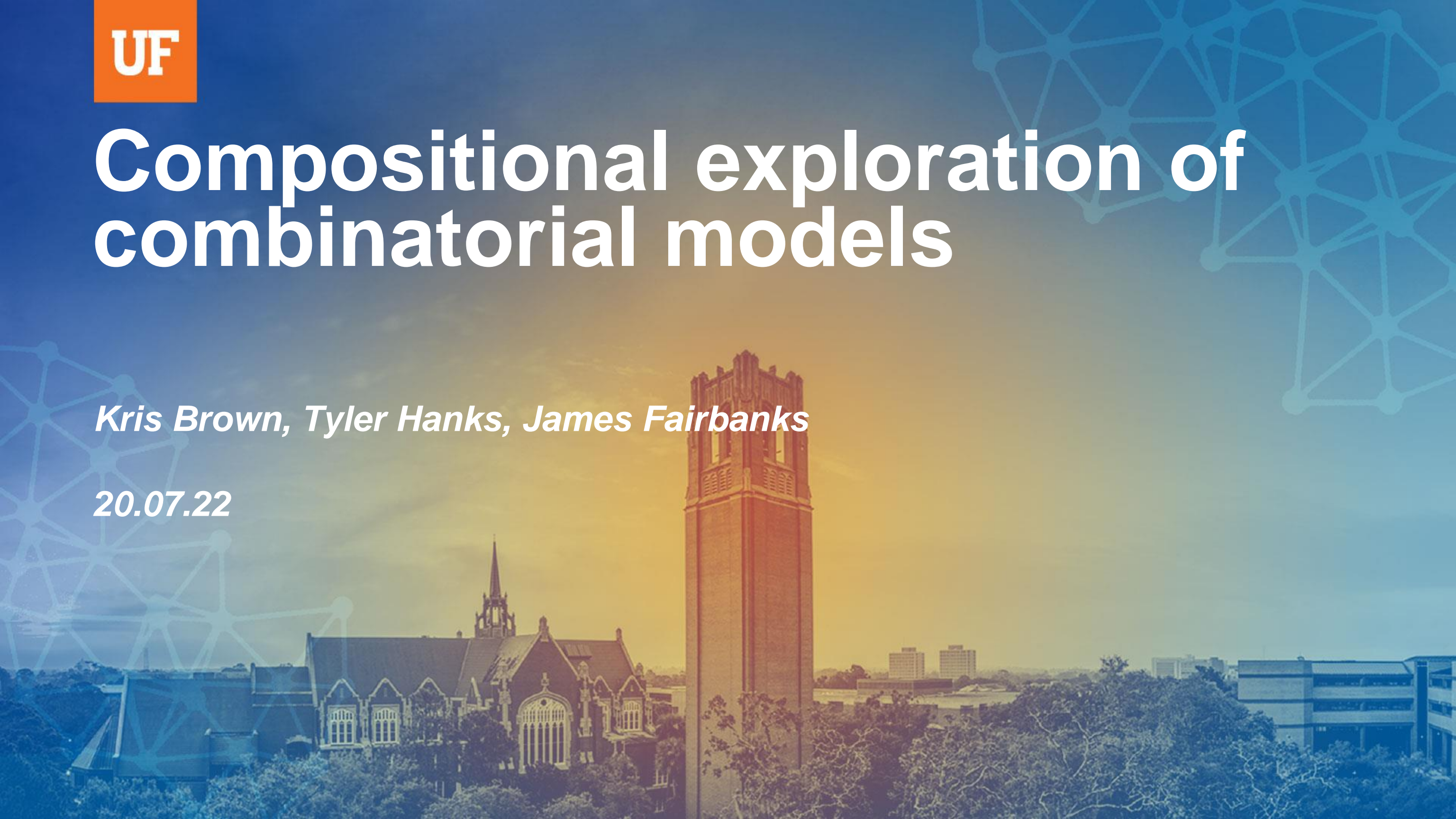


Compositional exploration of combinatorial models

Kris Brown, Tyler Hanks, James Fairbanks

20.07.22



Introduction

- Why represent scientific models combinatorially?

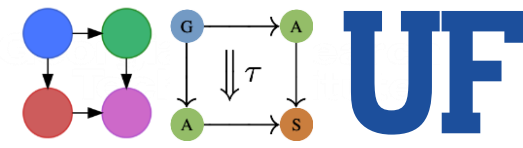
Model space exploration

- The category of diagrams as a category of model spaces
- Example limits and colimits
- Composition recipes
- Limits and colimits: implementation

Model Selection

- Best fit chemical reaction network example

Alternative: model as opaque code / math / logic



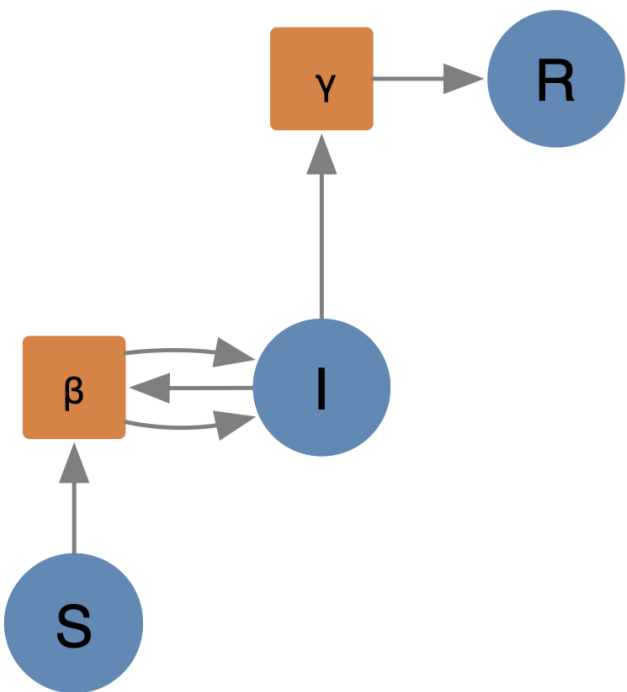
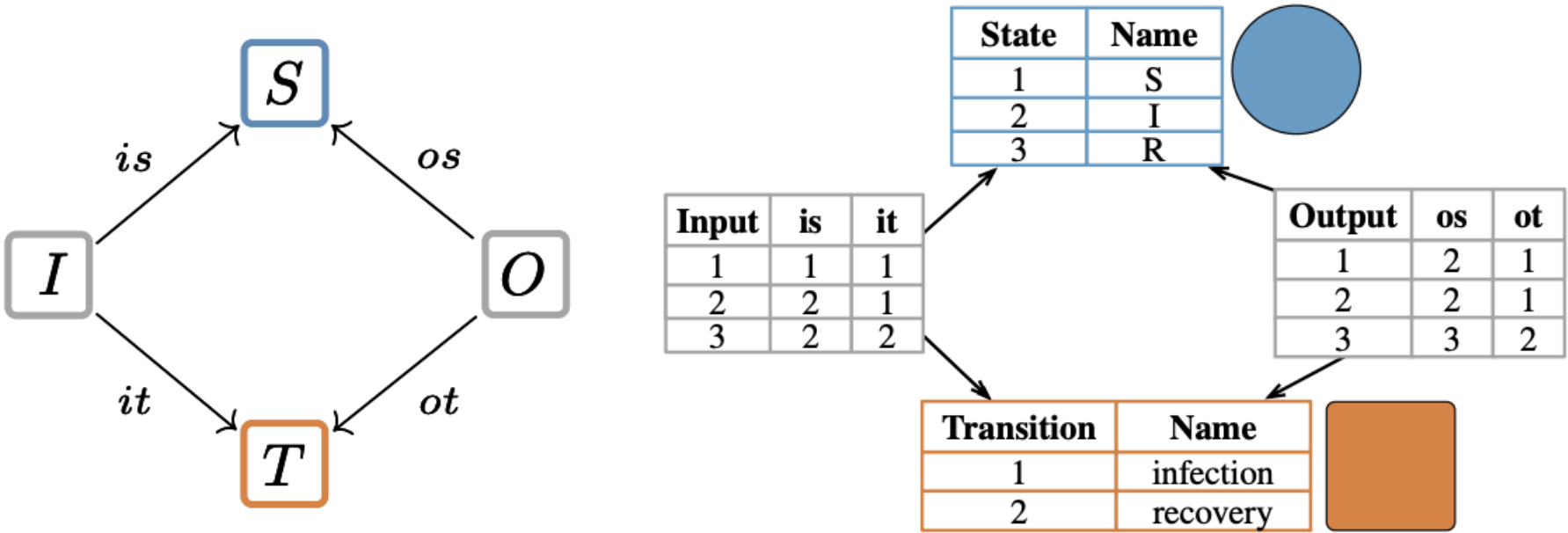
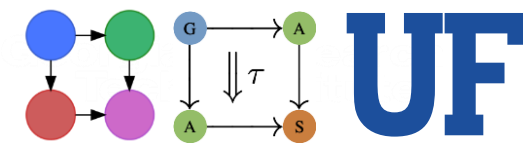
```
"""
2 H2 + O2 → 2 H2O. Mass-action kinetics. Compare to experimental data and plot.
"""
def main():
    # experimental data
    real_data = [0.0101, 0.012, 0.023, 0.037, 0.045, 0.053, 0.061, 0.069,
0.076, 0.083, 0.089, 0.096, 0.102, 0.108, 0.114, 0.119, 0.125, 0.130, 0.135,
0.140, 0.145, 0.150, 0.154, 0.159, 0.163, 0.167, 0.171, 0.175, 0.179, 0.183,
0.186, 0.190, 0.193, 0.197, 0.200, 0.203, 0.206, 0.209, 0.212, 0.215, 0.218,
0.221, 0.224, 0.227, 0.229, 0.232, 0.234, 0.237, 0.239]
    # Initial concentrations
    H2,O2, H2O = 1.0, 2.0, 0.0
    dt = 0.01
    results = []
    for step in range(1,50):
        print("Step ", step)
        rate = 0.5 * H2**2 * O2
        H2 -= 2*rate*dt
        O2 -= rate*dt
        H2O += rate*dt
        results.append(H2O)

    plot(results, real_data) # Figure 4 in the paper
```

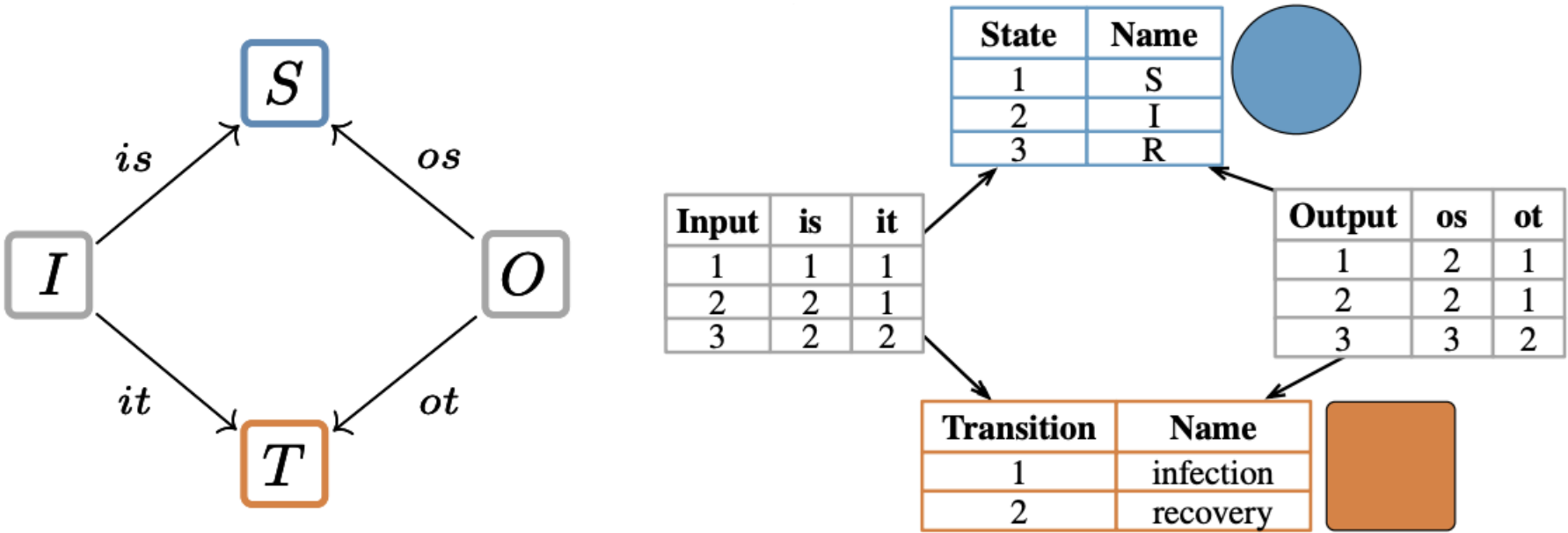
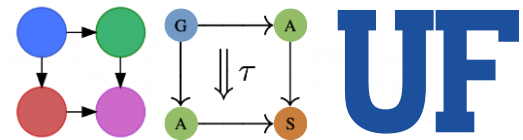
Many tasks we'd like to do cannot be done with arbitrary code (nor mathematical expressions).

1. Generate the entire code from just declaring the reaction
2. Easily alter semantics (e.g. stochastic-based simulation)
3. Construct models compositionally (operad)
4. Construct models compositionally (limits).
5. Explore alternate reaction networks to fit the data

1. Functorial generation of code



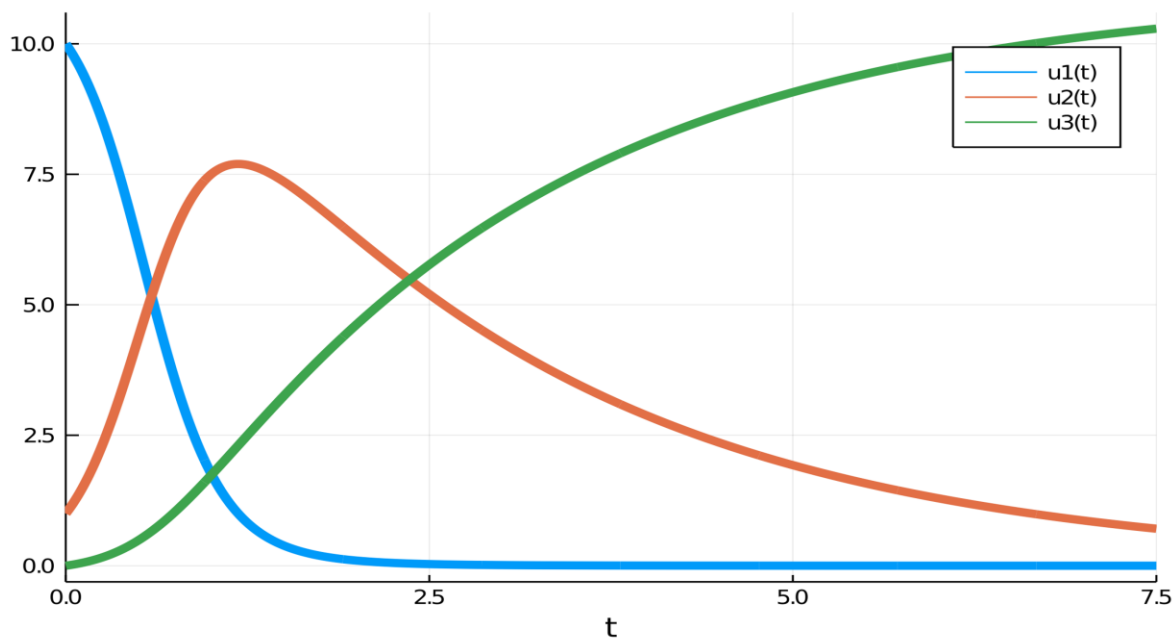
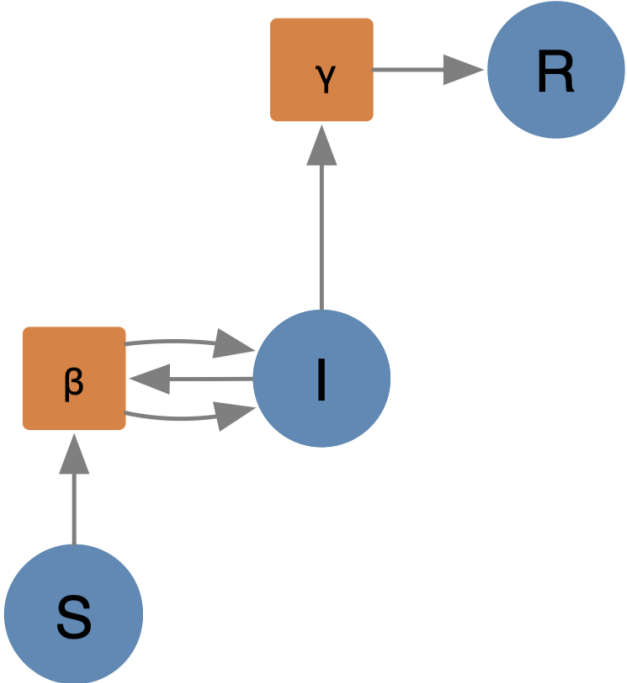
1. Functorial generation of code



