

Statistical Analysis of Networks

Introduction to Stochastic Actor-Based Models

Motivating Example

- ❖ Victim / Offender overlap: a curious phenomenon
 - ❖ Why do victimization and offending tend to co-occur?

Motivating Example



CRIMINOLOGY

**VIOLENT OFFENDING AND VICTIMIZATION IN
ADOLESCENCE: SOCIAL NETWORK MECHANISMS
AND HOMOPHILY***

JILLIAN J. TURANOVIC¹ and JACOB T.N. YOUNG²

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²School of Criminology and Criminal Justice, Arizona State University

- ❖ As a network question:
 - ❖ Why is there homophily for victimization? Offending?
 - ❖ *What do the authors say?*

Motivating Example

A screenshot of a journal article title page. The word "CRIMINOLOGY" is centered at the top in a large, blue, serif font, underlined. Below it, the title "VIOLENT OFFENDING AND VICTIMIZATION IN ADOLESCENCE: SOCIAL NETWORK MECHANISMS AND HOMOPHILY*" is written in a smaller, black, serif font. The authors' names, "JILLIAN J. TURANOVIC¹ and JACOB T.N. YOUNG²", are listed below the title. At the bottom, two footnotes provide affiliations: "¹College of Criminology and Criminal Justice, Florida State University" and "²School of Criminology and Criminal Justice, Arizona State University". A small icon with a red bookmark and a 'C' is visible in the top right corner of the screenshot.

CRIMINOLOGY

**VIOLENT OFFENDING AND VICTIMIZATION IN
ADOLESCENCE: SOCIAL NETWORK MECHANISMS
AND HOMOPHILY***

JILLIAN J. TURANOVIC¹ and JACOB T.N. YOUNG²

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- ❖ Interpersonal relational decision making shapes these outcomes.
- ❖ But, this can occur through different mechanisms.

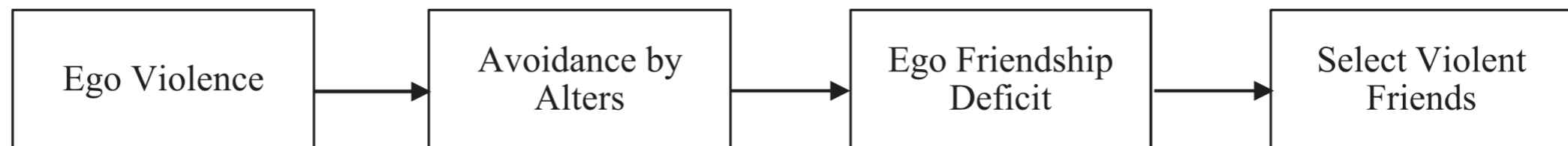
Motivating Example

Figure 1. Mechanisms of Violent Offending Homophily

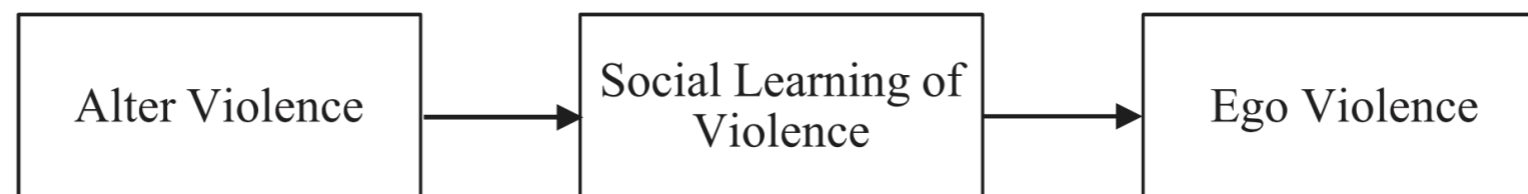
a Violence homophily through preference for similarity:



b Violence homophily through avoidance:



c Violence homophily through influence:

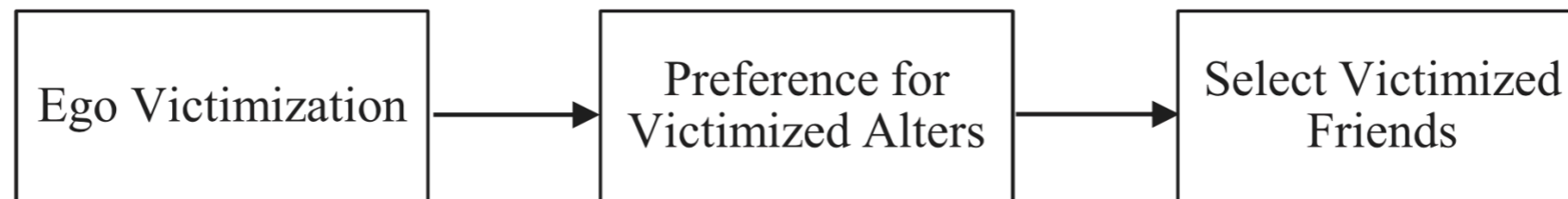


NOTE: "Ego" refers to the focal individual; "alter" refers to potential friends.

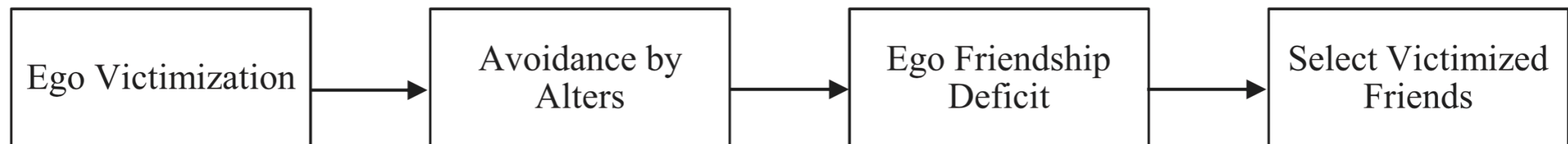
Motivating Example

Figure 2. Mechanisms of Violent Victimization Homophily

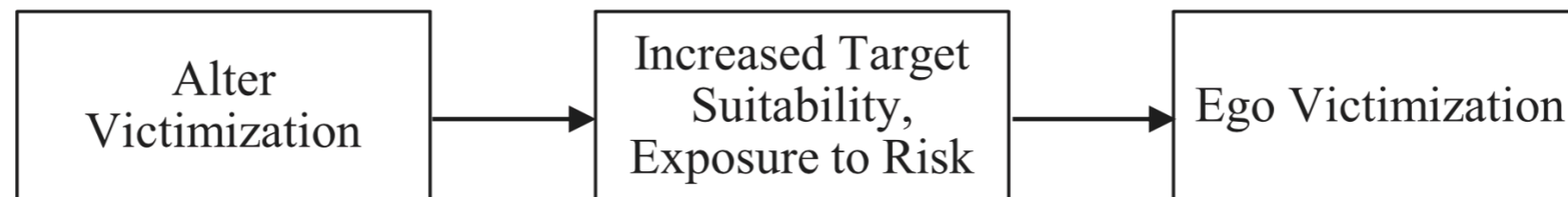
a Victimization homophily through preference for similarity:



b Victimization homophily through avoidance:



c Victimization homophily through influence:



NOTE: "Ego" refers to the focal individual; "alter" refers to potential friends.

Motivating Example

- ❖ What are the findings?

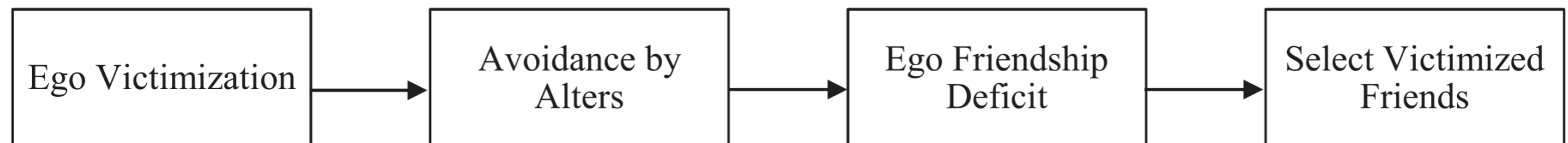
Motivating Example

❖ What are the findings?

a Violence homophily through preference for similarity:



b Victimization homophily through avoidance:



Motivating Example

- ❖ How did the authors go about coming to these conclusions?
- ❖ Stochastic Actor-Based Models!

