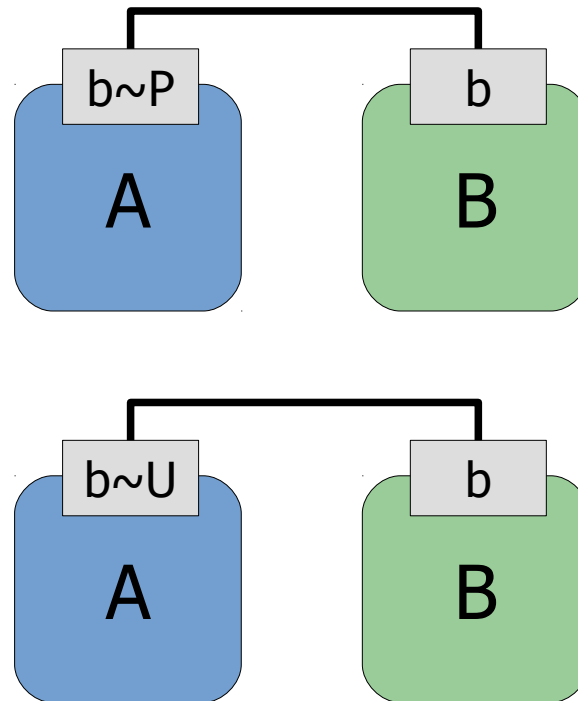
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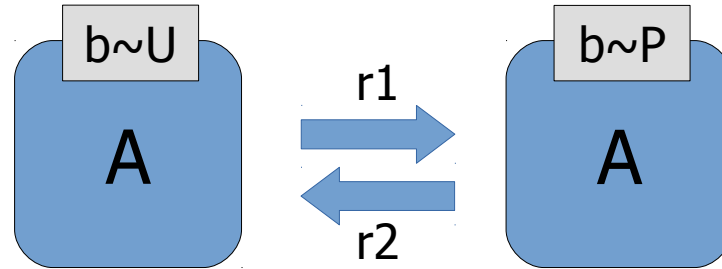
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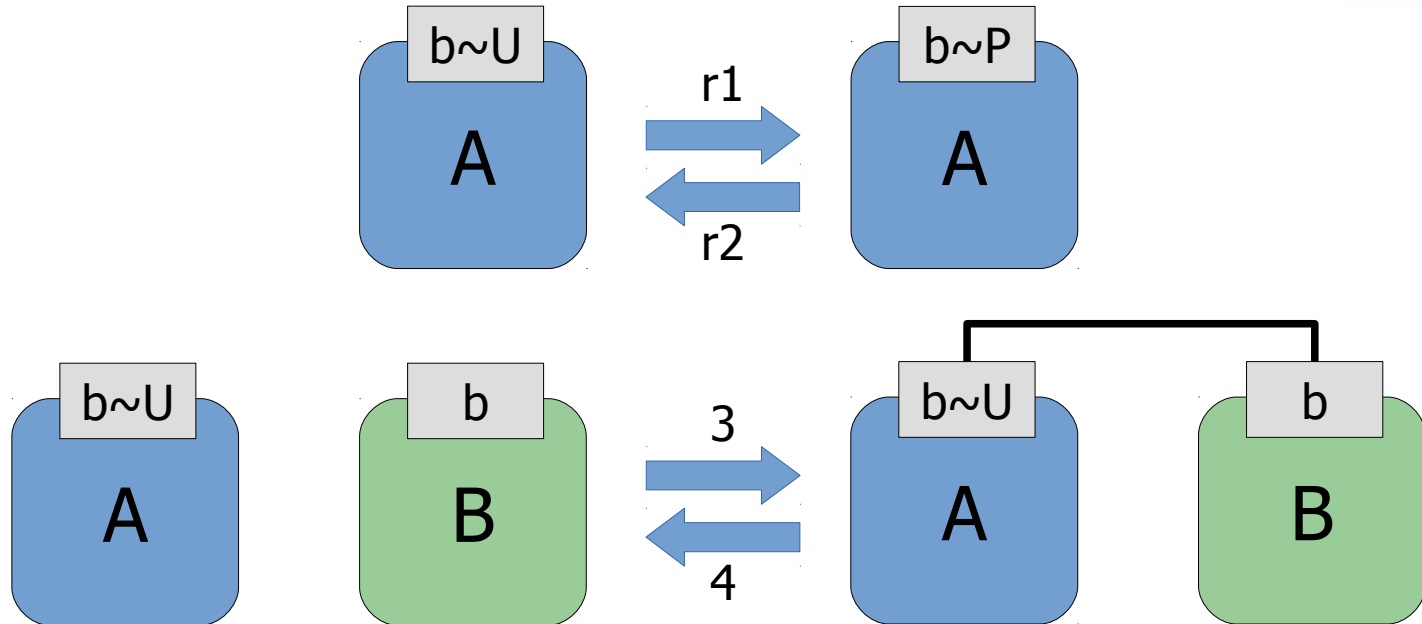
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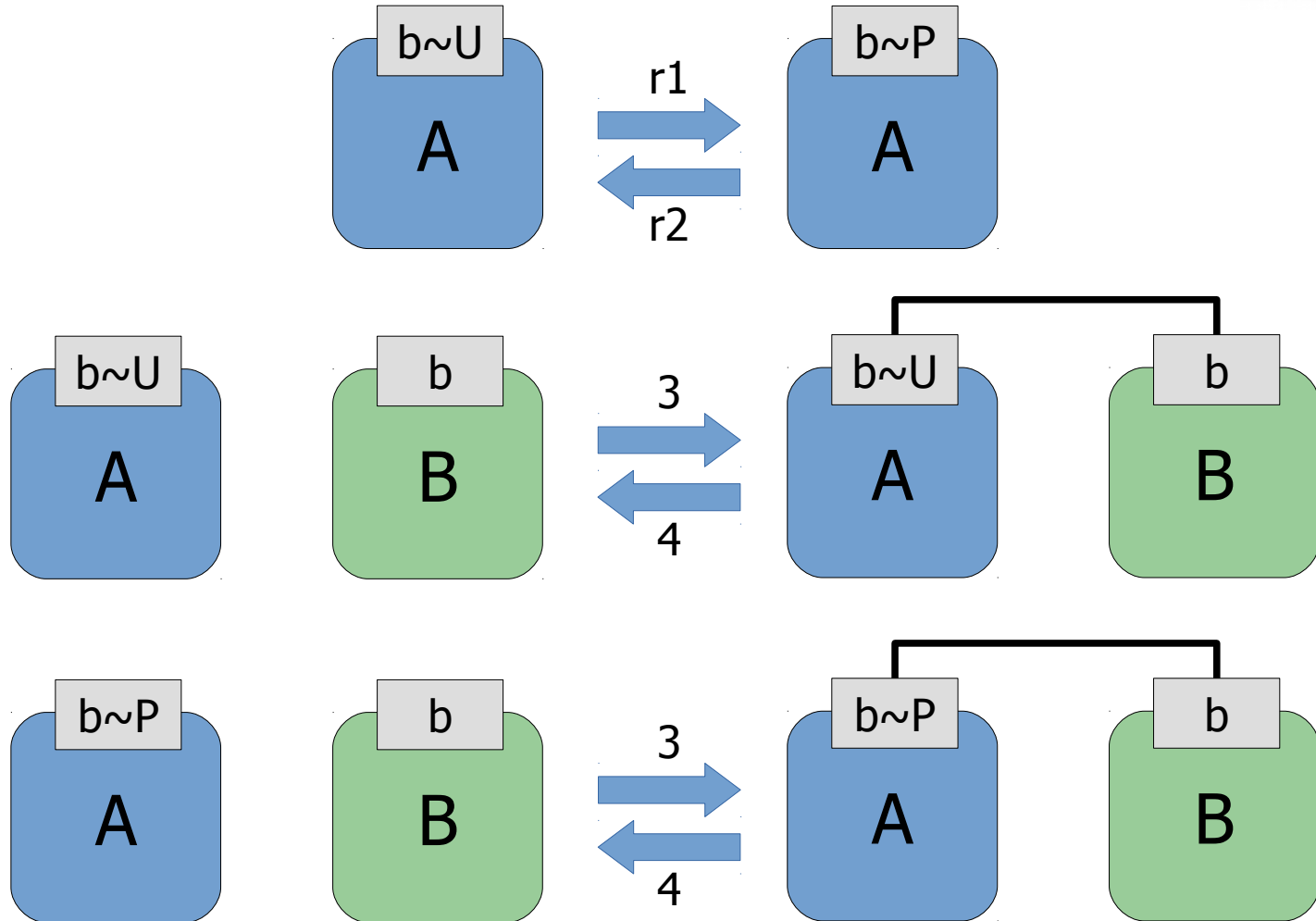
ERODE: Core – Reduction techniques