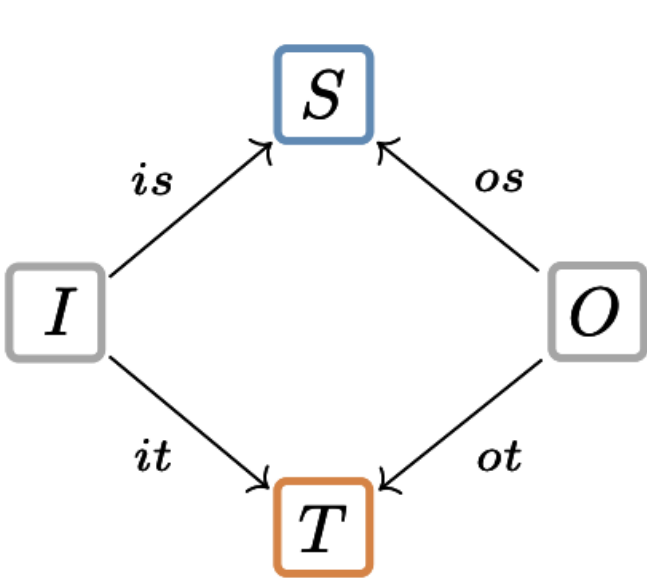
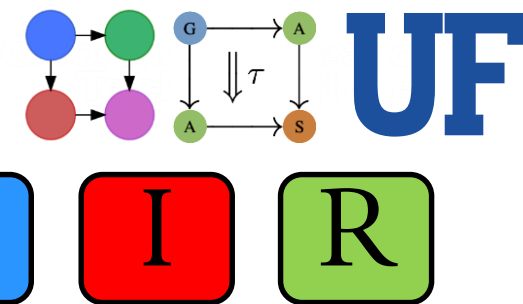
Specifying a reaction network as a Petri Net allows for automatically generating an ODE simulation.

$$r_{\beta} = k_{\beta} \cdot [S] \cdot [I]$$

$$r_{\gamma} = k_{\gamma} \cdot [I]$$



2. Easily change semantics with different functor



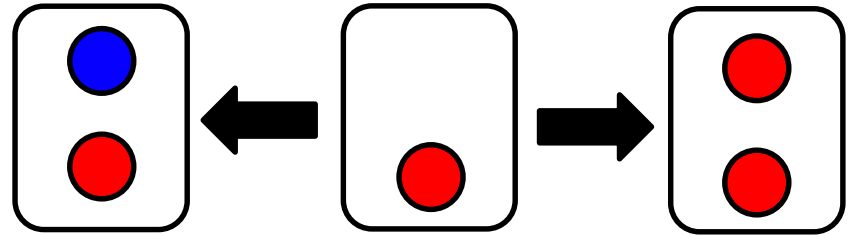
State	Name
1	S
2	I
3	R

Input	is	it
1	1	1
2	2	1
3	2	2

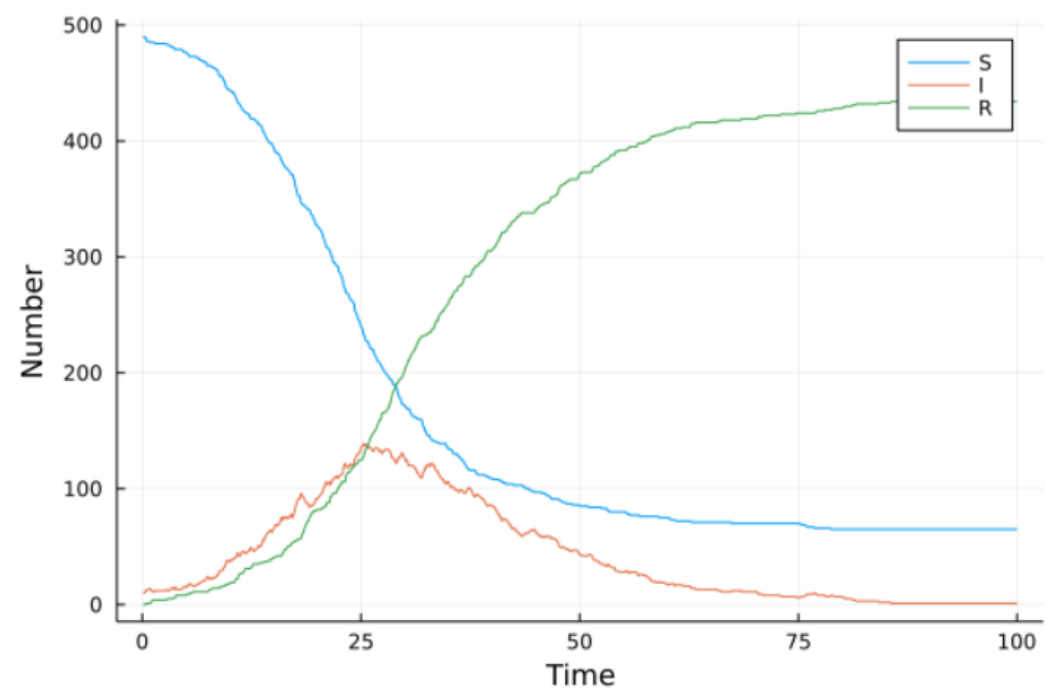
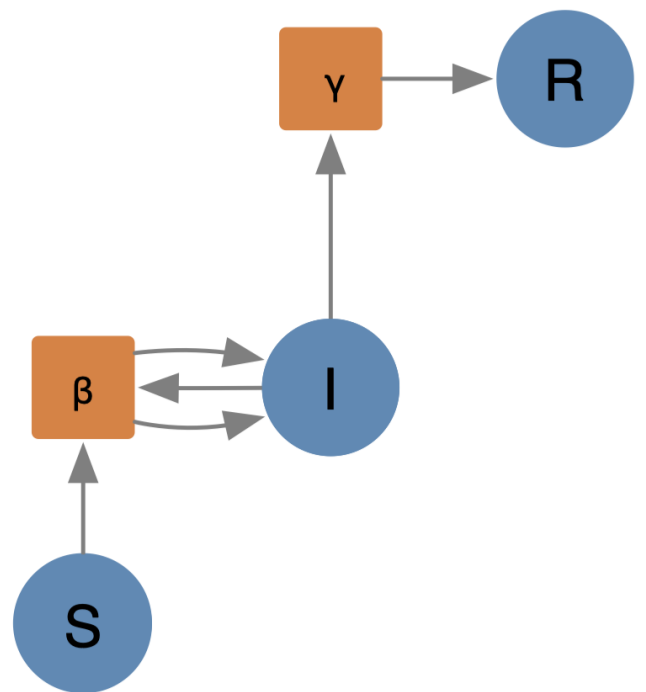
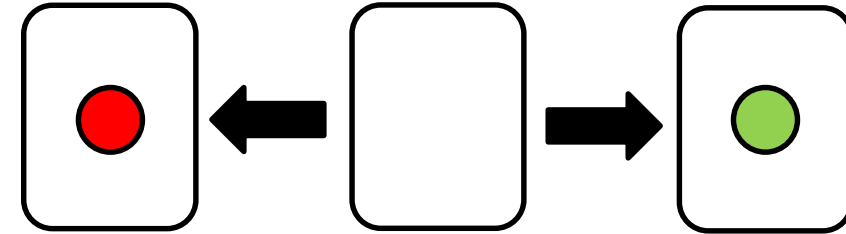
Output	os	ot
1	2	1
2	2	1
3	3	2

Transition	Name
1	infection
2	recovery

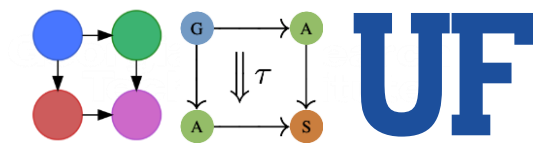
Infect:



Recover:

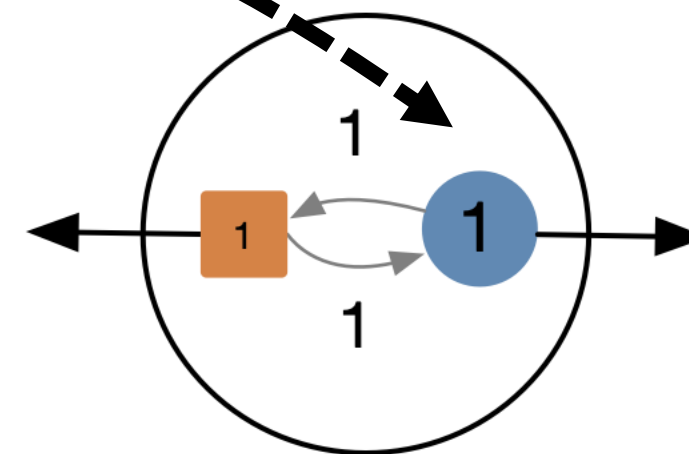
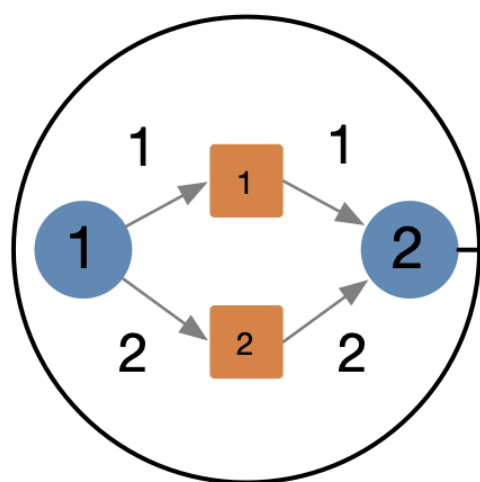


3. Build models compositionally (operad)



Database queries can be built hierarchically using a wiring diagram syntax. Raw SQL queries are not composable this way.

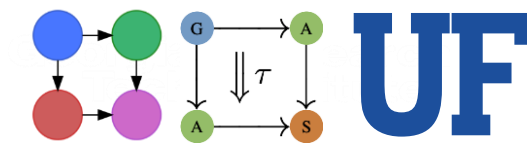
“Find all catalysts (*that are the product of two reactions*) and the reactions they catalyze”



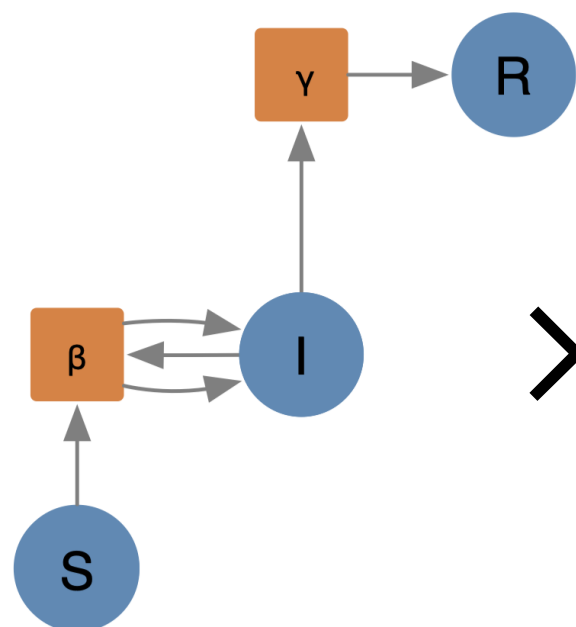
```
SELECT state2.id
FROM   S AS state1, S AS state2,
       T AS tran1,  T AS tran2,
       I AS in1,    I AS in2,
       O AS out1,   O AS out2
WHERE  in1.is = state1.id,
       in1.it = tran1.id
       out1.os = state2.id
       out1.ot = tran1.id
...
```

```
SELECT tran1.id, state1.id
FROM   S AS state1, T AS tran1,
       I AS in1,    O AS out1
WHERE  in1.is = state1.id,
       in1.it = tran1.id
       out1.os = state1.id
       out1.ot = tran1.id
```

4. Build models compositionally (limits)

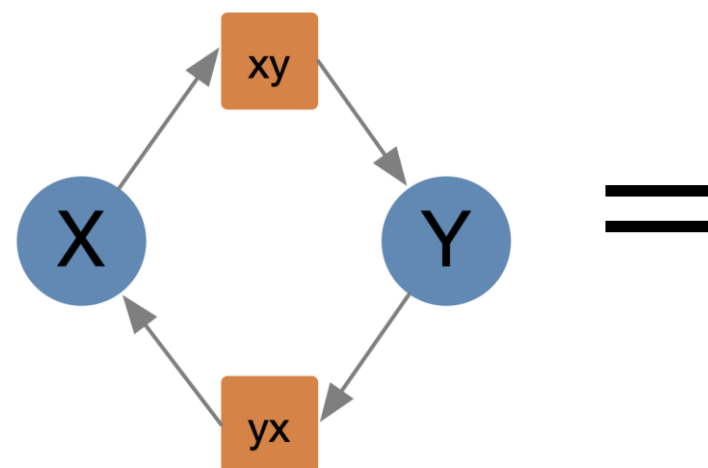


“SIR model”



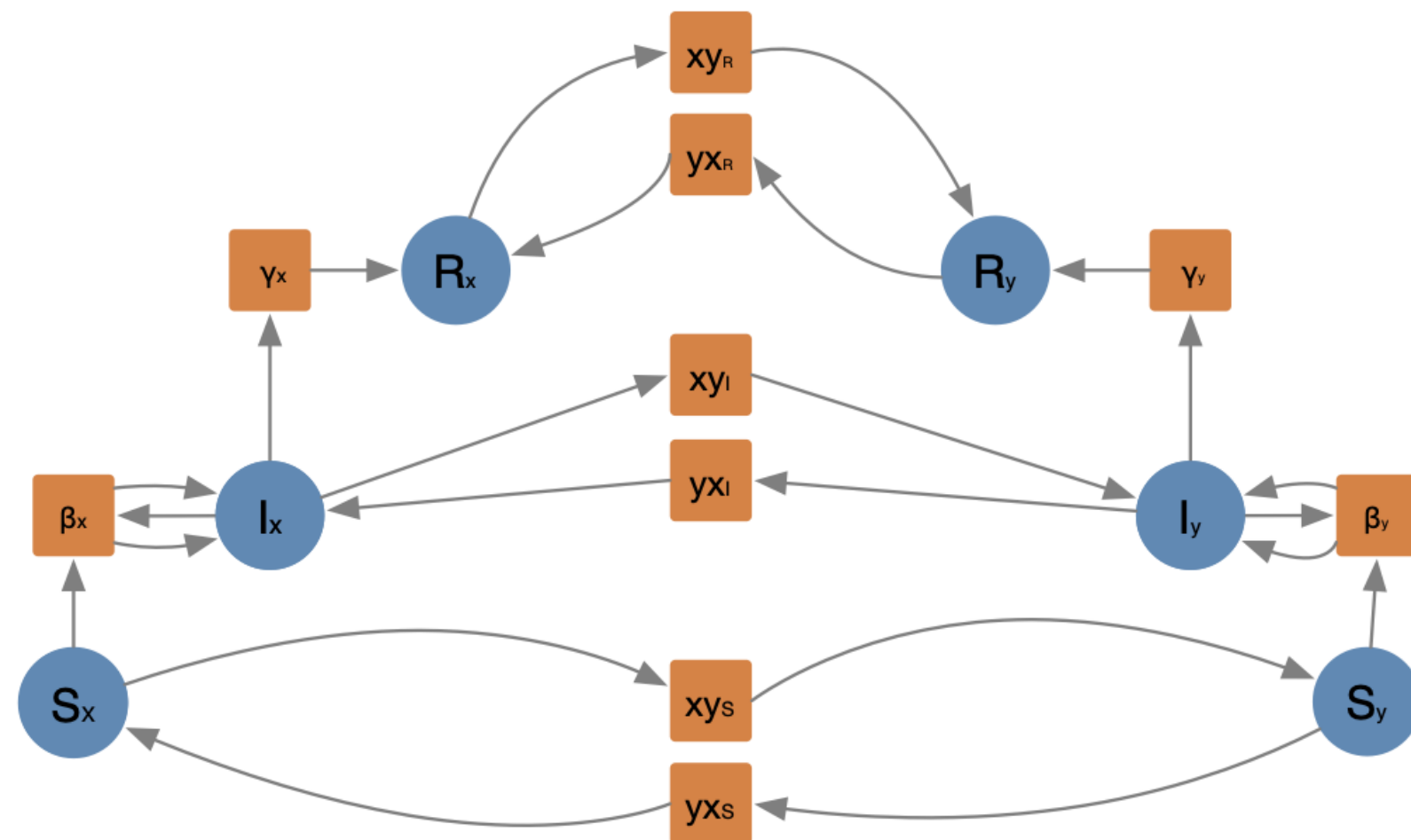
\times

“Two-city model”