Statistical Analysis of Networks

Introduction to Stochastic Actor-Based Models

Learning Goals

- ❖ By the end of this lecture, you should be able to answer these questions:
 - ❖ What is the difference between Exponential Random Graph Models (ERGMs) and Stochastic Actor-Based Models (SABMs)?
 - ❖ What is the logic of *micro-steps* and the simulation of networks using the **rate** function?
 - ❖ What is the logic of *preferences* and the simulation of networks using the **objective** function?

Introduction

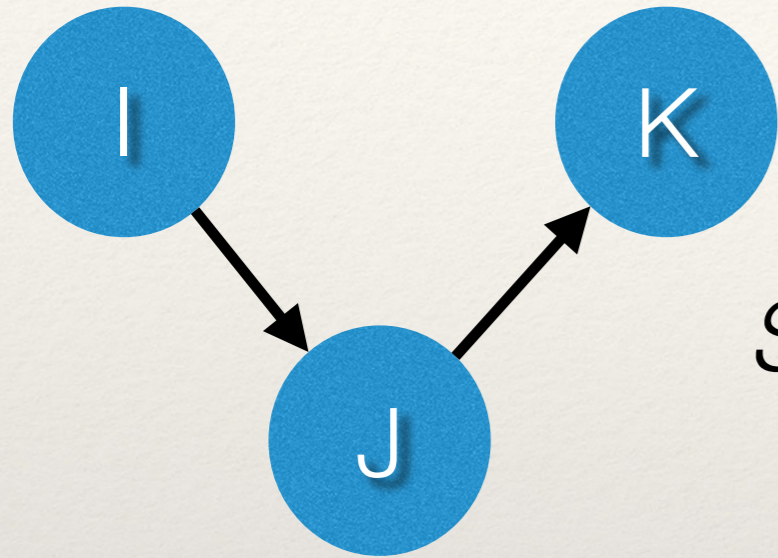
- ❖ With ERGMs, we asked: “what are the network configurations that generated this network?”
- ❖ We reviewed models that are “edge-based” in the sense that the probability of an edge is dependent (or not) on other edges in the network.
 - ❖ Remember *dyadic-dependence*?

Introduction

- ❖ ERGMs are for a single cross-section of a network. ***
- ❖ But, what if we have a **panel** of networks?
 - ❖ We want to model “network dynamics”

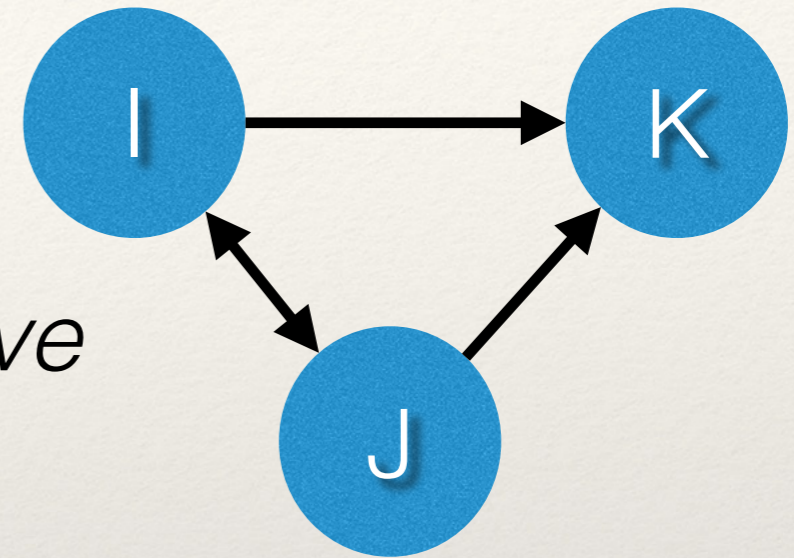
What do we mean by
“network dynamics”?