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ERODE: Core – Reduction techniques



Highlights

- RN specification
- ODE simulation
- ODE specification (via encoding)
- FDE reduction
 - With & without pre-partitioning
 - Explanation of reduction via plots
- BDE reduction
 - For $r_1=r_2$ and not
 - With & without IC pre-partitioning
 - Explanation of reduction via plots

Model

$$\dot{x}_1 = -x_1$$

$$\dot{x}_2 = k_1 \cdot x_1 - x_2$$

$$\dot{x}_3 = k_2 \cdot x_1 - x_3$$

Candidate partition

$$\left\{ \{x_1\}, \{x_2, x_3\} \right\}$$

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Candidate partition

$$\left\{ \{x_1\}, \{x_2, x_3\} \right\}$$

Equivalence condition (quantifier free)

$$\phi \equiv (x_2 = x_3) \implies k_1 \cdot x_1 - x_2 = k_2 \cdot x_1 - x_3$$

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SMT check

$$\text{sat}(\neg\phi) = \text{false}$$

Model

$$\dot{x}_1 = -x_1$$

$$\dot{x}_2 = k_1 \cdot x_1 - x_2$$

$$\dot{x}_3 = k_2 \cdot x_1 - x_3$$

Candidate partition

$$\{\{x_1\}, \{x_2, x_3\}\}$$

Equivalence condition (quantifier free)

$$\neg\phi \equiv$$

$$k_1 \cdot x_1 - x_2 \neq k_2 \cdot x_1 - x_2$$

SMT check

$$\text{sat}(\neg\phi) = \text{false}$$

Algorithm. Compute the largest equivalence that refines a given partition of variables:

$$k_1 = k_2 = 3$$

$$\dot{x}_1 = -x_1$$

$$\dot{x}_2 = 3 \cdot x_1 - x_2$$

$$\dot{x}_3 = 3 \cdot x_1 - x_3$$

$$\left\{ \{x_1, x_2, x_3\} \right\}$$

$$k_1 = k_2 = 3$$

Algorithm. Compute the largest equivalence that refines a given partition of variables:

1. SMT check of negation of

$$x_1 = x_2 = x_3 \implies$$

$$-x_1 = 3 \cdot x_1 - x_2 = 3 \cdot x_1 - x_3$$

$$\dot{x}_1 = -x_1$$

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$$\{\{x_1, x_2, x_3\}\}$$



$$x_1 = x_2 = x_3 = 1$$

$$\dot{x}_1 = -1$$

$$\dot{x}_2 = 3 \cdot 1 - 1 \quad \text{sat}$$

$$\dot{x}_3 = 3 \cdot 1 - 1$$

$$k_1 = k_2 = 3$$

Algorithm. Compute the largest equivalence that refines a given partition of variables:

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$$-x_1 = 3 \cdot x_1 - x_2 = 3 \cdot x_1 - x_3$$
2. If **sat** get witness and split partition preserving its uniformity. Goto 1.

$$\dot{x}_1 = -x_1$$

$$\dot{x}_2 = 3 \cdot x_1 - x_2$$

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$$\{\{x_1, x_2, x_3\}\}$$



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$$\{\{x_1\}, \{x_2, x_3\}\} \quad \mathbf{unsat}$$

$$k_1 = k_2 = 3$$

Algorithm. Compute the largest equivalence that refines a given partition of variables:

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$$x_1 = x_2 = x_3 \implies -x_1 = 3 \cdot x_1 - x_2 = 3 \cdot x_1 - x_3$$
2. If **sat** get witness and split partition preserving its uniformity. Goto 1.
3. If **unsat** the current partition is the coarsest refinement. End.

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$$\{\{x_1, x_2, x_3\}\}$$



$$x_1 = x_2 = x_3 = 1$$

$$\dot{x}_1 = -1$$

$$\dot{x}_2 = 3 \cdot 1 - 1 \quad \mathbf{sat}$$

$$\dot{x}_3 = 3 \cdot 1 - 1$$



$$\{\{x_1\}, \{x_2, x_3\}\} \quad \mathbf{unsat}$$

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Equivalence condition (quantifier free)

$$\neg\phi \equiv k_1 \cdot x_1 - x_2 \neq k_2 \cdot x_1 - x_2$$

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<http://rise4fun.com/Z3/IW7d1>

$$\text{sat}(\neg\phi) = \text{false}$$

Highlights

- RN Encoding
- FDE & FB reduction
 - Same FDE reductions as for ODE
 - FB is stricter than FDE
- BDE & BB reduction
 - Same BDE reductions as for ODE
 - The two reductions coincide
 - Comparison of the BDE/BB reduced RNs

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 - Frontend
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<i>Original model</i>		<i>FB reduction</i>			<i>BB reduction</i>		
$ R $	$ S $	<i>Red.(s)</i>	$ R $	$ S $	<i>Red.(s)</i>	$ R $	$ S $
Continuous-time Markov chains							
22 871 849	3 101 445	2.01E+3	1 069 777	135 752	1.34E+3	1 166 931	148 092
11 583 520	2 373 652	9.78E+1	5 792 531	1 187 597	3.07E+2	5 814 622	1 187 597
10 485 761	1 048 576	1.76E+1	3301	792	1.23E+1	5083	792
Biochemical reaction networks							
3 538 944	262 146	7.49E+0	990	222	1.21E+1	2 614	222
786 432	65 538	1.58E+0	720	167	2.51E+0	1 873	167
172 032	16 386	2.89E−1	504	122	6.03E−1	1 305	122
194 054	14 531	3.88E−1	142 165	10 855	6.00E−1	91 001	6 634
187 468	10 734	6.09E−1	57 508	3 744	1.40E+0	145 650	5 575
32 776	2 506	1.19E−1	16 481	1 281	2.14E−1	<i>32 776</i>	<i>2 506</i>
41 233	2 562	2.69E−1	33 075	1 897	3.97E−1	<i>41 233</i>	<i>2 562</i>

Largest models from MRMC distribution

Comparison with state-of-the-art model checker to assess scalability and correctness

<i>Original model</i>		<i>FB reduction</i>			<i>BB reduction</i>		
$ R $	$ S $	<i>Red.(s)</i>	$ R $	$ S $	<i>Red.(s)</i>	$ R $	$ S $
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Biochemical models from BioNetGen repository