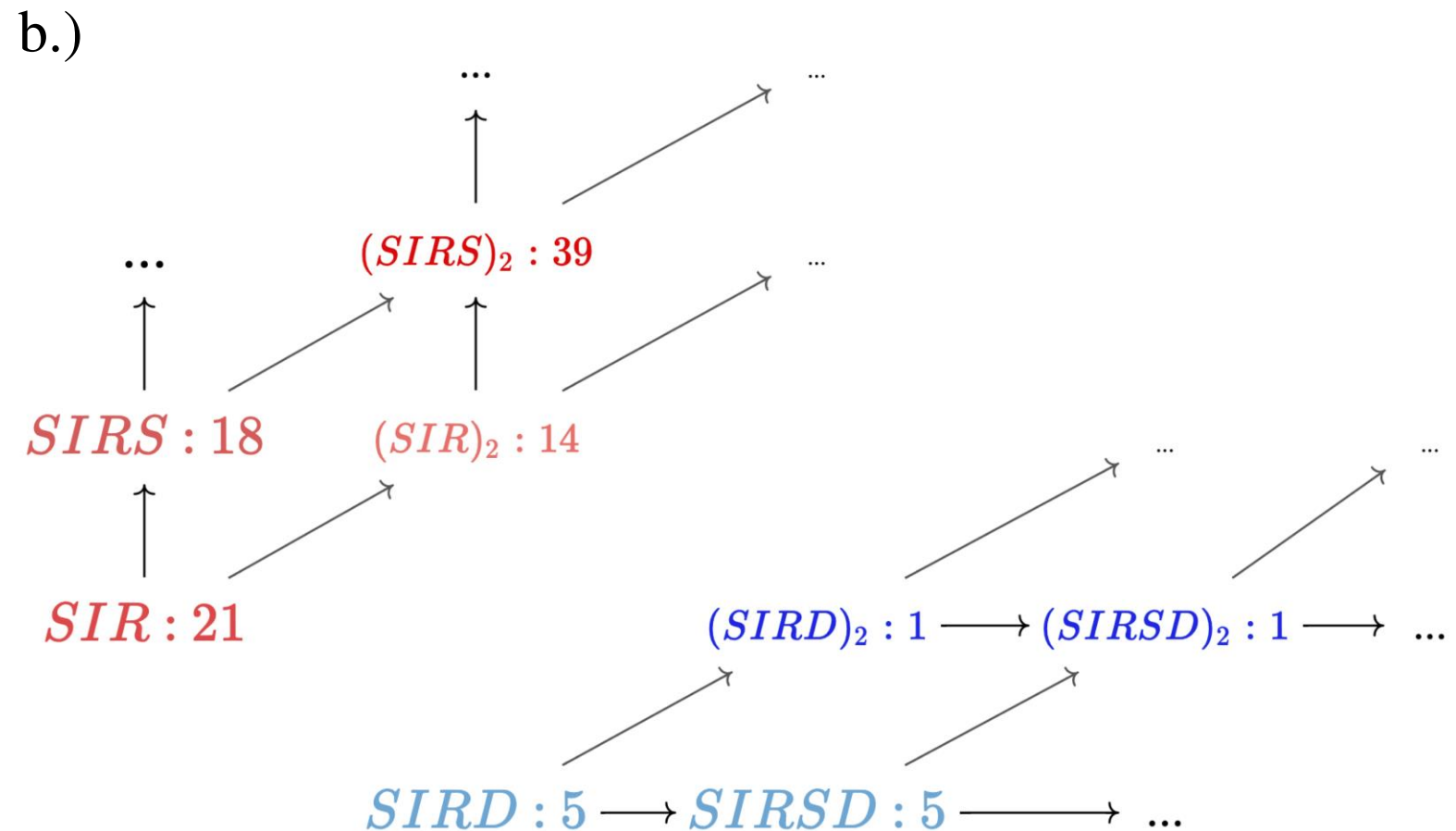
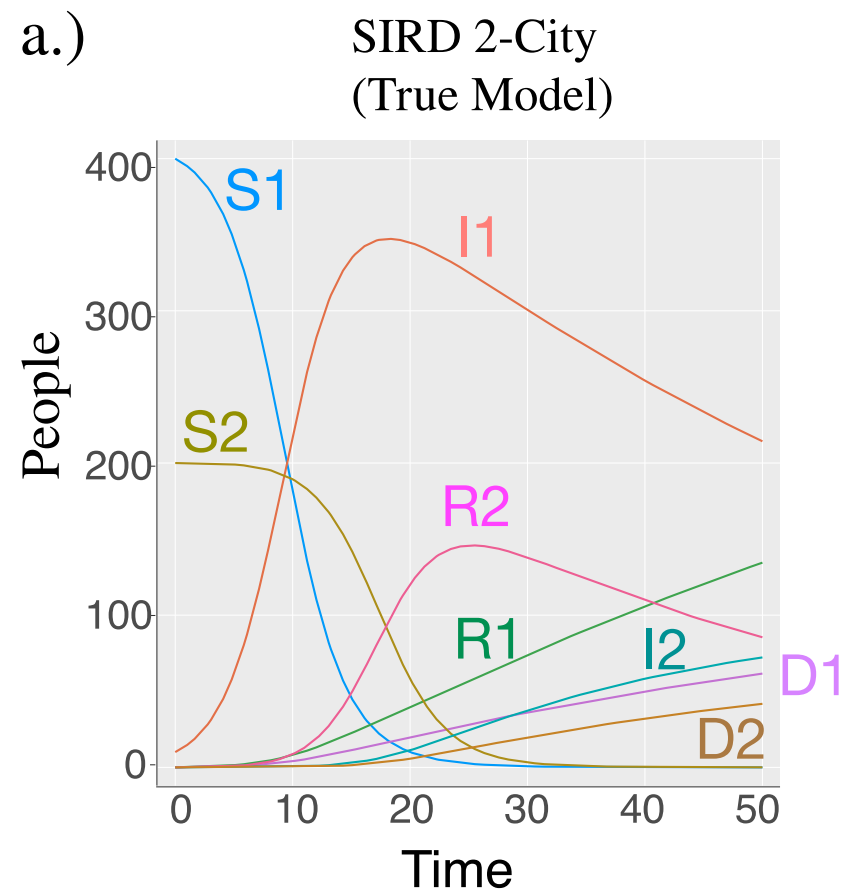
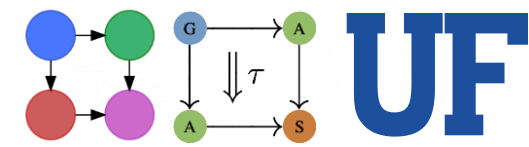
$=$

“Two-city SIR model”

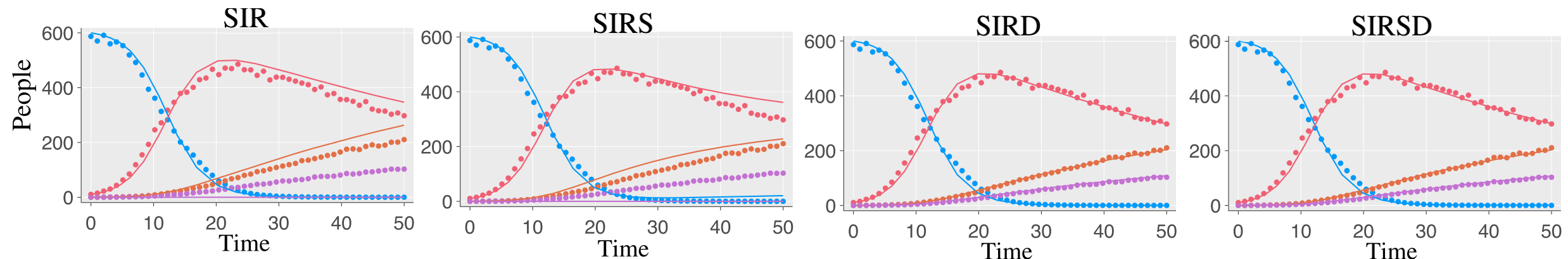


High-level operations on Petri nets like products do the ‘right’ thing’ for reaction networks, unlike for symbolic syntax or raw ODEs.

5. Automated Model Selection



c.) — Model ••••• Data Σ Susceptible Σ Infected Σ Recovered Σ Dead



Introduction

- Why represent scientific models combinatorially?

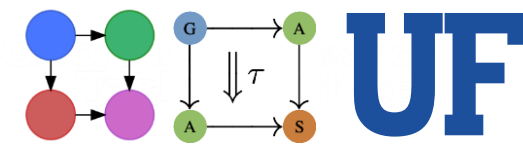
Model space exploration

- The category of diagrams as a category of model spaces
- Example limits and colimits
- Composition recipes
- Limits and colimits: implementation

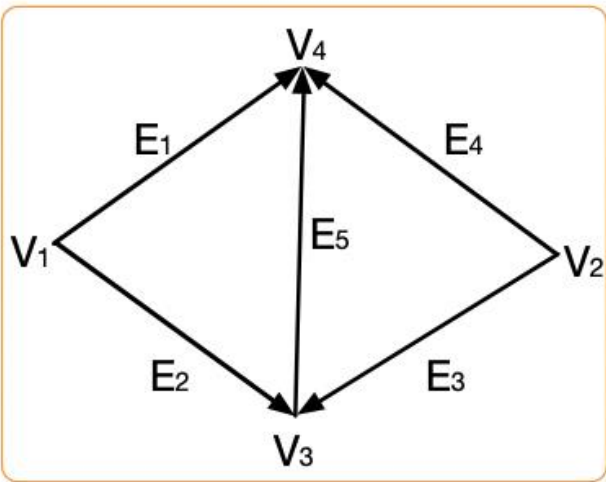
Model Selection

- Best fit chemical reaction network example

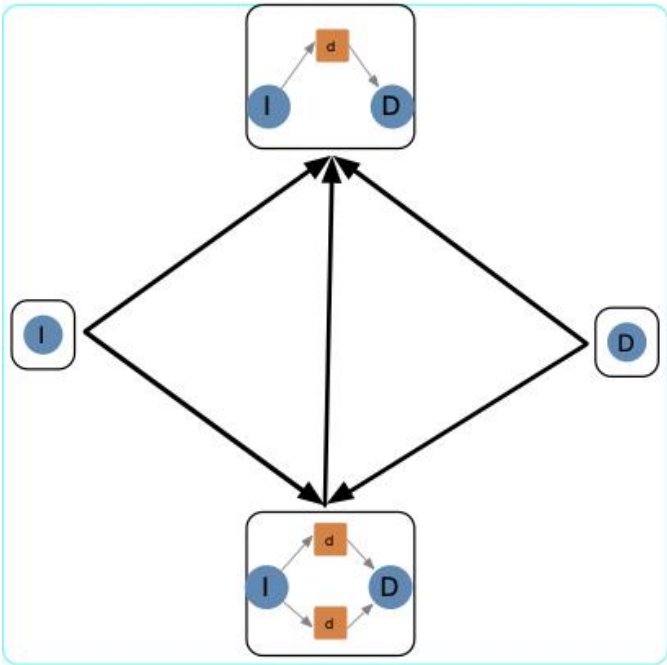
The category of diagrams



A particular
diagram in Petri

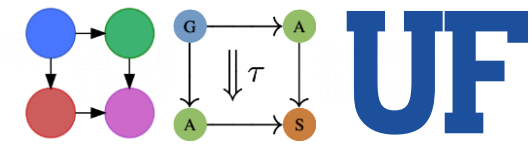


Shape

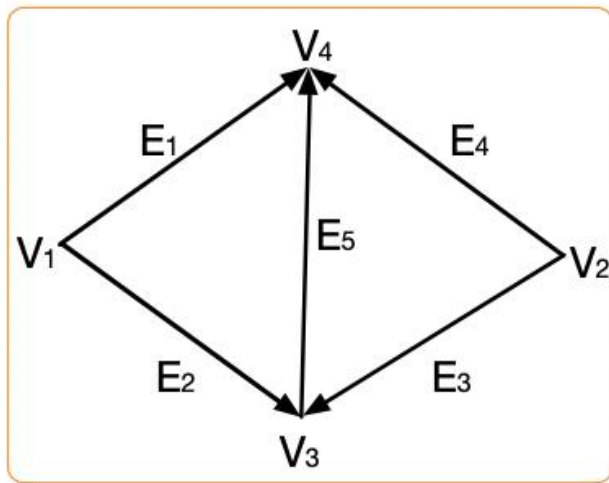


Diagram

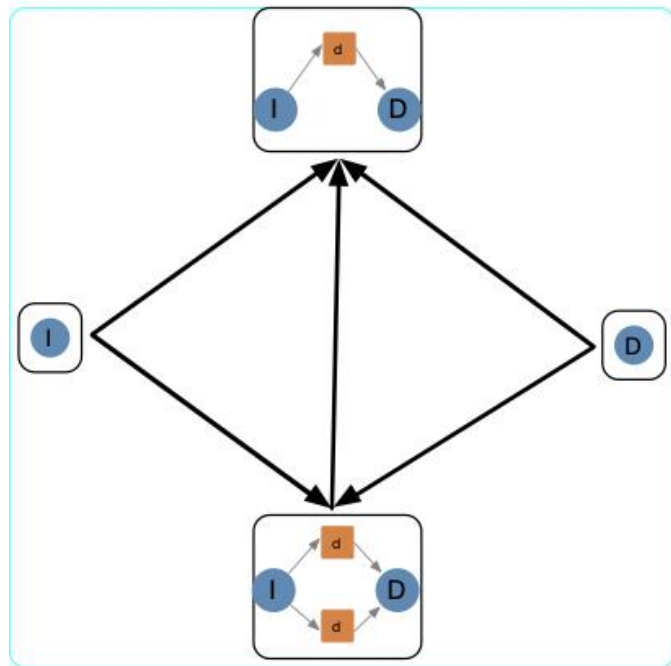
The category of diagrams: a lax slice category