Network Dynamics



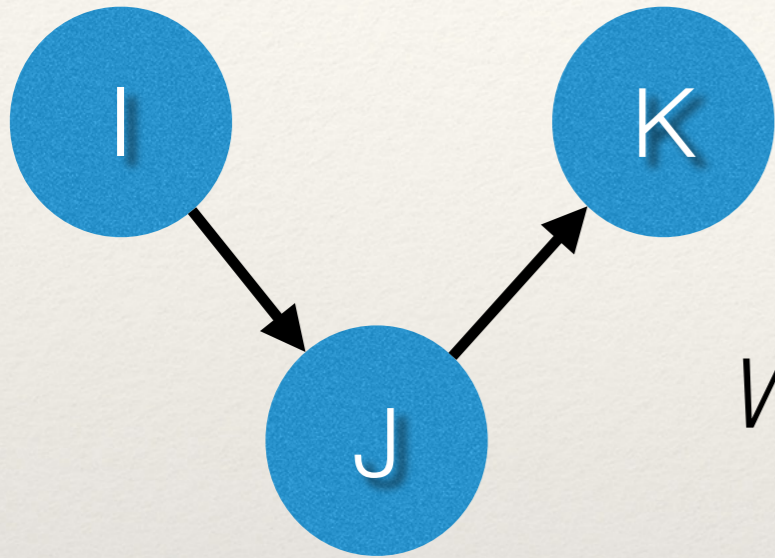
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*Suppose we observe
two discrete time
points.*



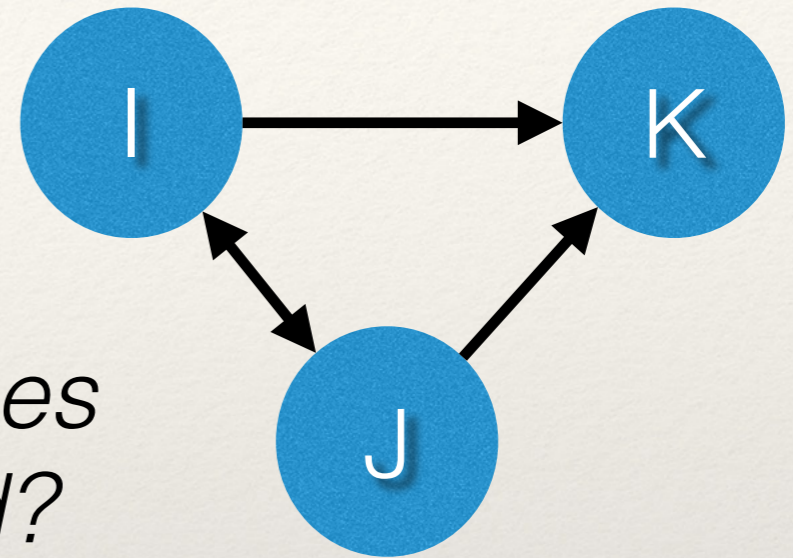
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Network Dynamics



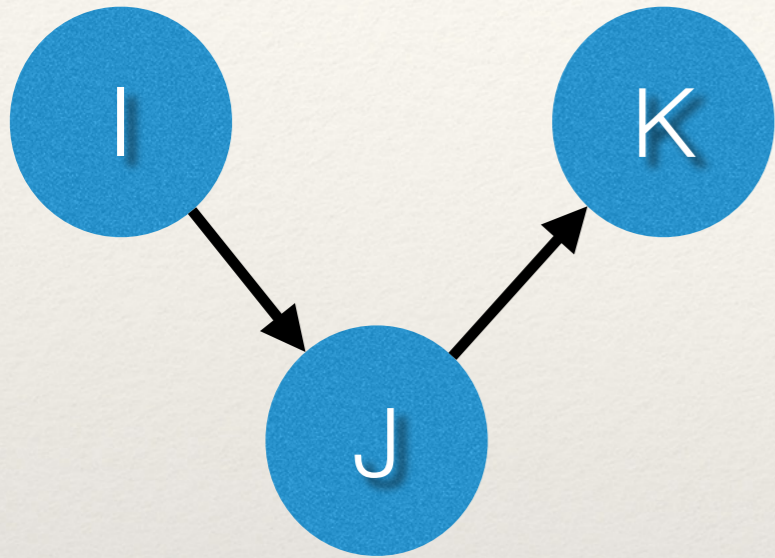
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*What are the changes
that have occurred?*