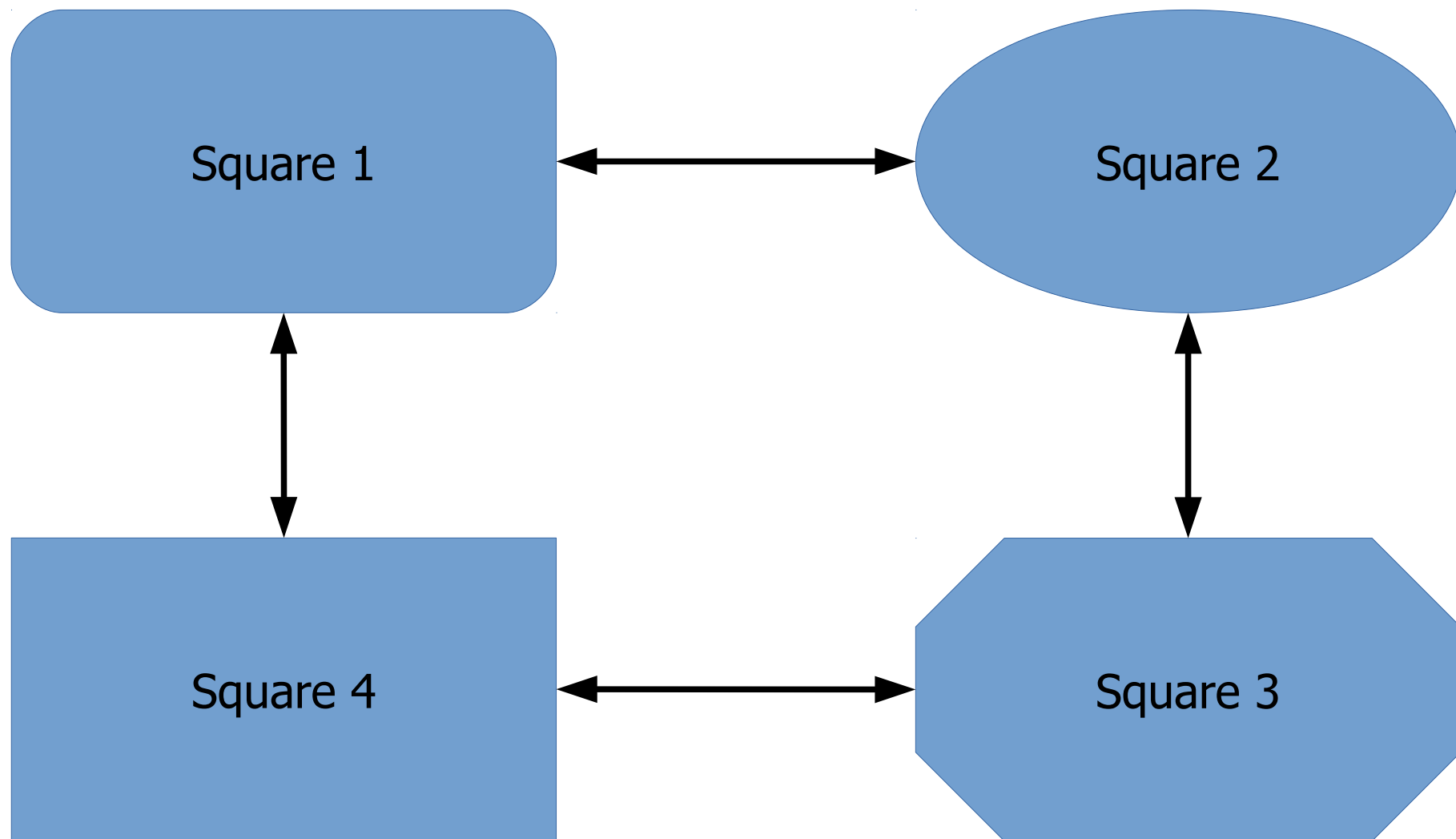
Support for state-of-the-art tool for the analysis of biological networks