A particular
diagram in Petri



Shape

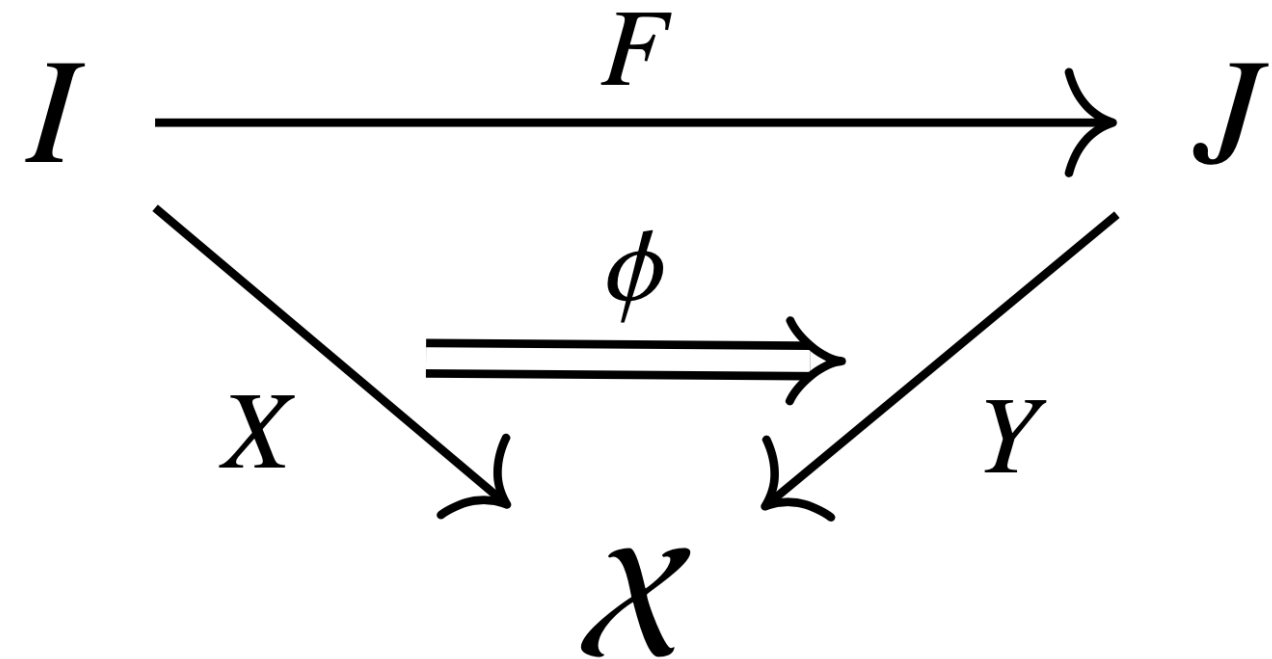


Diagram

$$(F, \phi): (I, X) \rightarrow (J, Y)$$

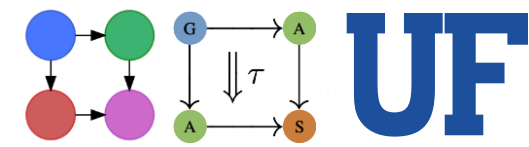
Shape map: F

Diagram map: ϕ

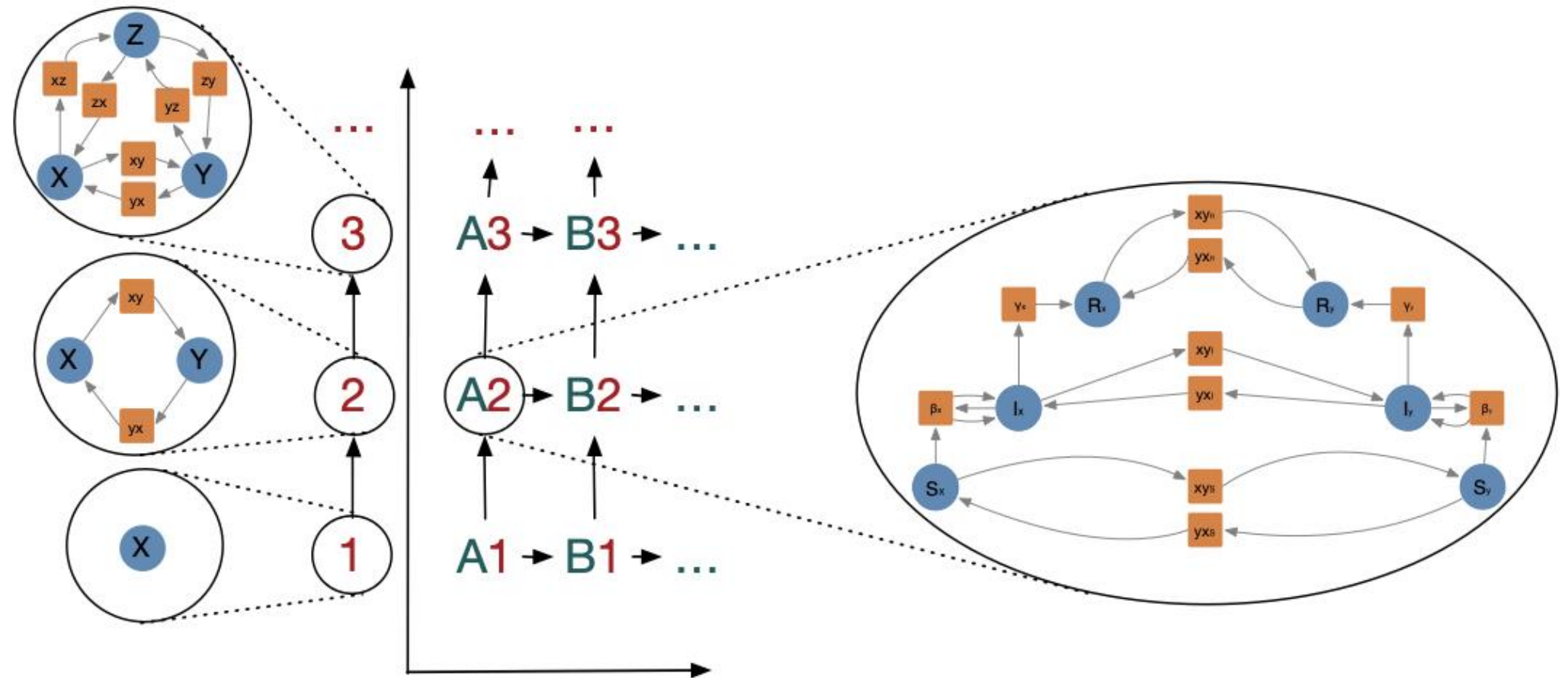


We want pushouts and pullbacks of diagrams

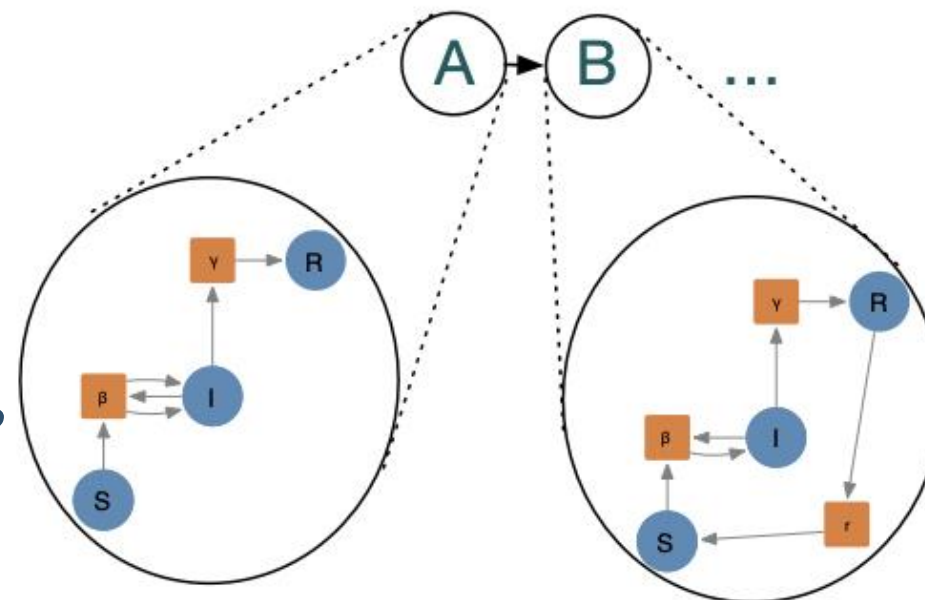
Model space product



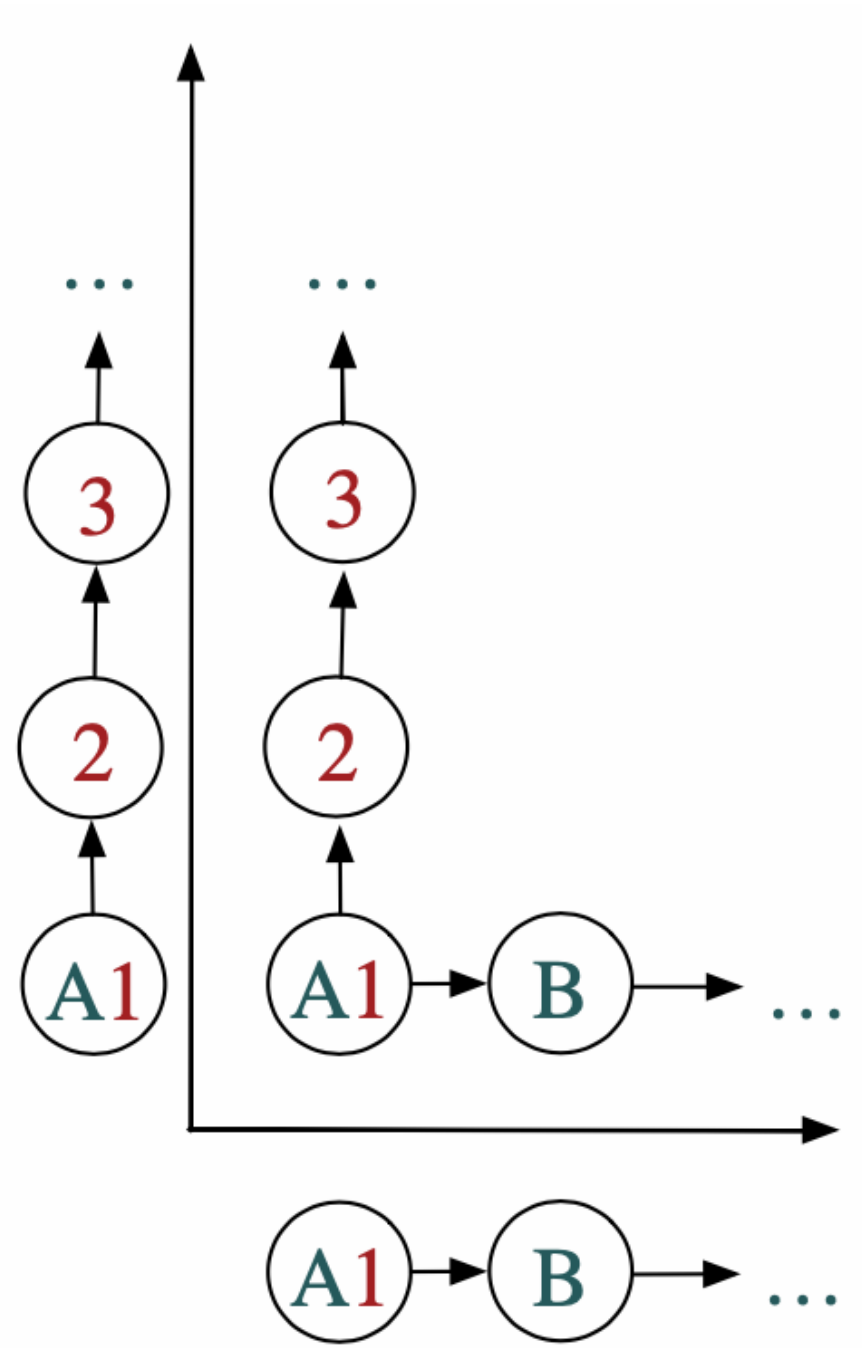
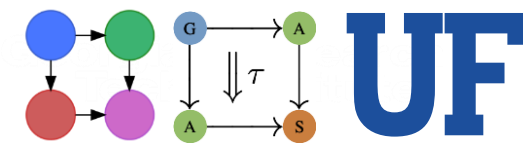
“City dimension”



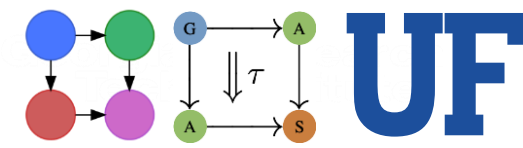
“Disease dimension”



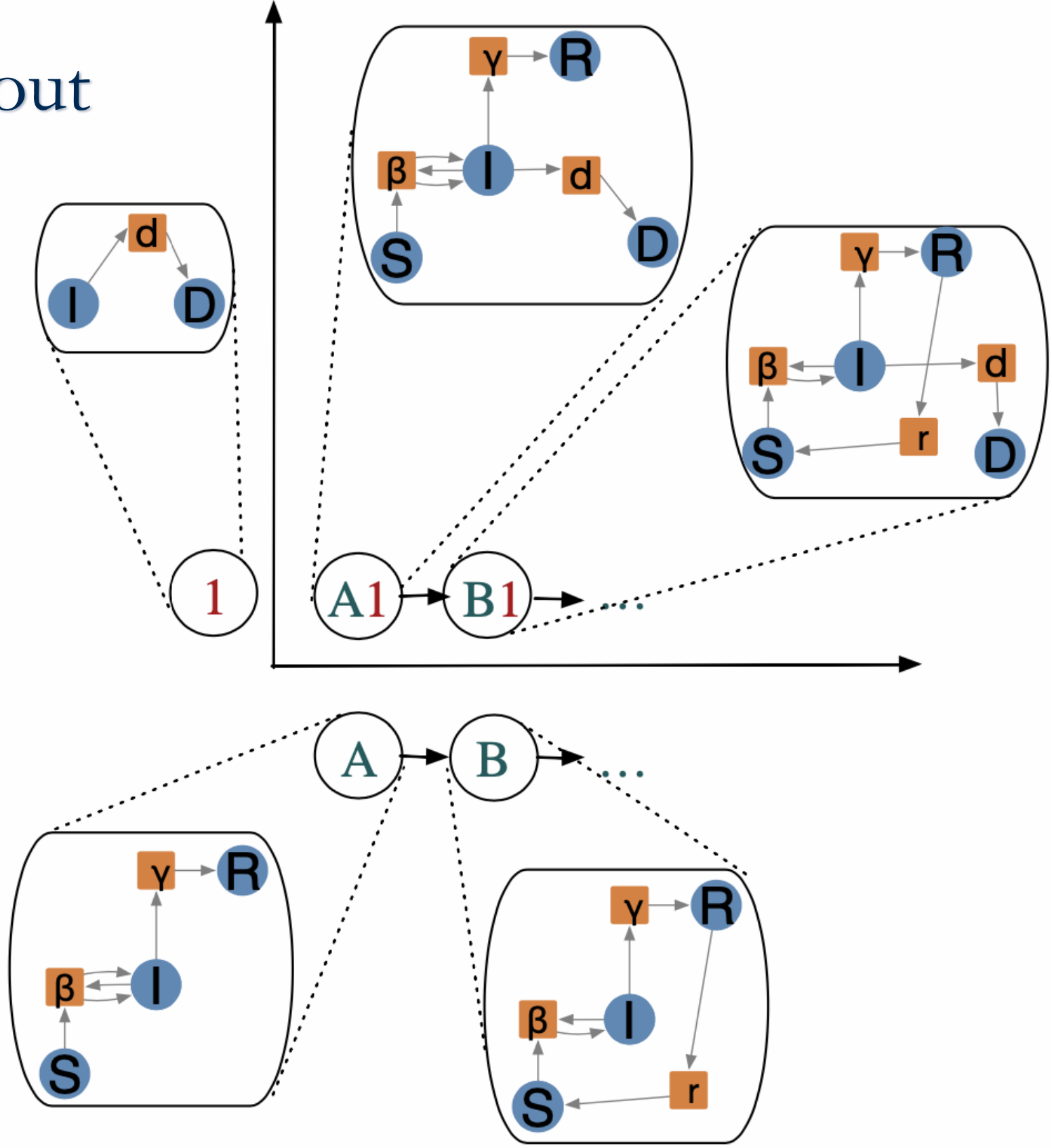
Model space pushout



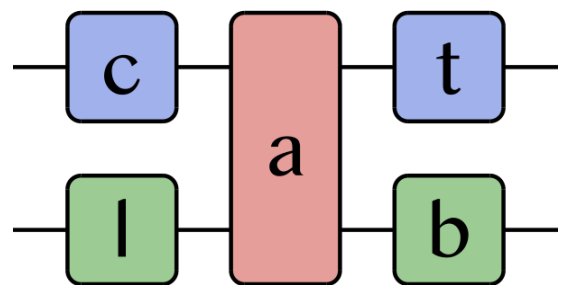
Model space pushout



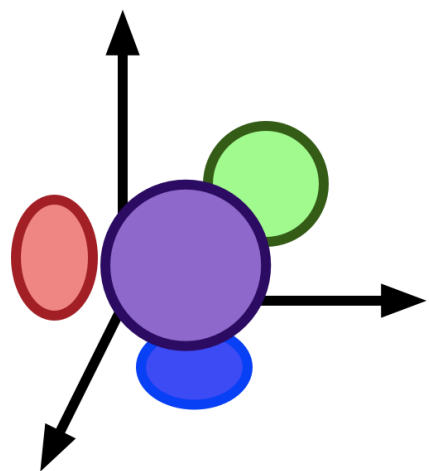
“*Glue*”



Composition

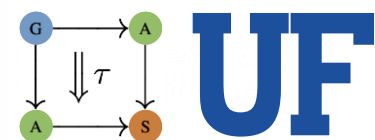
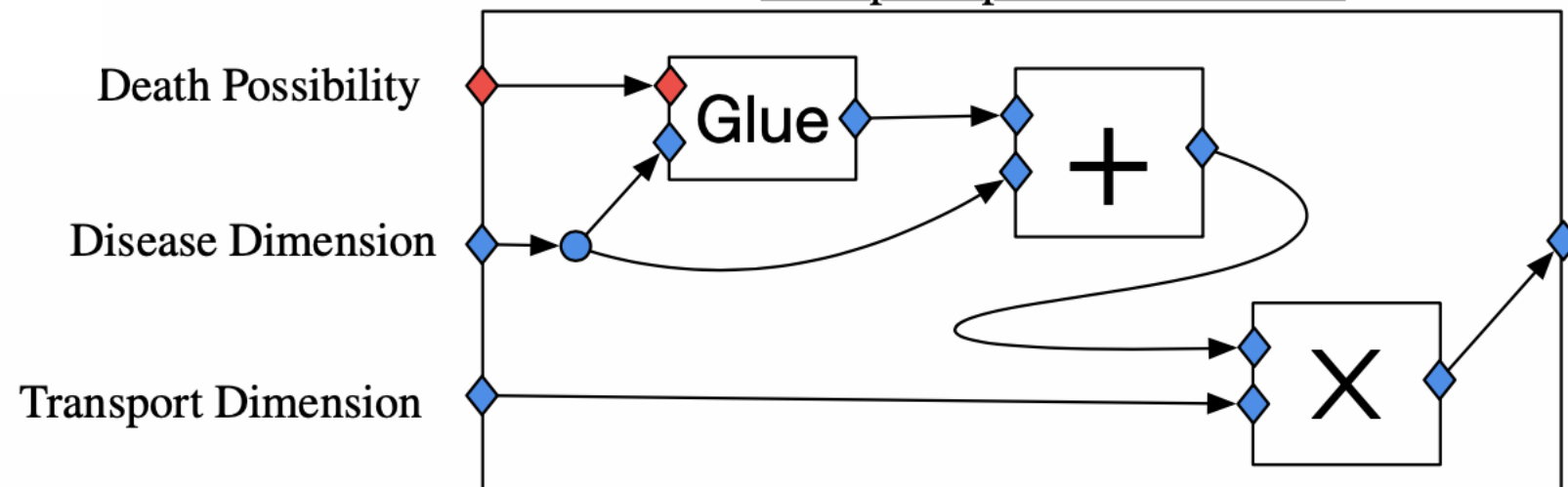


Catlab.jl

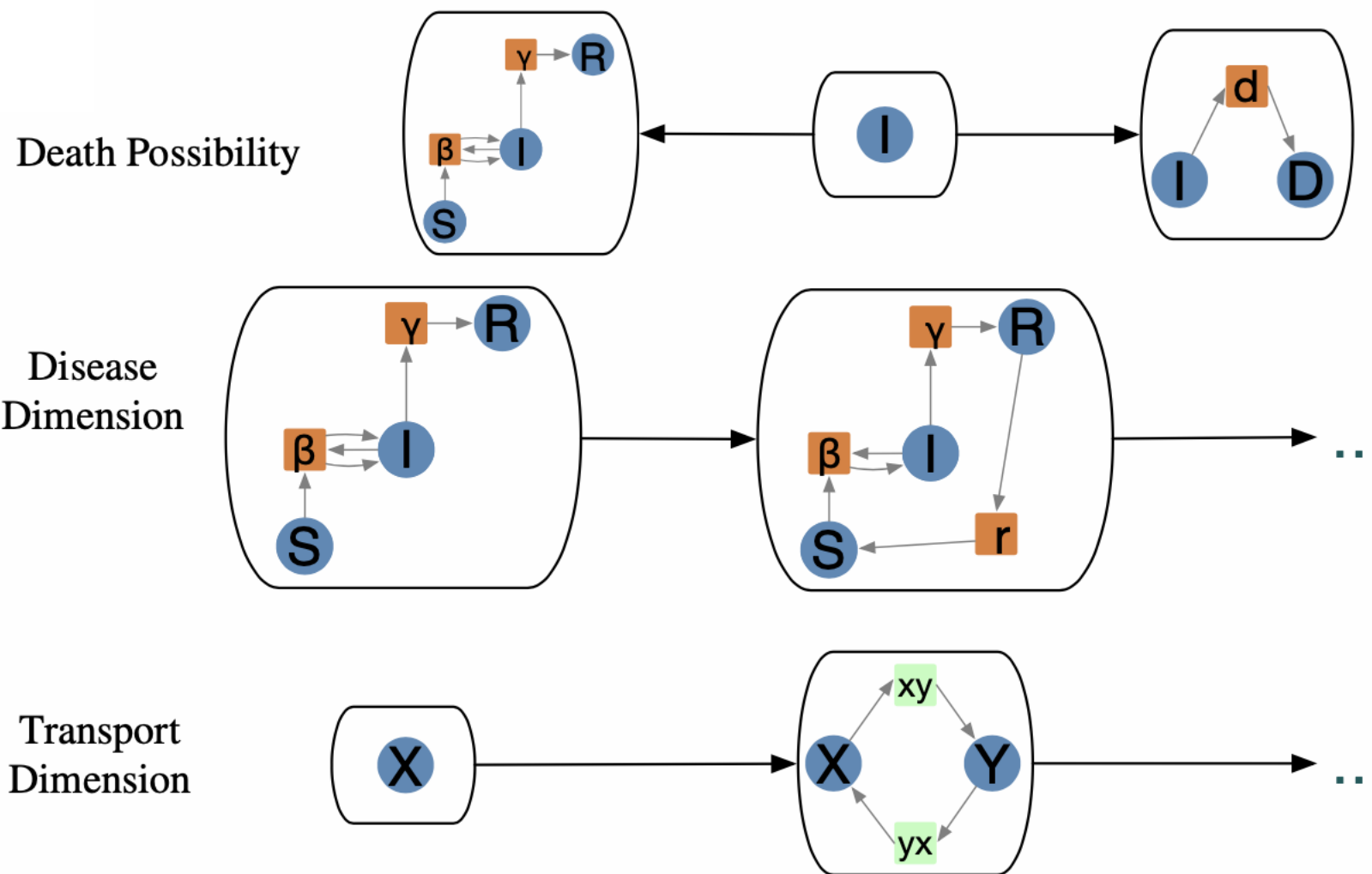


ModelExploration.jl

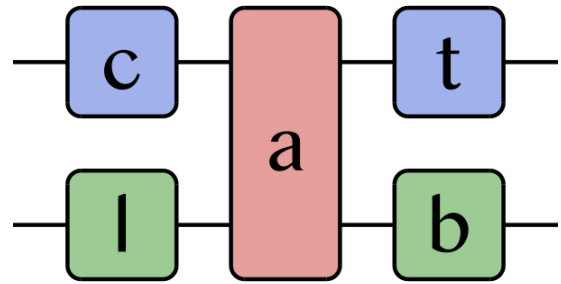
Example exploration workflow



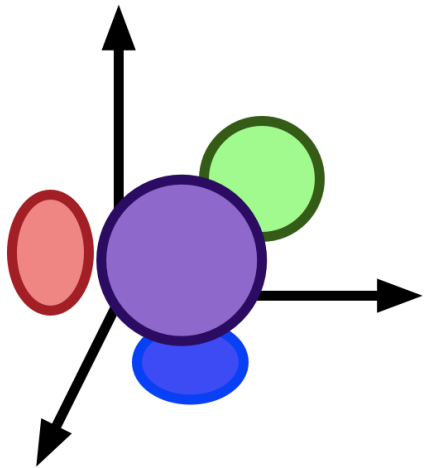
UF



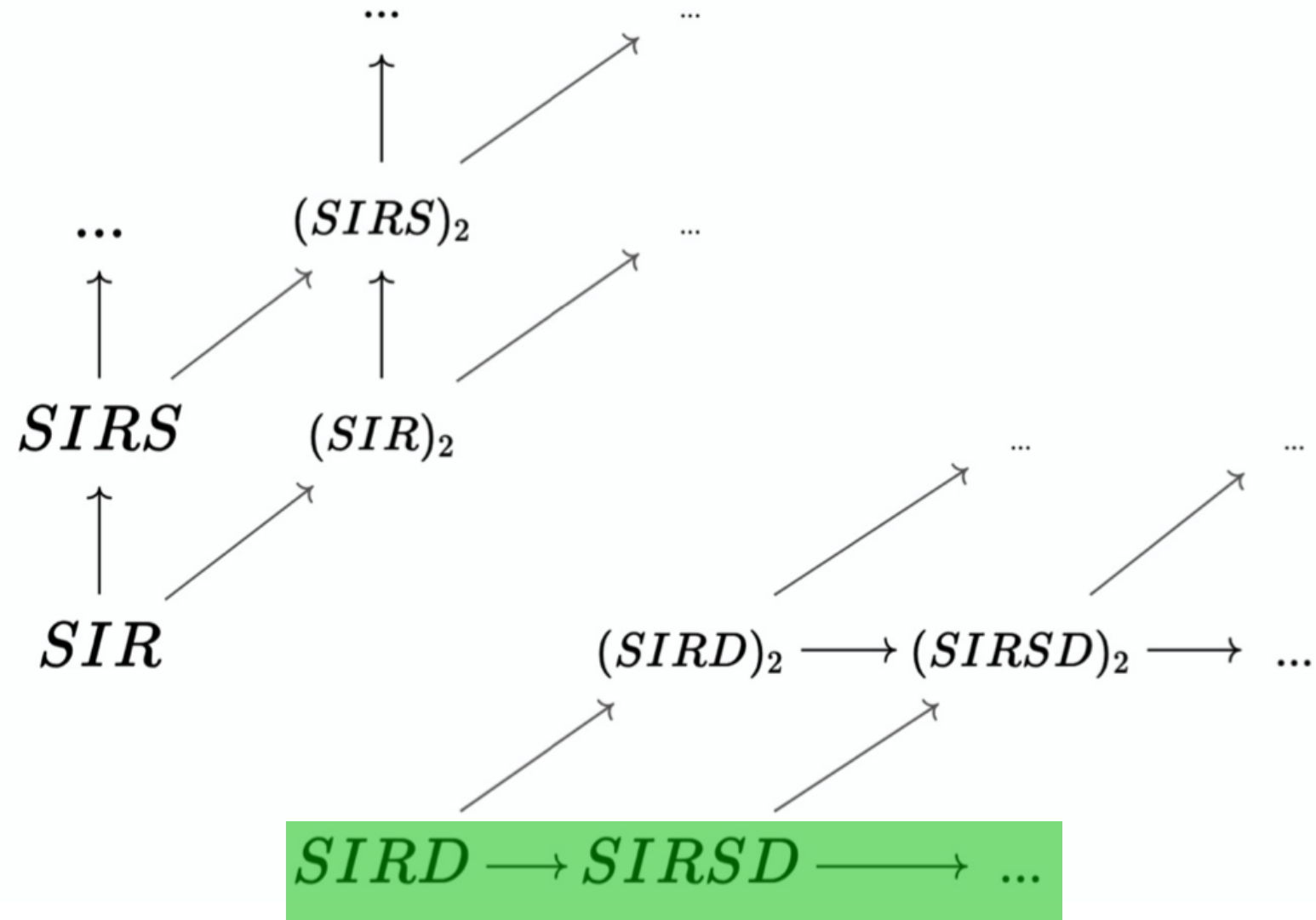
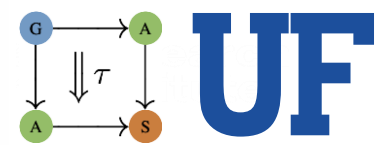
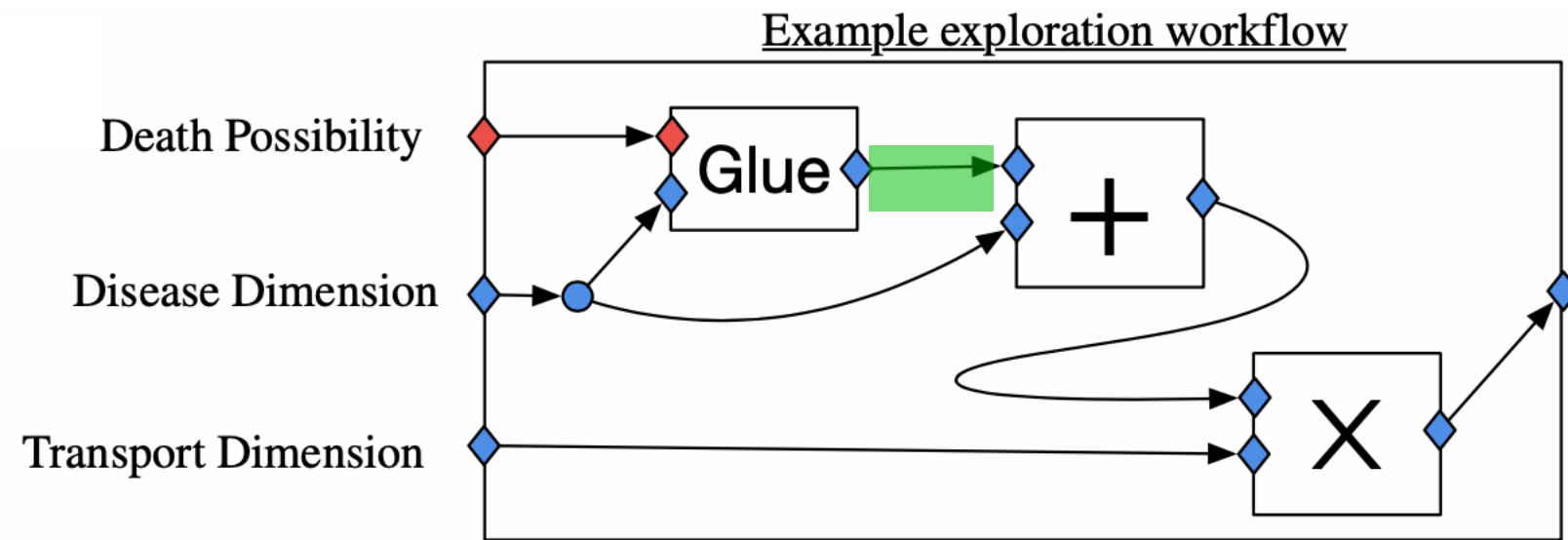
Composition result



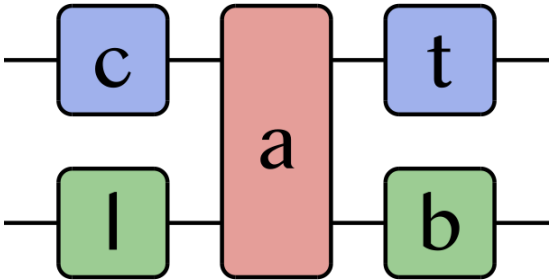
Catlab.jl



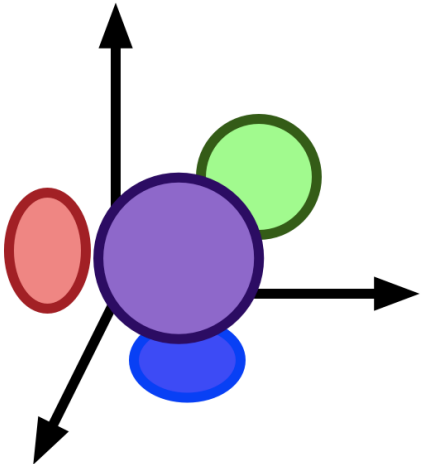
ModelExploration.jl



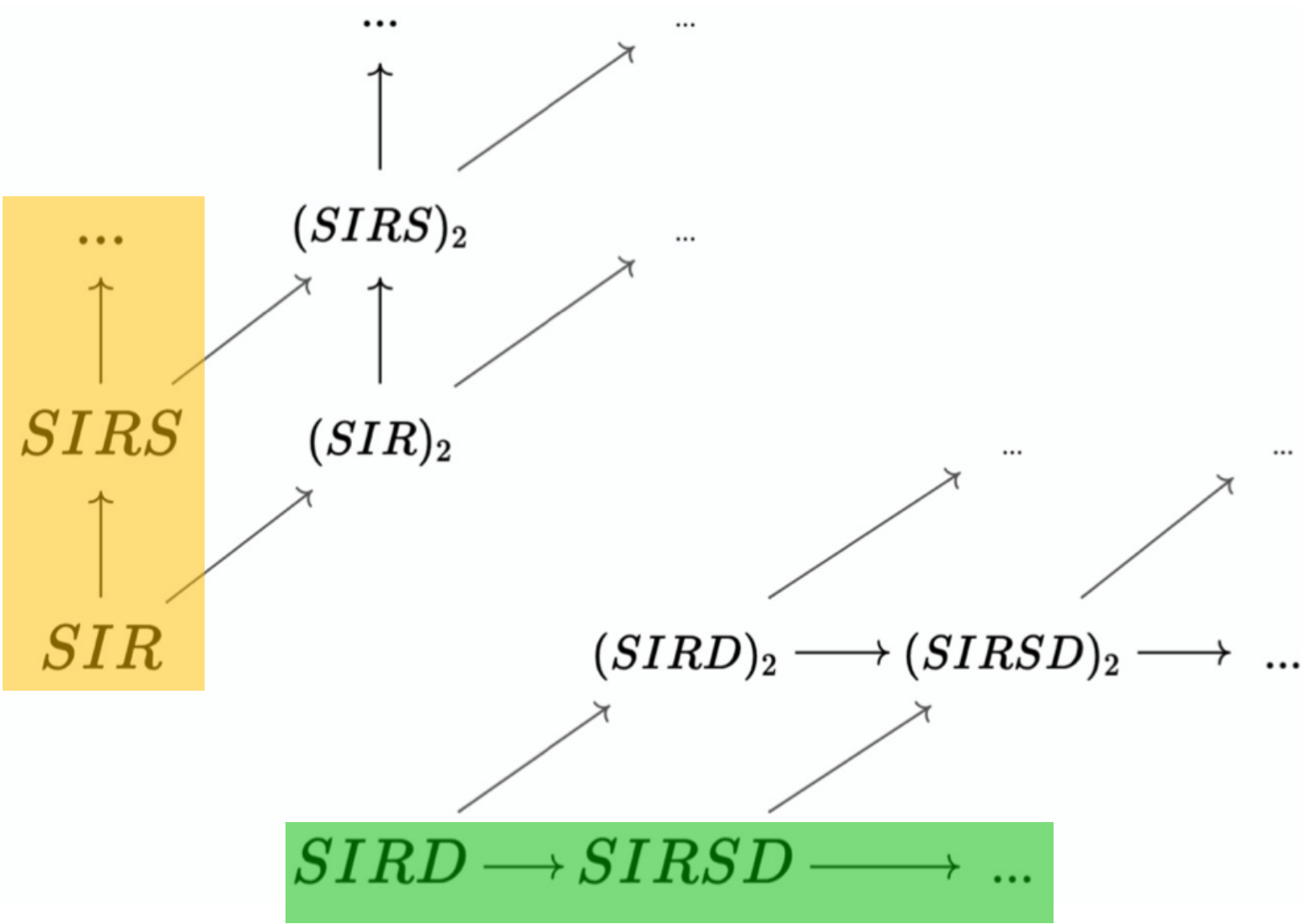
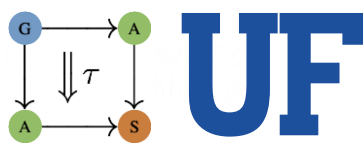
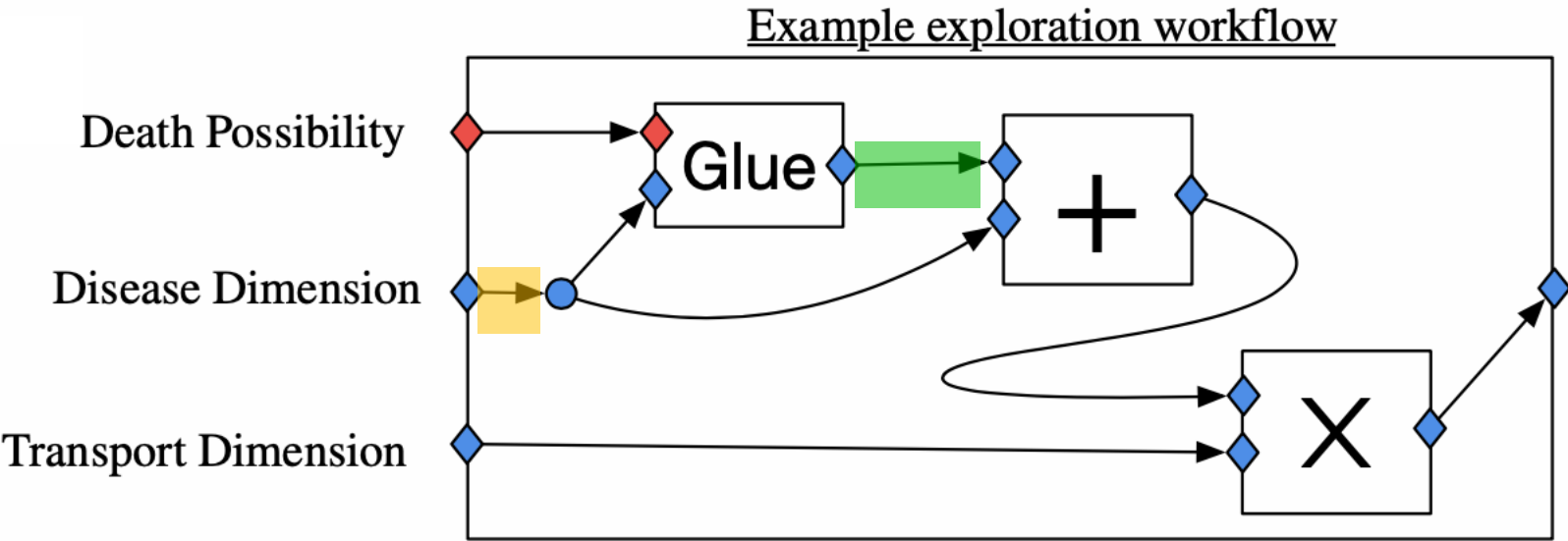
Composition result