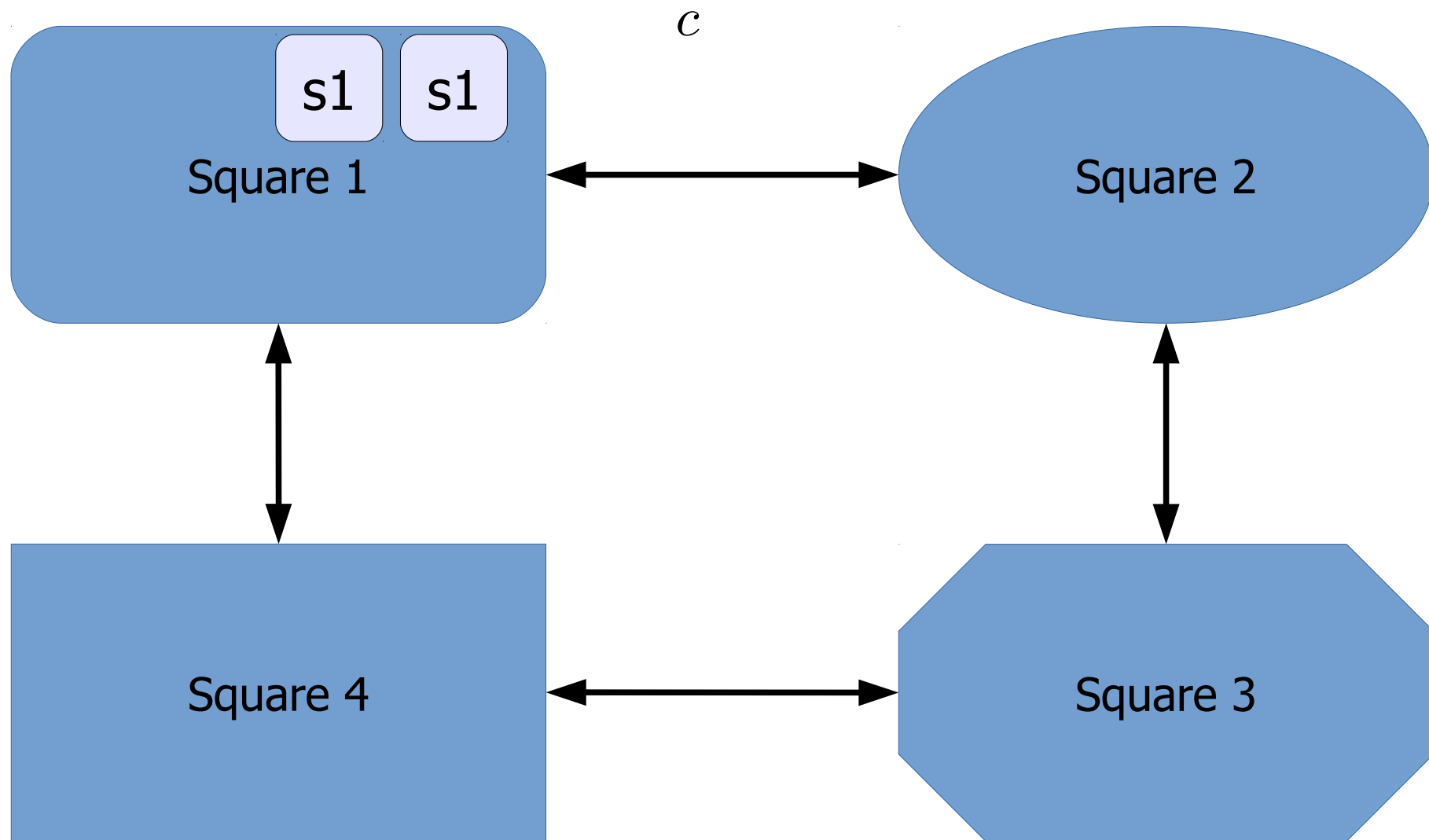
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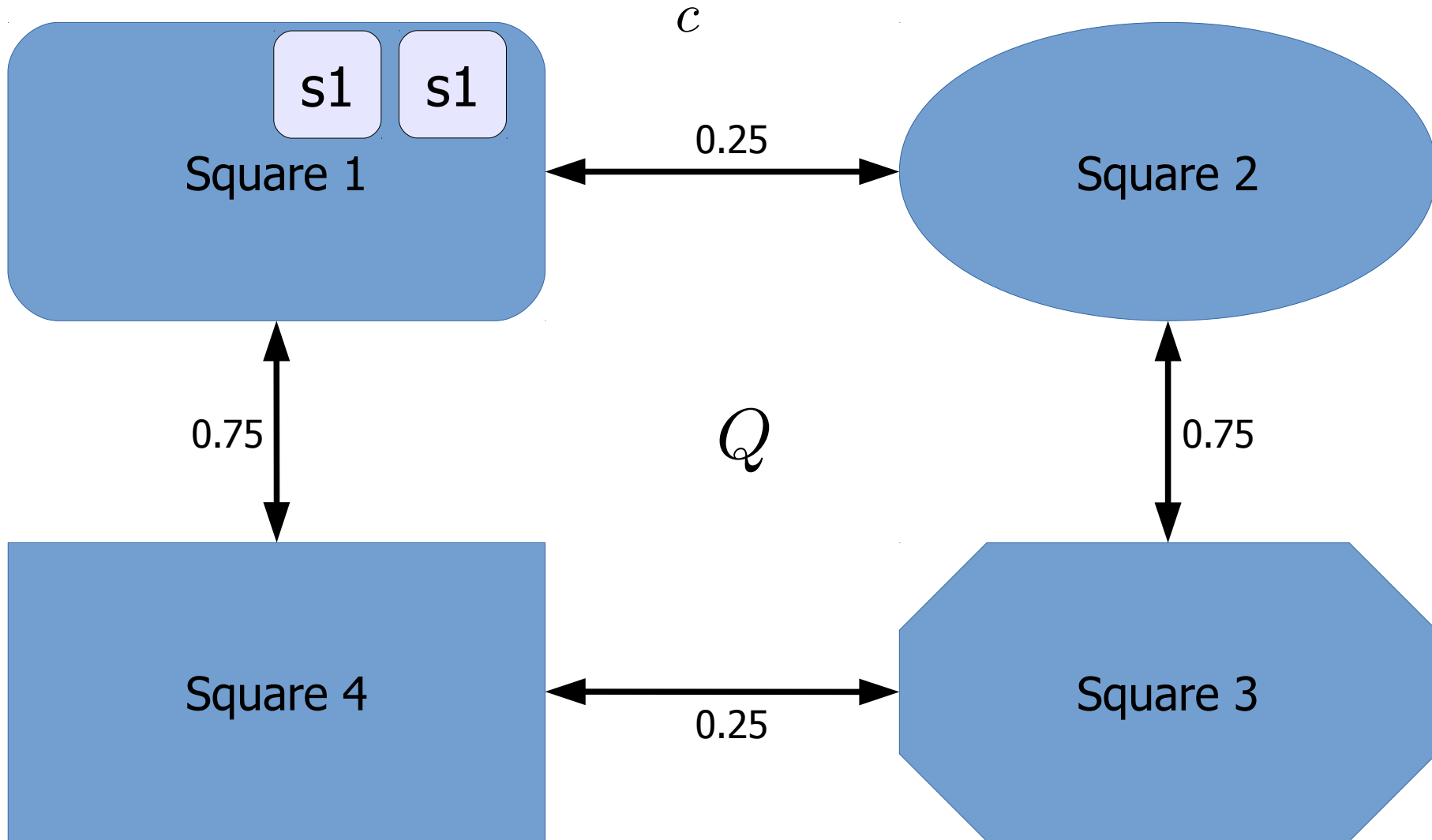
<i>Original model</i>		<i>FB reduction</i>			<i>FB reduction with prep.</i>			
$ R $	$ S $	$Red.(s)$	$ R $	$ S $	$ Prep. $	$Red.(s)$	$ R $	$ S $
194 054	14 531	3.88E−1	142 165	10 855	1 345	3.28E−1	147 797	12 037
5 797	796	1.90E−2	4 210	503	18	4.10E−2	4 210	503
487	85	2.00E−3	264	56	4	2.00E−3	362	69

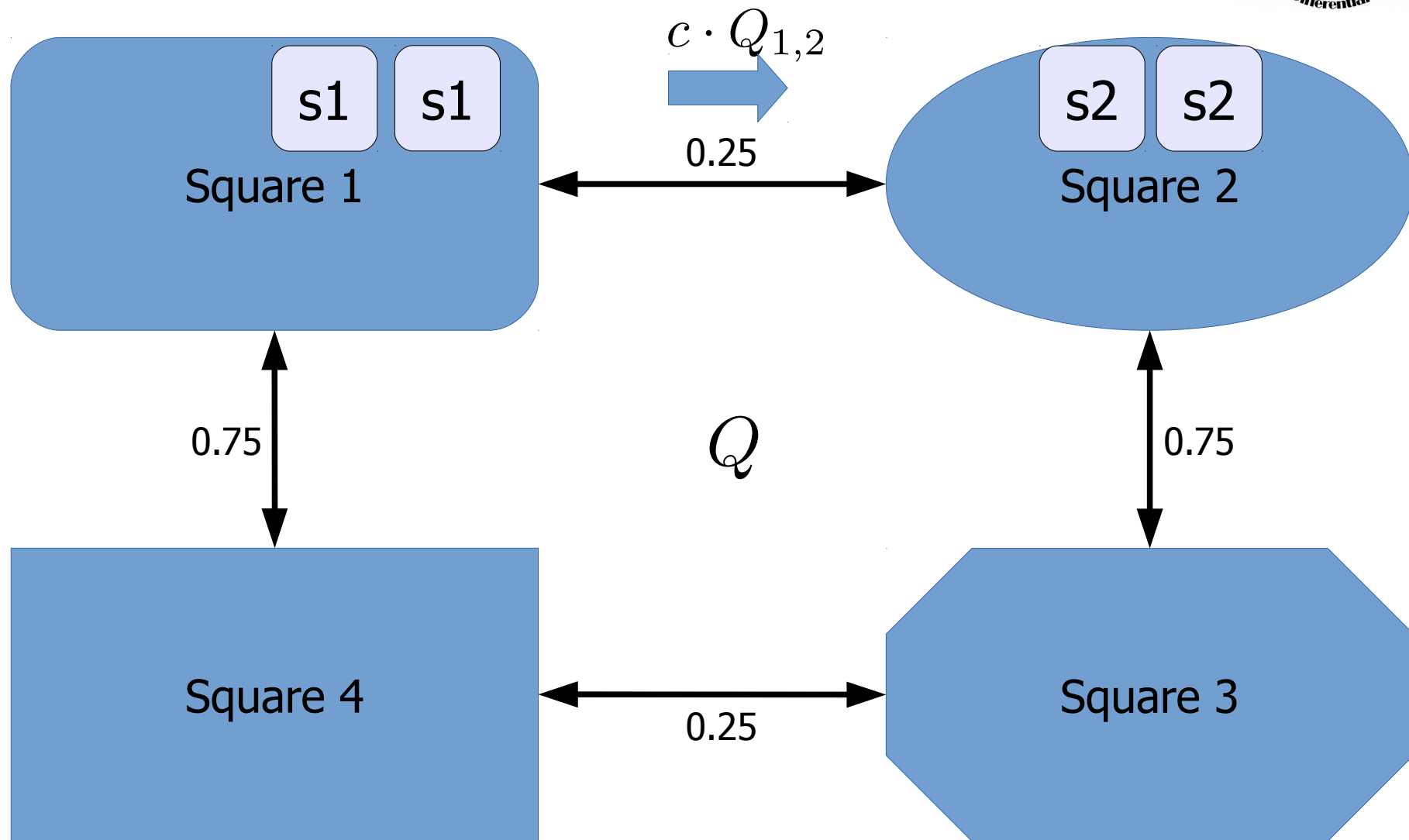
<i>Original model</i>		<i>FB reduction</i>			<i>FB reduction with prep.</i>			
$ R $	$ S $	$Red.(s)$	$ R $	$ S $	$ Prep. $	$Red.(s)$	$ R $	$ S $
194 054	14 531	3.88E−1	142 165	10 855	1 345	3.28E−1	147 797	12 037
5 797	796	1.90E−2	4 210	503	18	4.10E−2	4 210	503
487	85	2.00E−3	264	56	4	2.00E−3	362	69