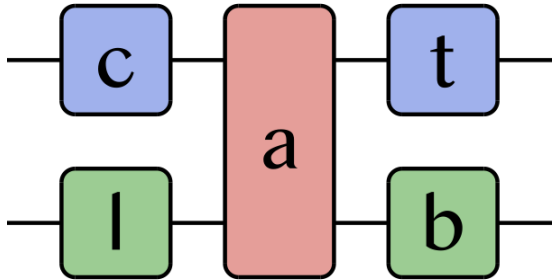
Catlab.jl



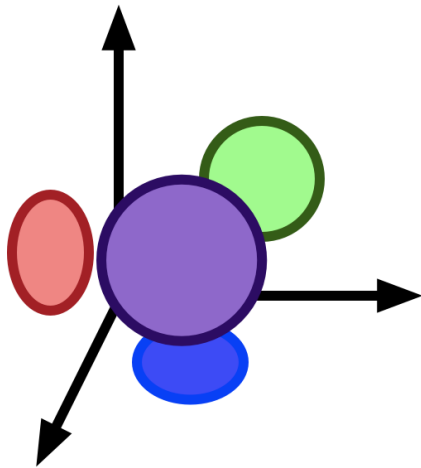
ModelExploration.jl



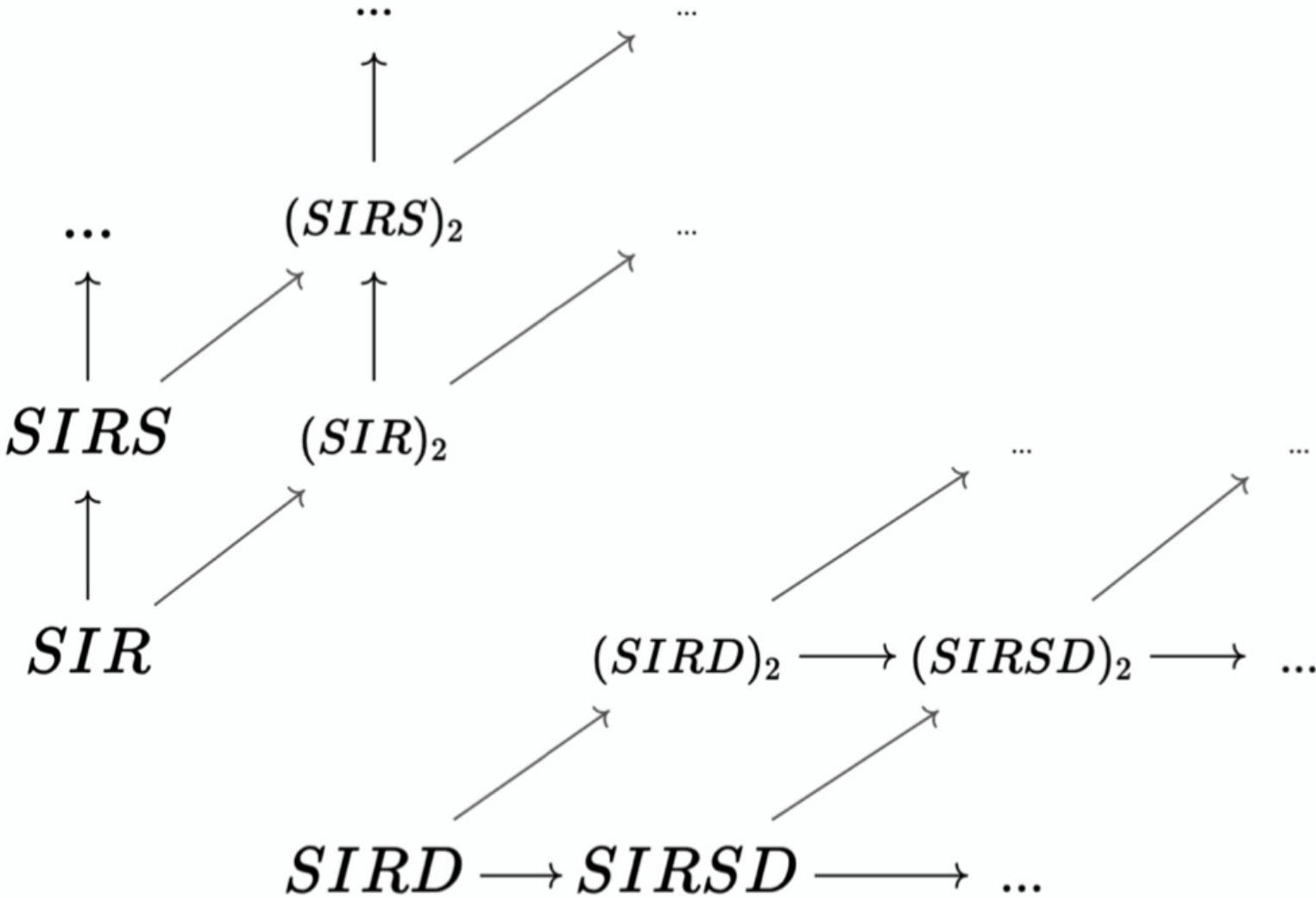
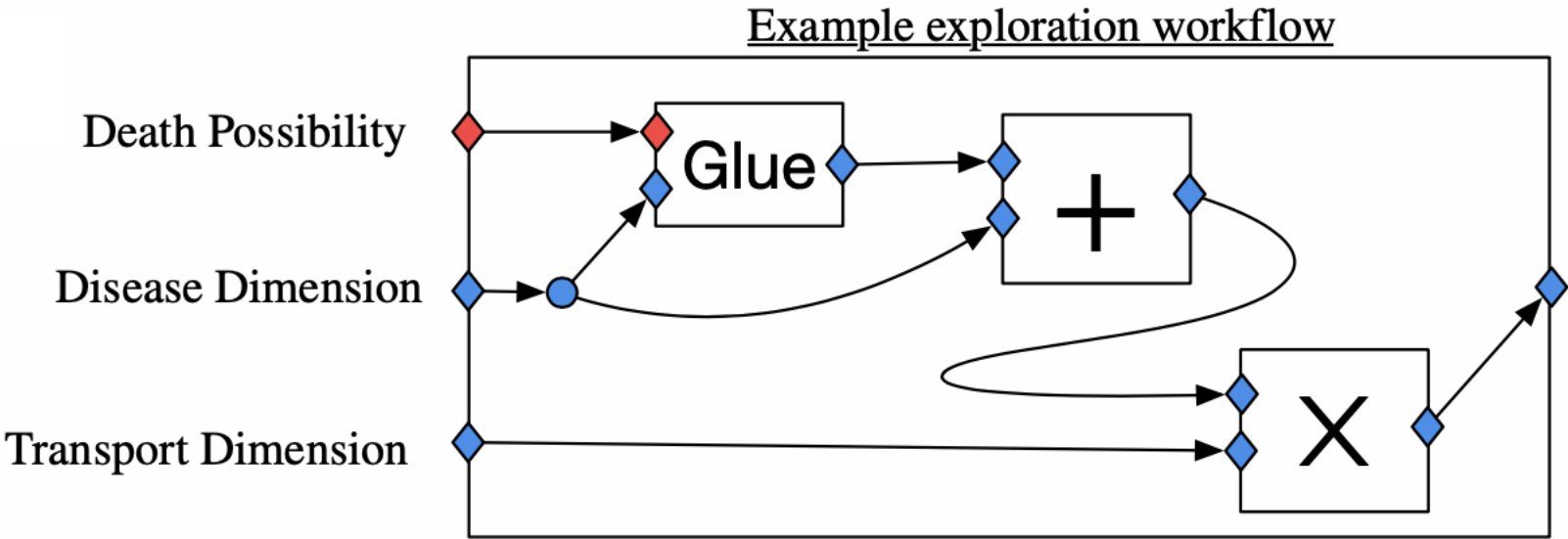
Composition result



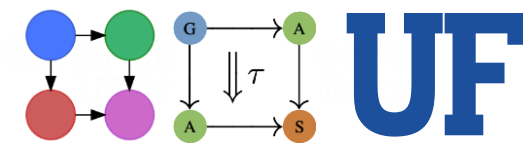
Catlab.jl



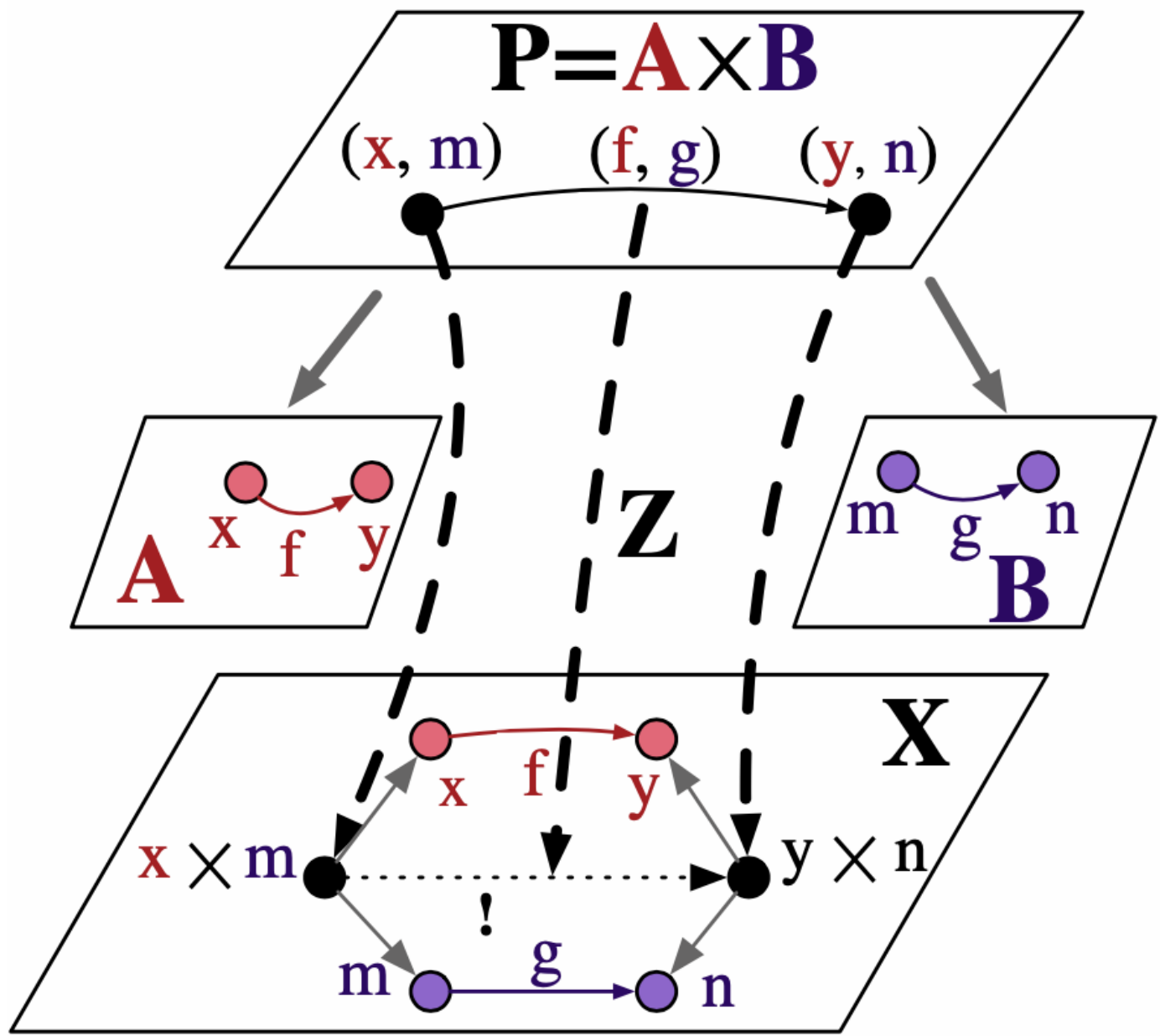
ModelExploration.jl



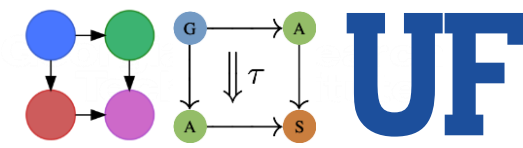
Limits - implementation



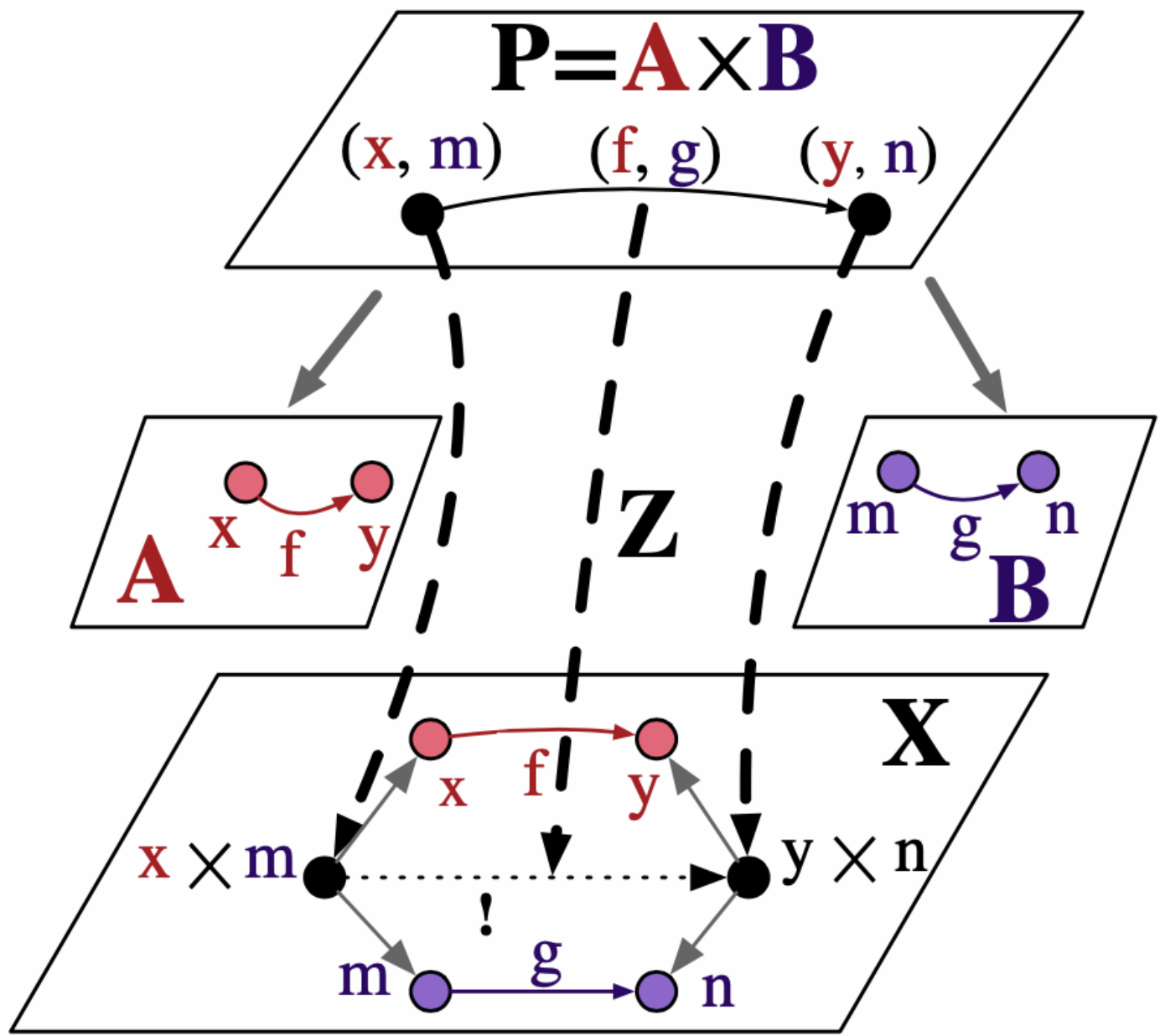
Product



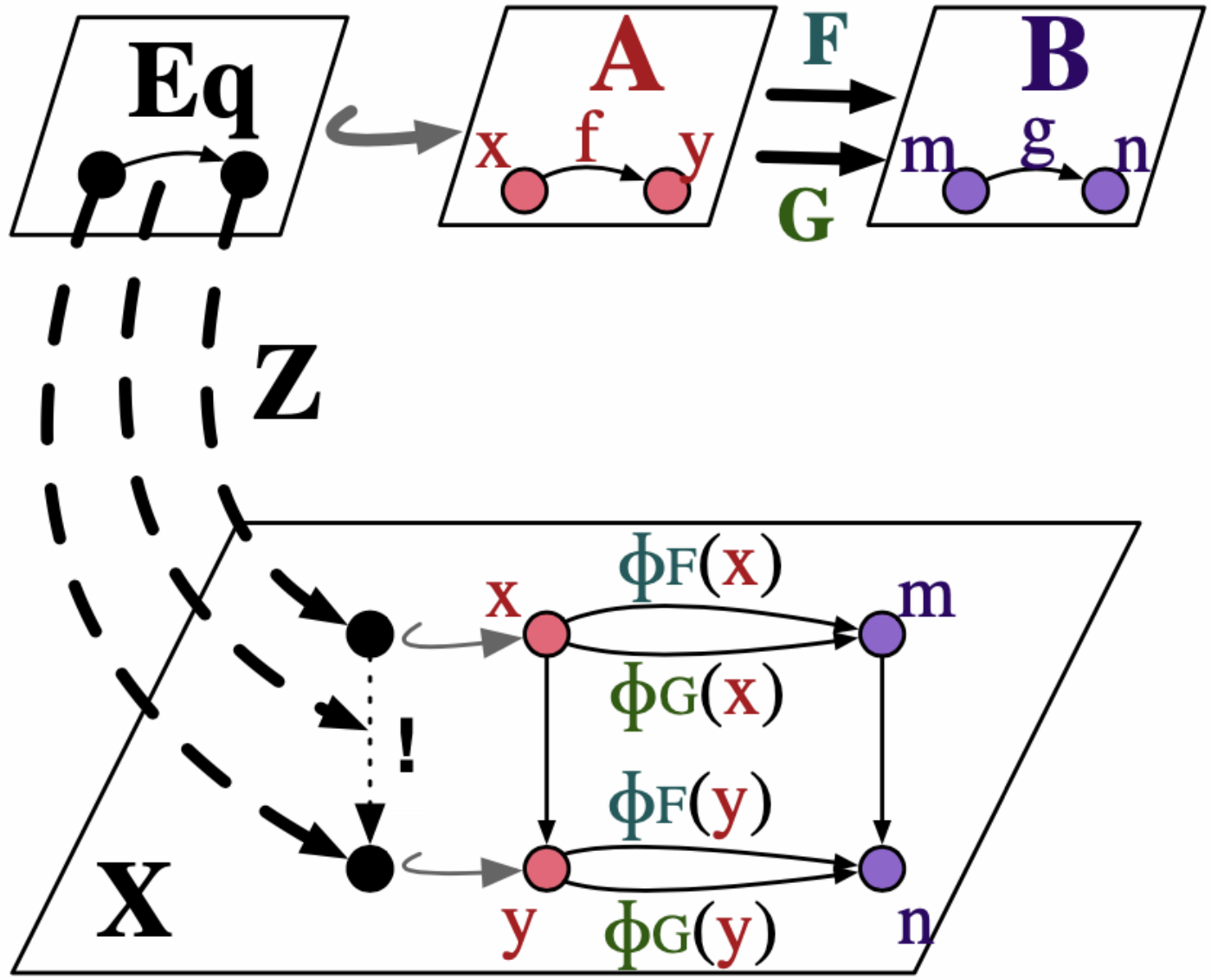
Limits - implementation



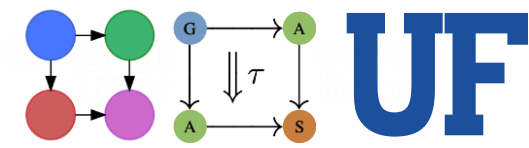
Product



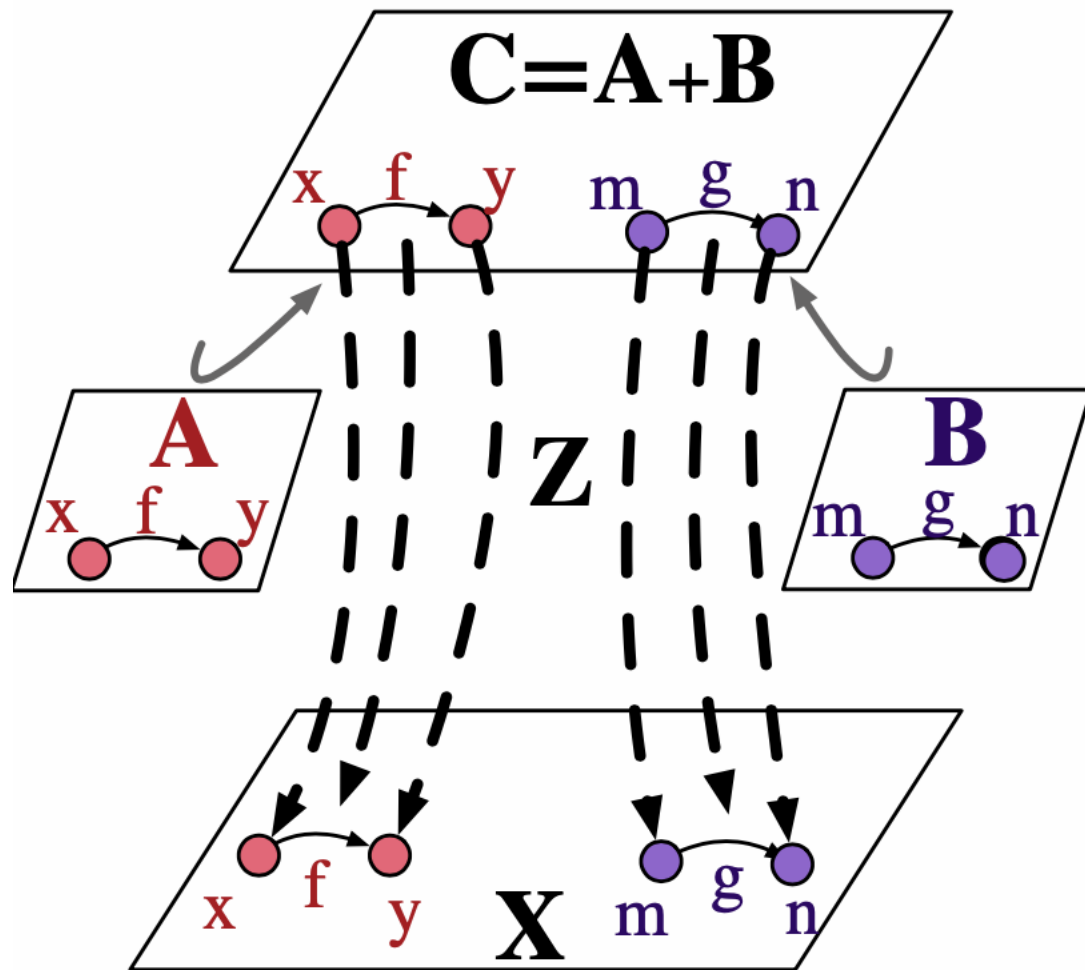
Equalizer



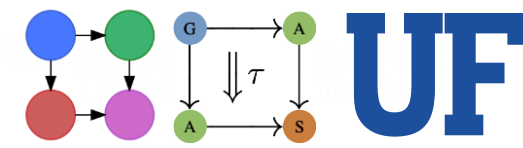
Colimits - implementation



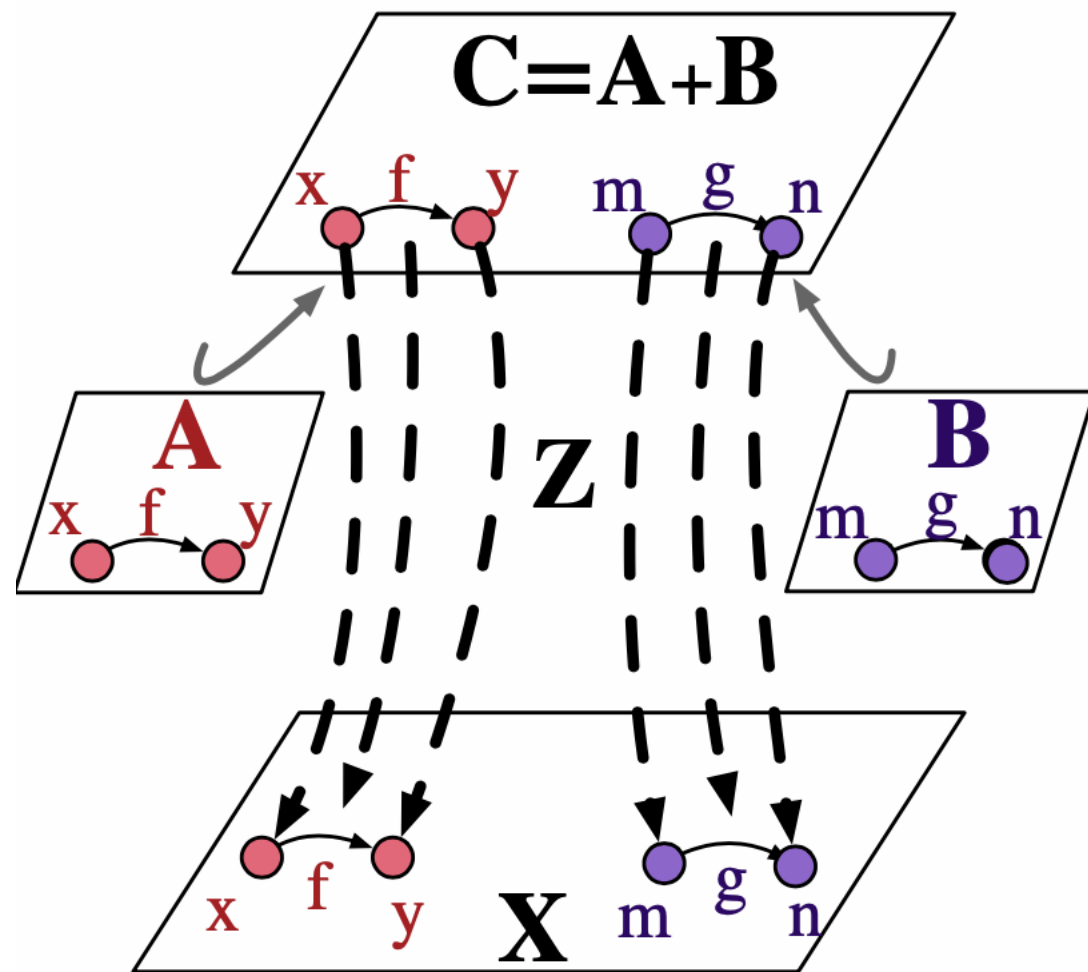
Coproduct



Colimits - implementation



Coproduct



Coequalizer

$$\begin{array}{ccc}
 (A, X) & \xrightarrow{(FH, \lambda)} & (C, Lan_{FH} X) \\
 (F, \phi_i) \downarrow \downarrow \downarrow & & \downarrow \downarrow \downarrow (id_C, \alpha_i) \\
 (B, Y) & \xrightarrow{(H, \kappa)} & (C, Lan_H Y) \\
 & \searrow (H, \kappa\gamma) & \downarrow (id_C, \gamma) \\
 & & (C, Z)
 \end{array}$$



Introduction

- Why represent scientific models combinatorially?

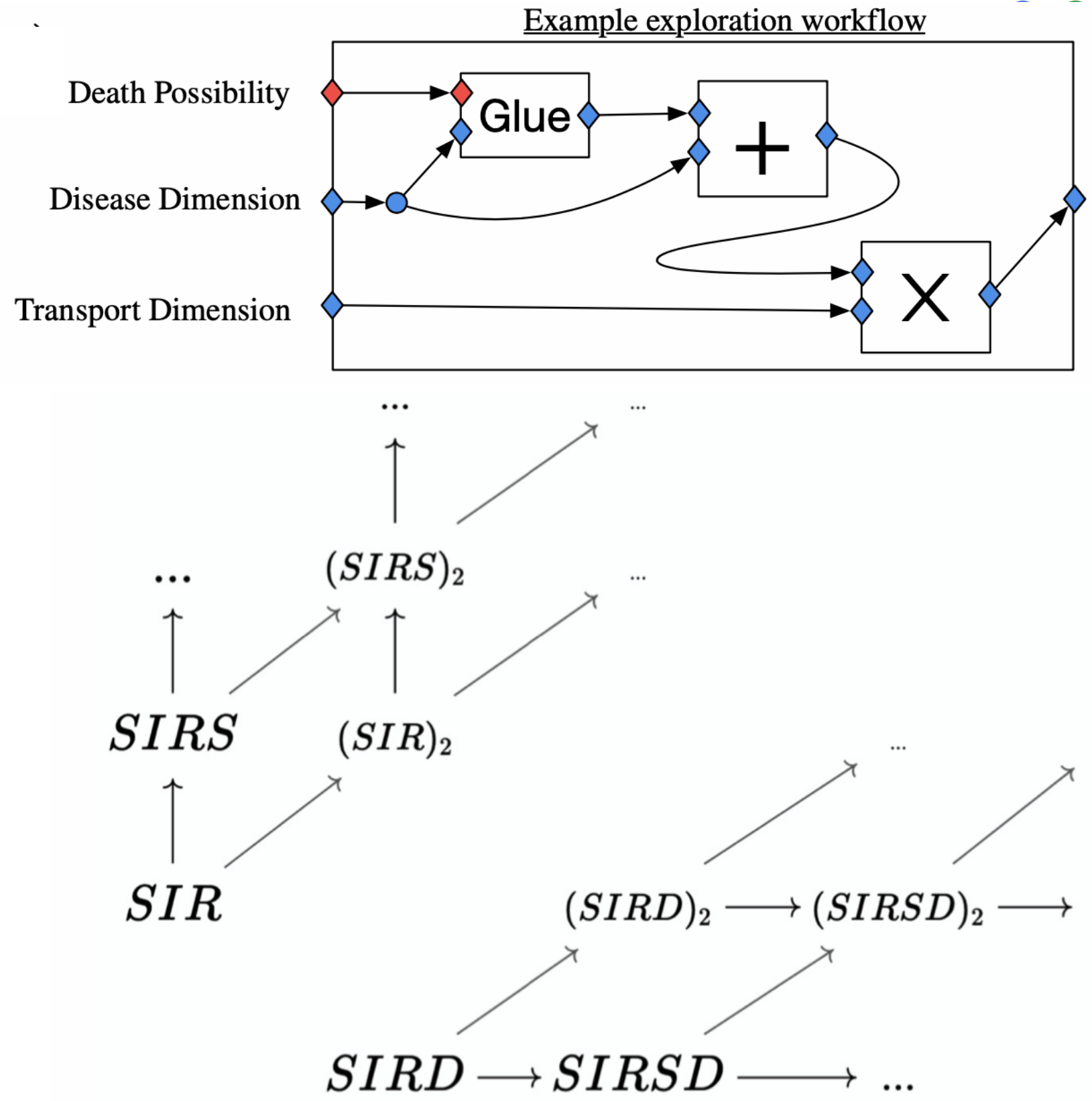
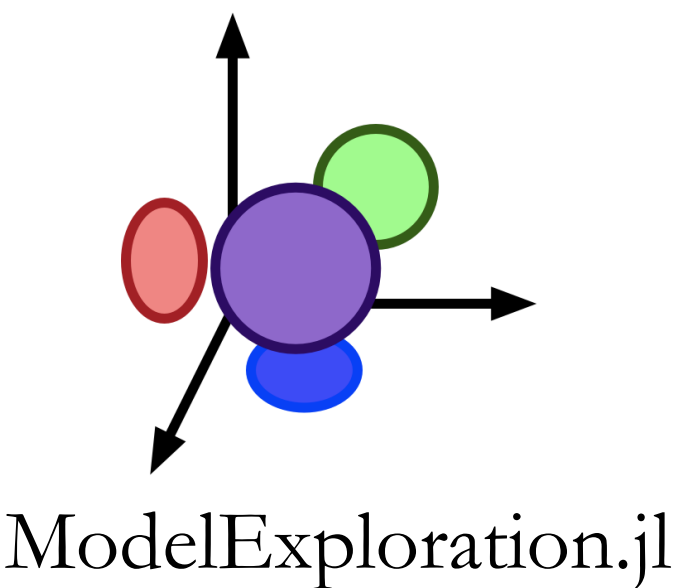
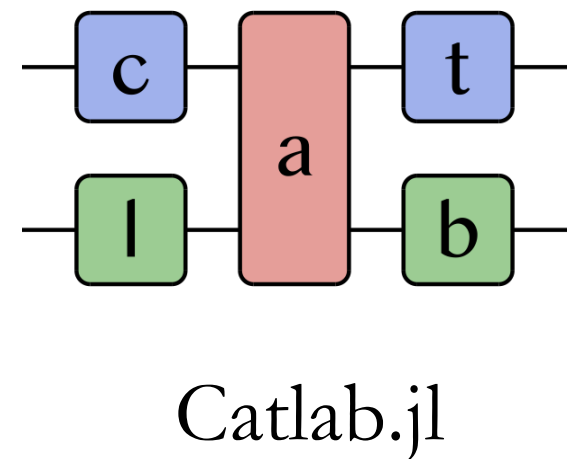
Model space exploration

- The category of diagrams as a category of model spaces
- Example limits and colimits
- Composition recipes
- Limits and colimits: implementation