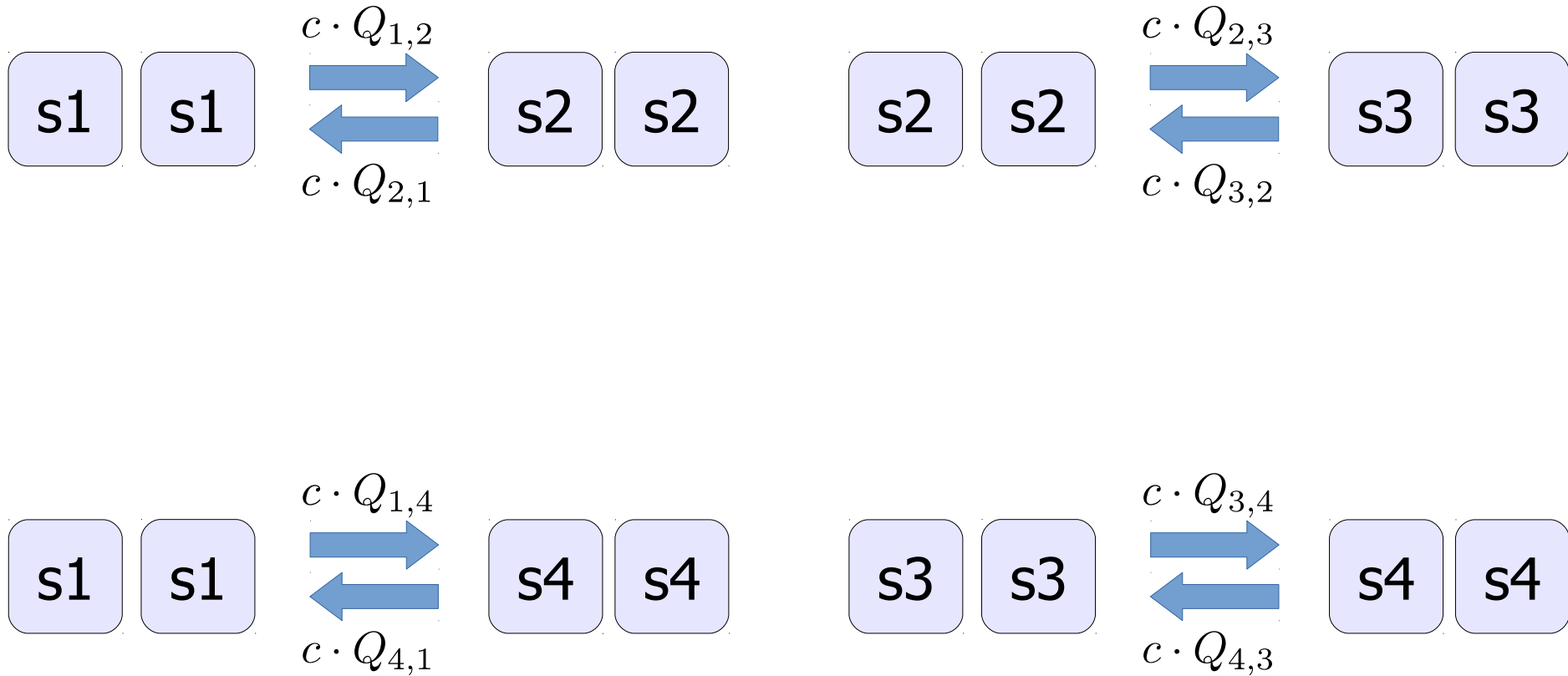
- ERODE: A tool for the Evaluation and Reduction of ODEs
 - Frontend
 - Core
 - Reduction Techniques
- Case Studies
 - Large case studies
 - Continuous time Markov chains from MRMC distribution
 - Chemical Reaction Networks from the literature
 - Spatial dynamics
 - Crowd Dynamics
 - Multi-community Epidemiology





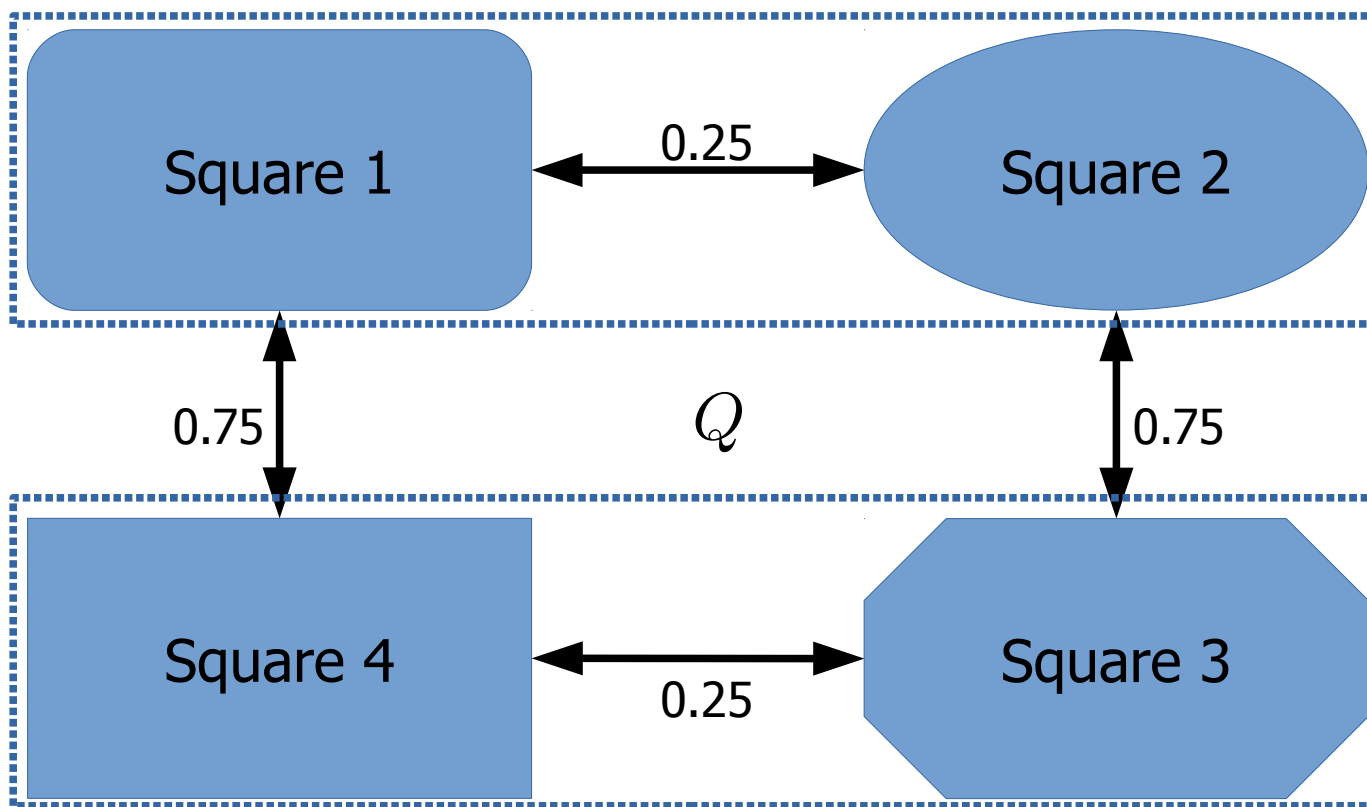


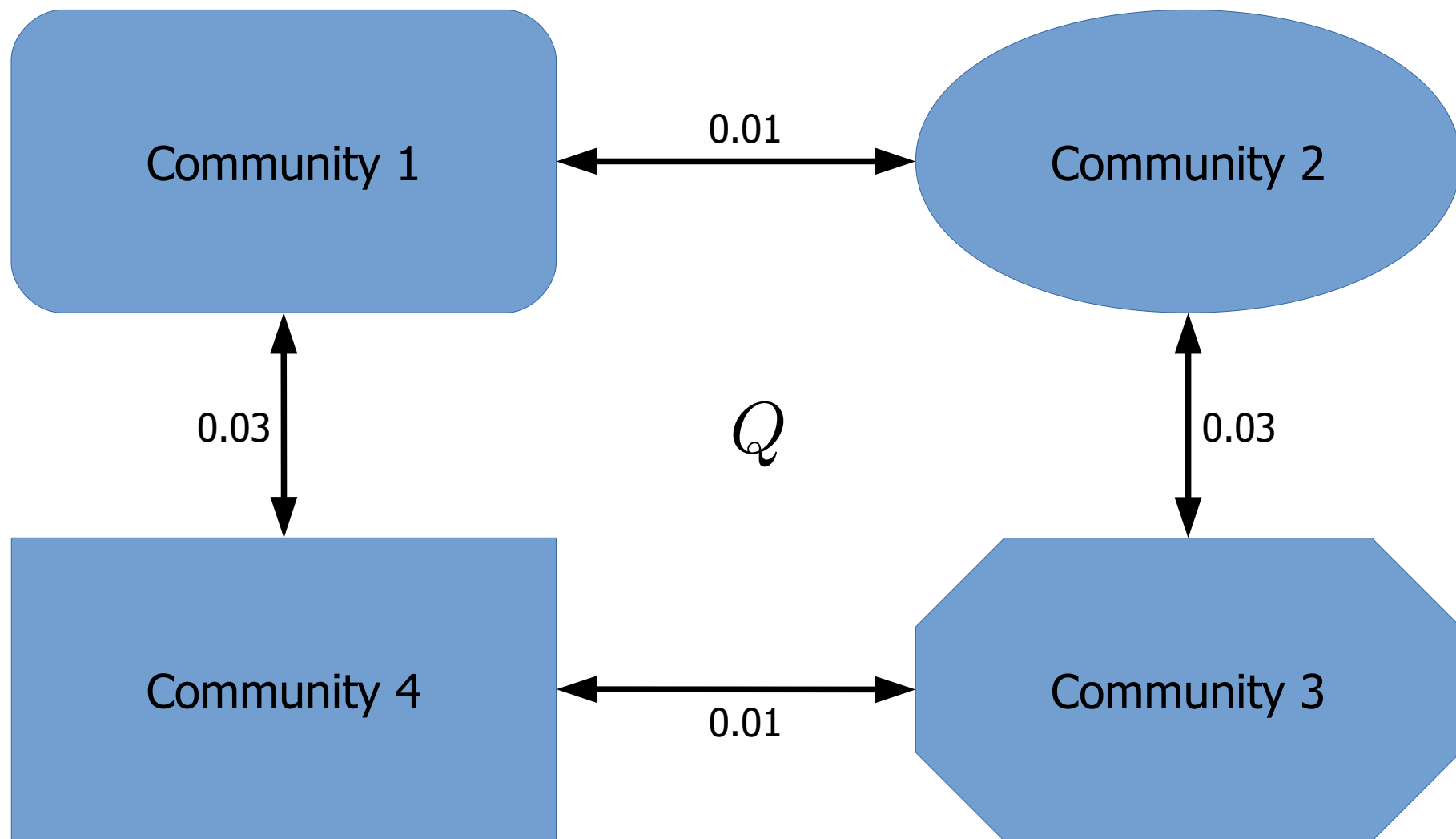


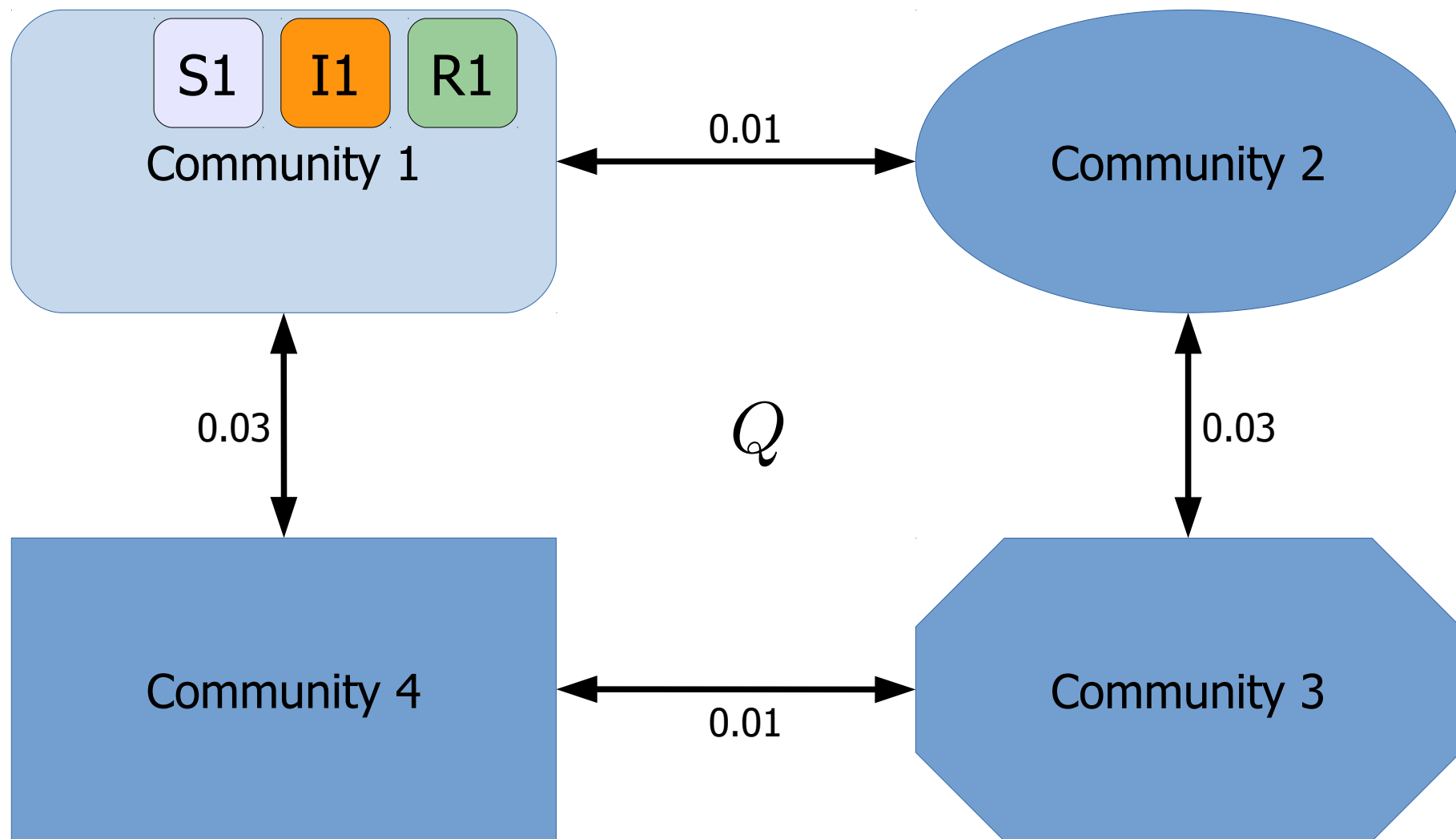


Highlights

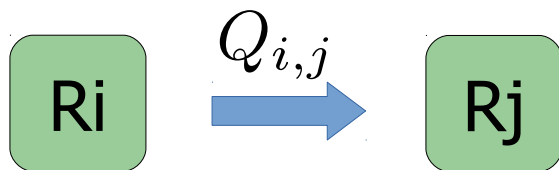
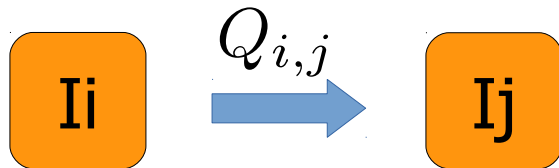
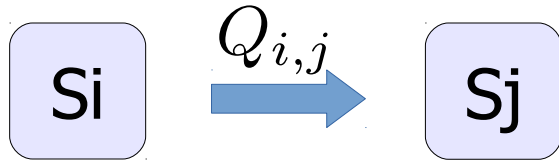
- Plots for different IC
- BDE/BB formally proves the presence of “symmetric squares”



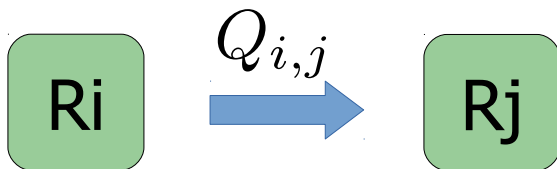