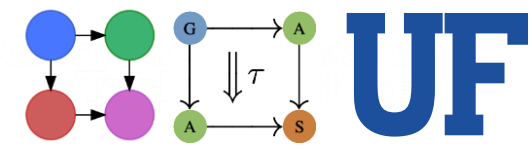
Model Selection

- Best fit chemical reaction network example

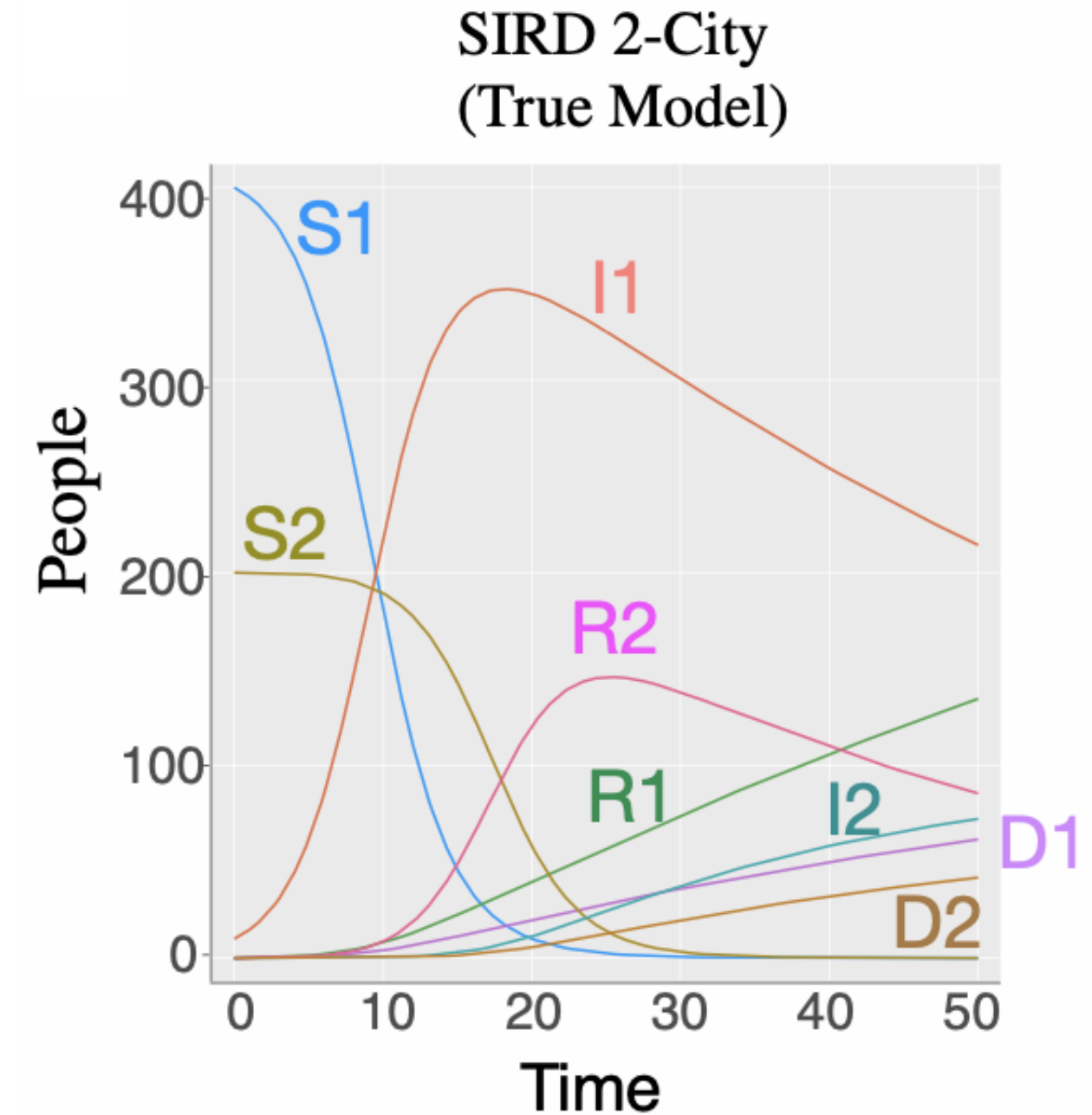
Composition result



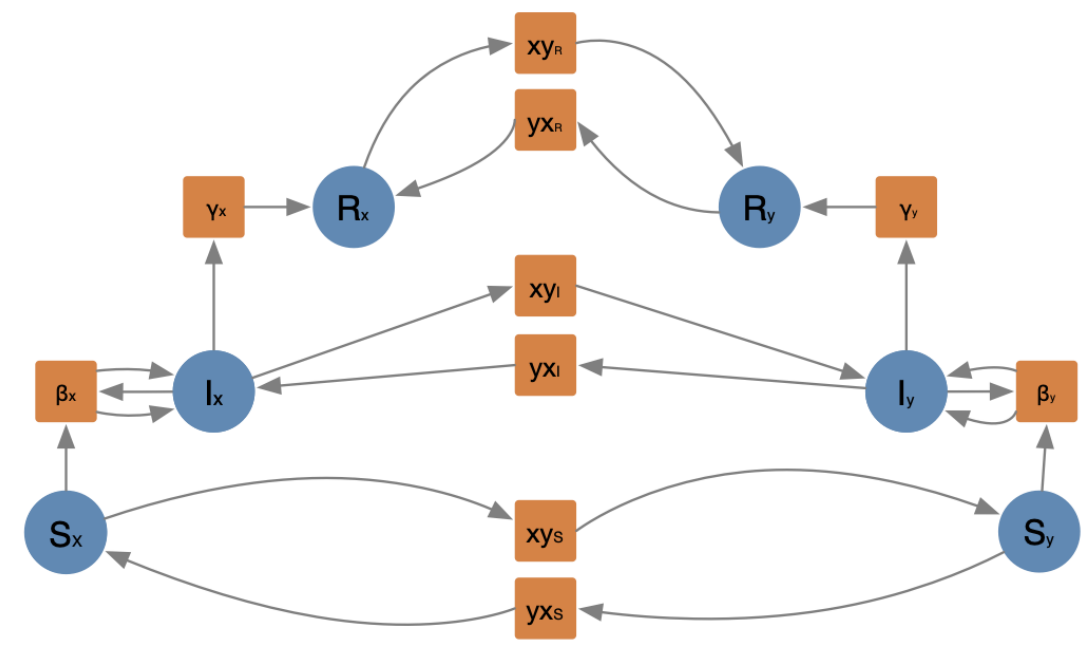
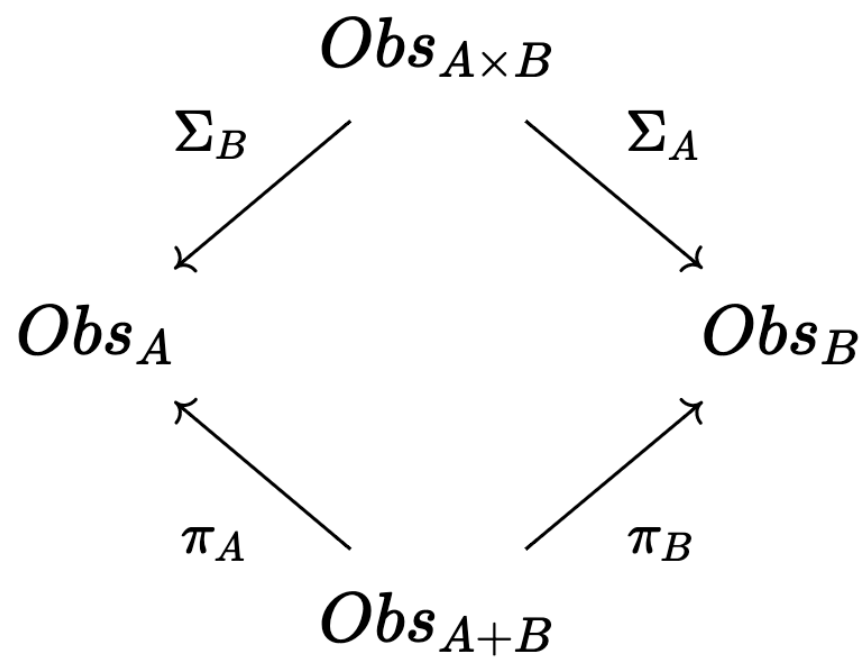
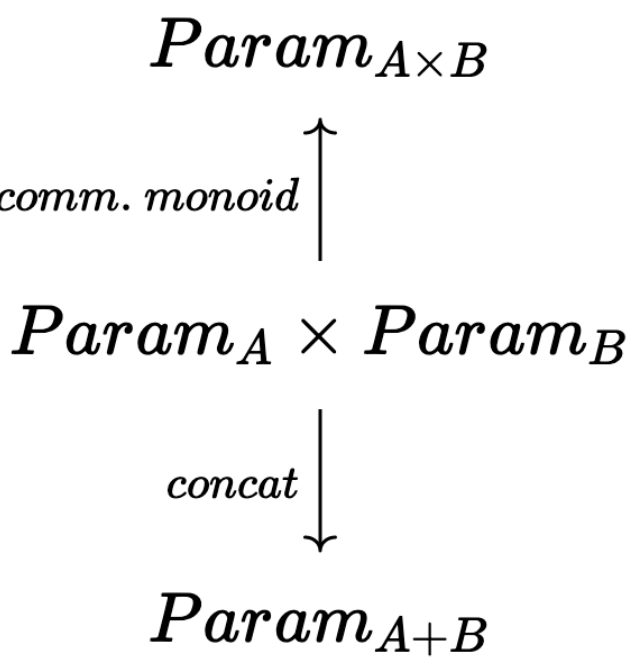
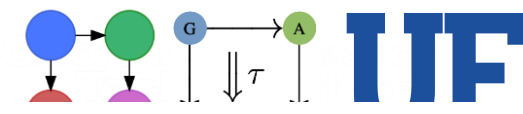
One particular model selection strategy



$$Loss(y, \hat{y}, \vec{p}) = \sum_{t \in T} \sum_{s \in \{S, I, R\}} (y_s(t) - \hat{y}_s(\hat{p}, t))^2$$



Model selection results



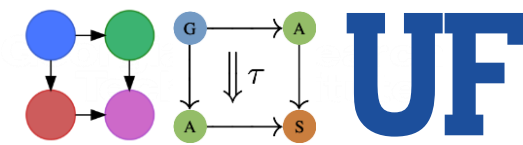
S: 90%, I: 10% and City X: 33%, City Y: 67%

For (SIRD)₂, we have S_X: 10%, S_Y: 25%

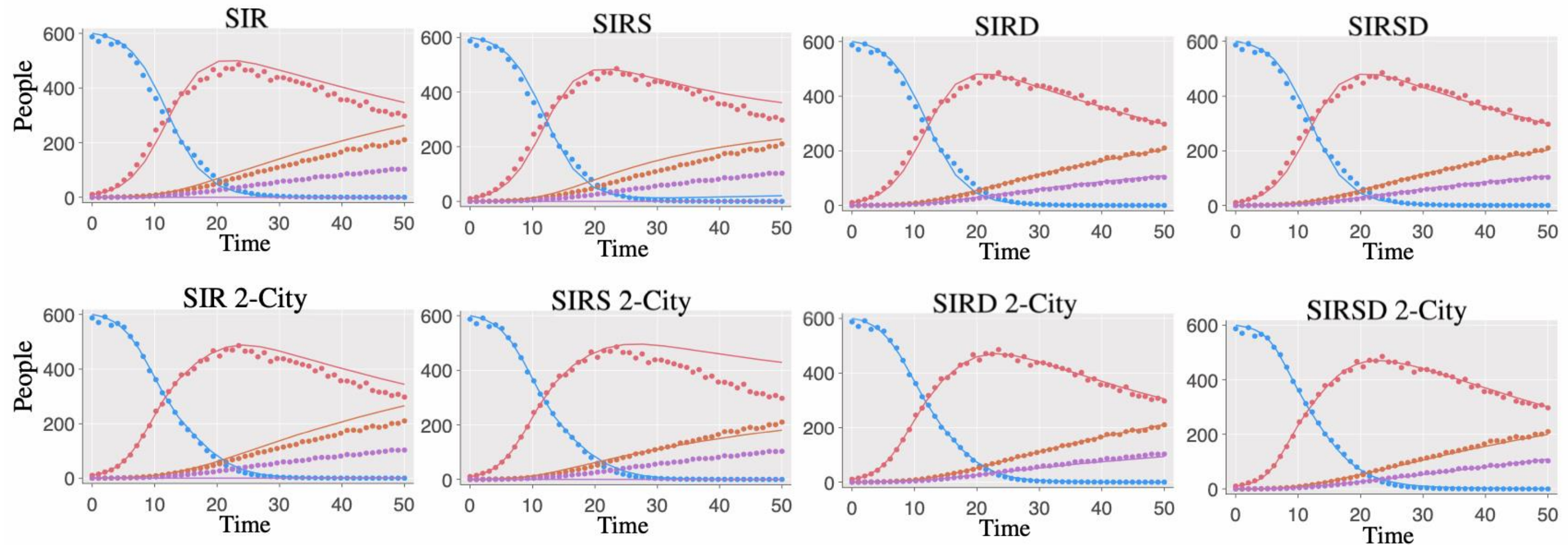
For (SIRD)₂, we have S_X: 30%, D_X: 0%

S: 35%

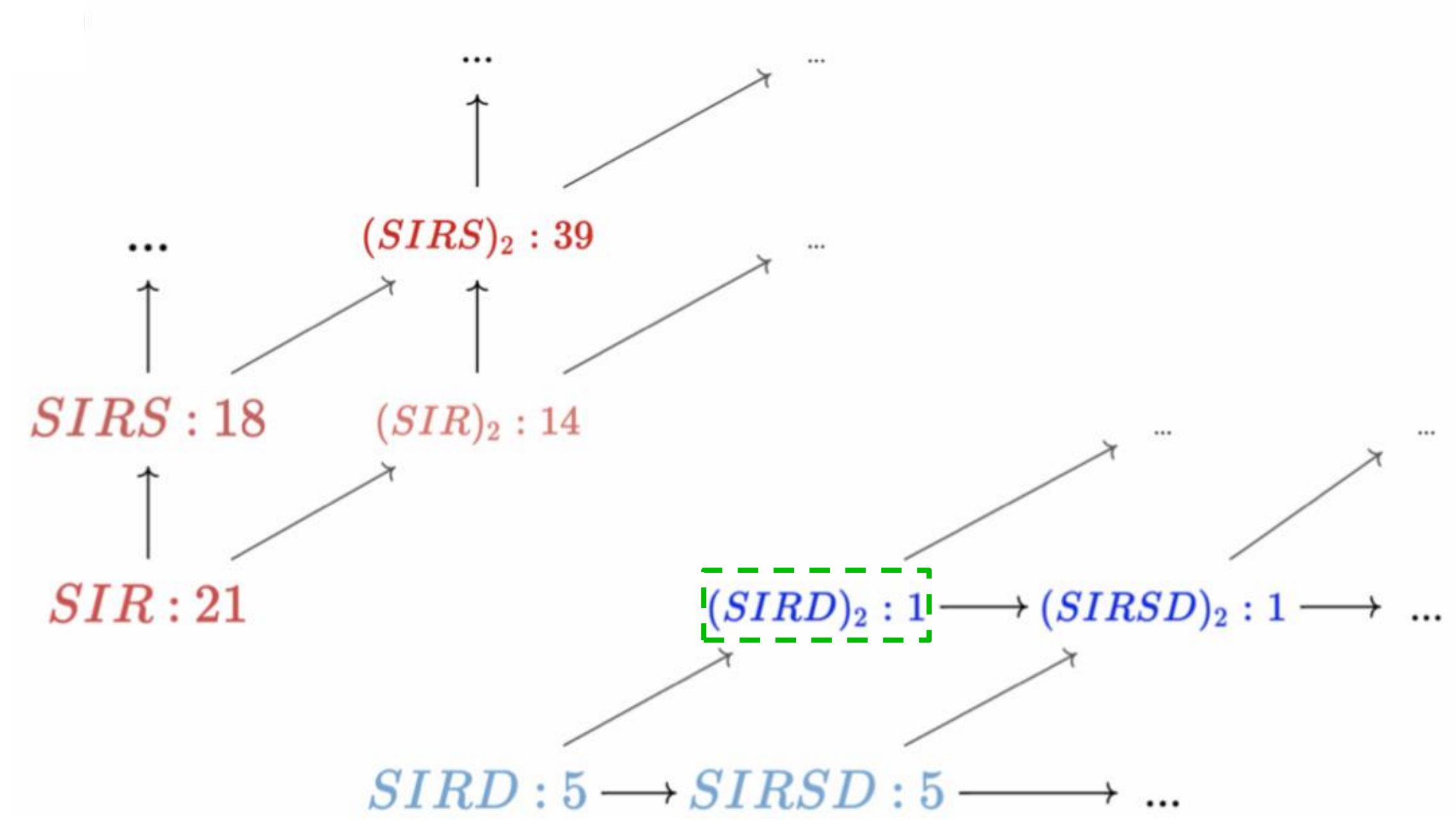
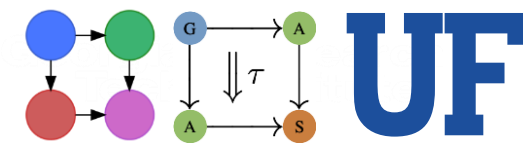
Model selection results



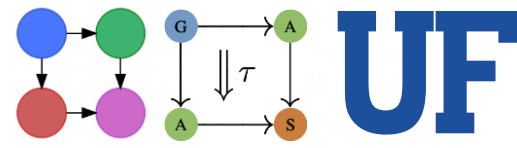
— Model ●●●● Data Σ Susceptible Σ Infected Σ Recovered Σ Dead



Model selection results

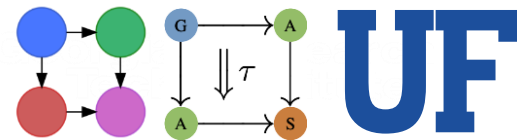


Future work



- More interesting “primitive” model space constructors
- Wiring diagram visualizer / GUI
- Lazy state space exploration
- Alternative applications as demonstrations (Boolean functions, circuits, NN)
- Hierarchical loss functions (optimizing overall goal + subgoals, together)

Thanks!



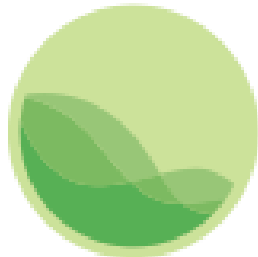
T. Hanks



Sean Wu



James Fairbanks



IHME
Institute for Health Metrics
and Evaluation



Andrew Baas