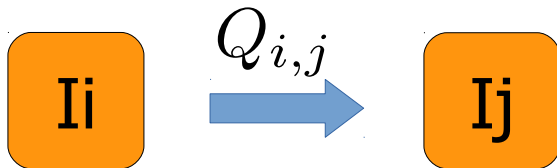
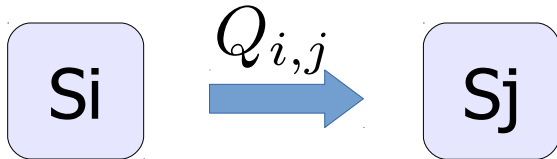




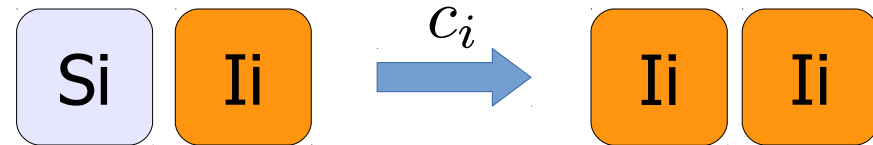
Movements



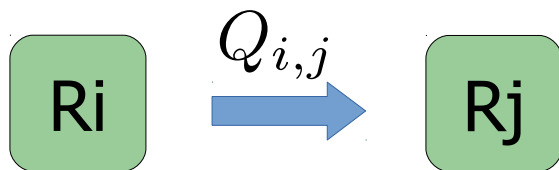
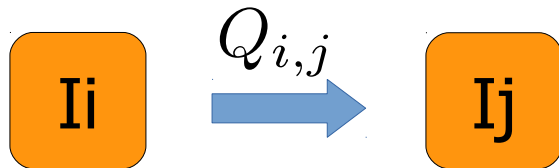
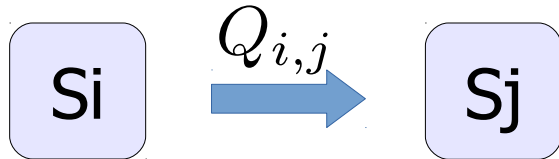
Movements



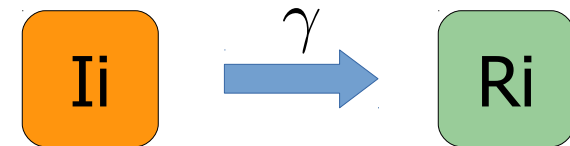
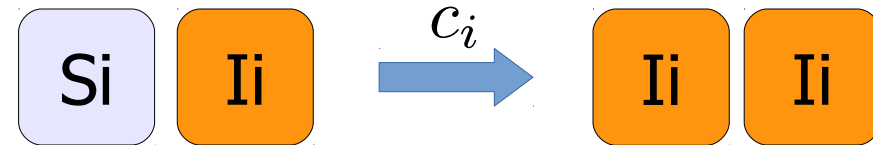
Local interactions



Movements



Local interactions



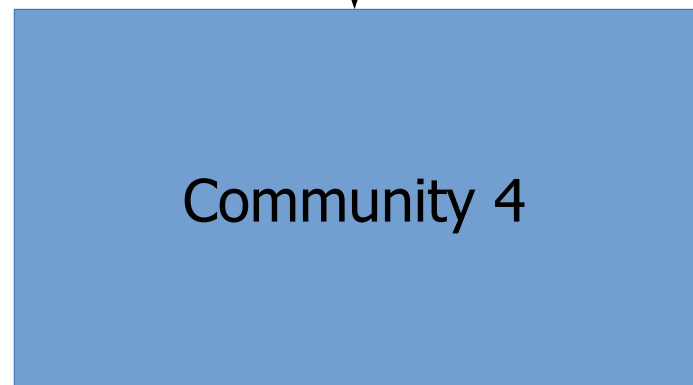
$$c_1 = 0.03$$



$$c_2 = 0.03$$

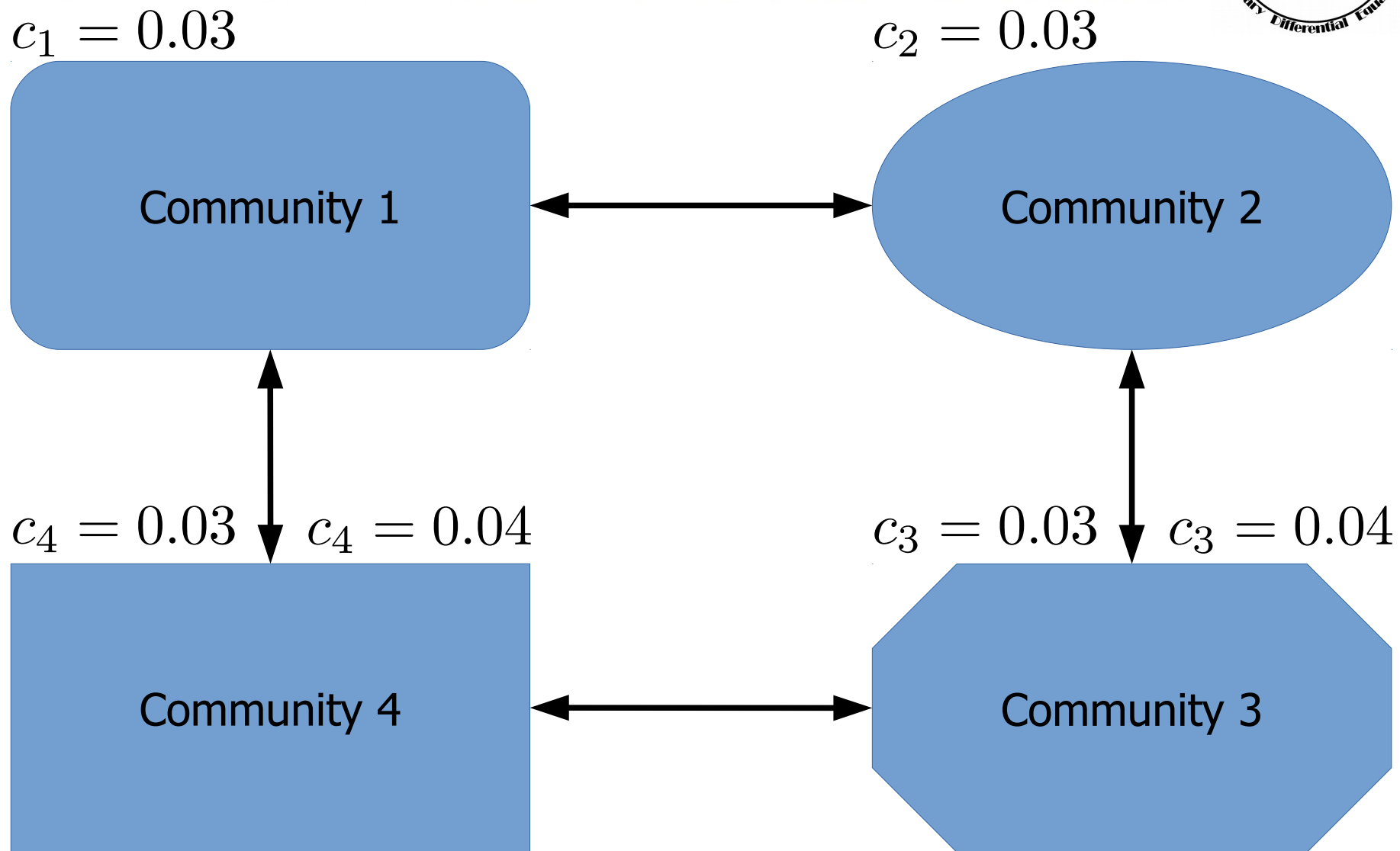


$$c_4 = 0.03$$



$$c_3 = 0.03$$





<http://groups.inf.ed.ac.uk/paloma/SIR.paloma>

```
#params
param p =0.5;

#states
//an agent in state "S" can receive a "contact" msg and become infected, or move to another location.
S(0,0):= ?(contact,p)@Pr{1/(|S(0,0)|+|I(0,0)|+|R(0,0)|)}.I(0,0)+(move_0_1,0.01).S(0,1)+(move_0_2,0.01).S(1,0);
S(0,1):= ?(contact,p)@Pr{1/(|S(0,1)|+|I(0,1)|+|R(0,1)|)}.I(0,1)+(move_1_0,0.01).S(0,0)+(move_1_3,0.01).S(1,1);
S(1,0):= ?(contact,p)@Pr{1/(|S(1,0)|+|I(1,0)|+|R(1,0)|)}.I(1,0)+(move_2_0,0.01).S(0,0)+(move_2_3,0.01).S(1,1);
S(1,1):= ?(contact,p)@Pr{1/(|S(1,1)|+|I(1,1)|+|R(1,1)|)}.I(1,1)+(move_3_1,0.01).S(0,1)+(move_3_2,0.01).S(1,0);

//an agent in state "I" can send out "contact" broadcast msg to its current location, or move to another location.
I(0,0):= !(contact,0.03)@IR{local}.I(0,0)+(recover,0.2).R(0,0)+(move_0_1,0.01).I(0,1)+(move_0_2,0.01).I(1,0);
I(0,1):= !(contact,0.03)@IR{local}.I(0,1)+(recover,0.2).R(0,1)+(move_1_0,0.01).I(0,0)+(move_1_3,0.01).I(1,1);
I(1,0):= !(contact,0.03)@IR{local}.I(1,0)+(recover,0.2).R(1,0)+(move_2_0,0.01).I(0,0)+(move_2_3,0.01).I(1,1);
I(1,1):= !(contact,0.03)@IR{local}.I(1,1)+(recover,0.2).R(1,1)+(move_3_1,0.01).I(0,1)+(move_3_2,0.01).I(1,0);

//an agent in state "R" can only move to another location.
R(0,0):= (move_0_1,0.01).R(0,1)+(move_0_2,0.01).R(1,0);
R(0,1):= (move_1_0,0.01).R(0,0)+(move_1_3,0.01).R(1,1);
R(1,0):= (move_2_0,0.01).R(0,0)+(move_2_3,0.01).R(1,1);
R(1,1):= (move_3_1,0.01).R(0,1)+(move_3_2,0.01).R(1,0);

#agents
S(0,0)[150] || I(0,0)[11] || R(0,0)[12] ||
S(0,1)[150] || I(0,1)[11] || R(0,1)[12] ||
S(1,0)[150] || I(1,0)[11] || R(1,0)[12] ||
S(1,1)[150] || I(1,1)[11] || R(1,1)[12]
```

```

fluidflow.m  x  +
1  function dy = fluidflow(y)
2  -    dy=zeros(90,1);
3
4  %First-order moments (12 ODEs)
5  -    dy(1)=0-1*0.03*y(1)-1*0.01*y(1)+1*0.01*y(2)+1*0.03*y(9)+1*0.2*y(10);
6  -    ...
7  -    dy(12)=0+1*0.03*0.5*1/(y(5)+y(12)+y(2))*y(69)+1*0.01*y(10)+1*0.03*y(1);
8
9  %Second-order moments (12 ODEs)
10 -    dy(13)=0+0.03*(1)^2*y(1)-2*y(13)*0.03*1+0.01*(1)^2*y(1)-2*y(13)*0.01*y(2);
11 -    ...
12 -    dy(24)=0+0.03*(0.5*1/(y(5)+y(12)+y(2)))^2*(y(17)*((y(69))^2)/(y(12)+eps));
13
14 %Further 66 auxiliary ODEs
15 -    dy(25)=0-0.03*1*y(25)-0.01*1*y(25)+0.01*1*y(13)-0.01*(1)^2*y(1)-0.03*y(1)*y(2);
16 -    ...
17 -    dy(90)=0+0.03*0.5*1/(y(5)+y(12)+y(2))*(y(69)*y(90)*y(68)/(y(12)+eps));
18
19 -    end

```