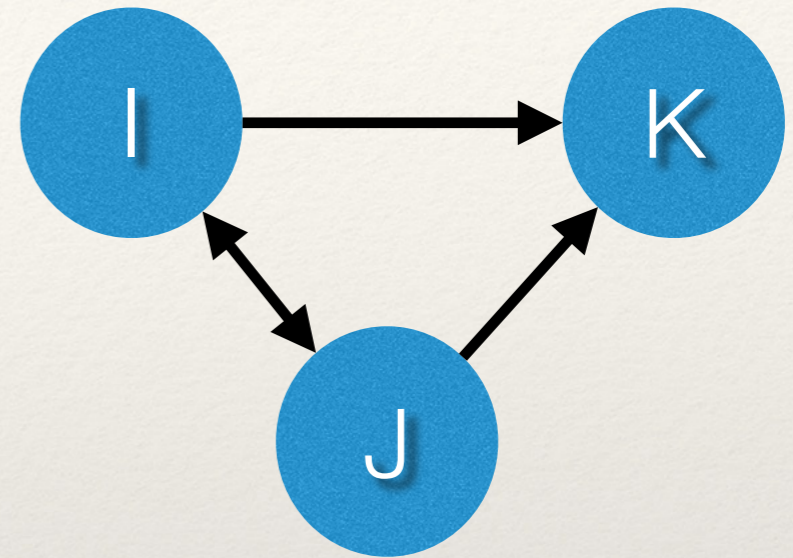


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Network Dynamics

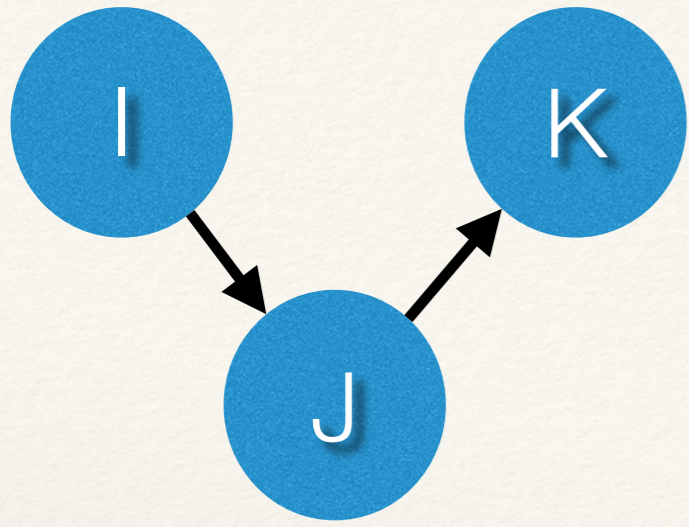


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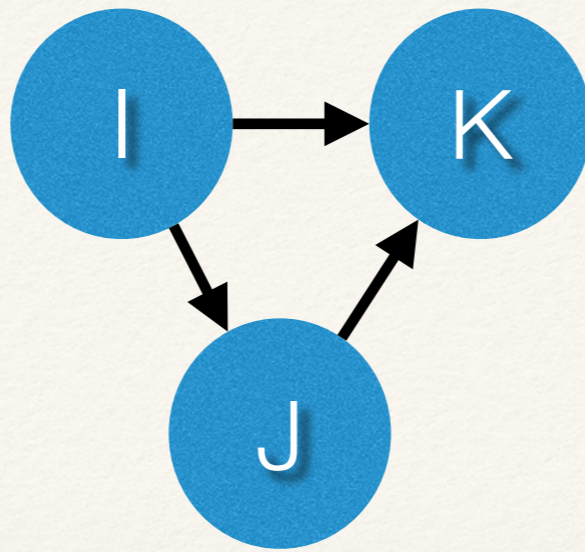
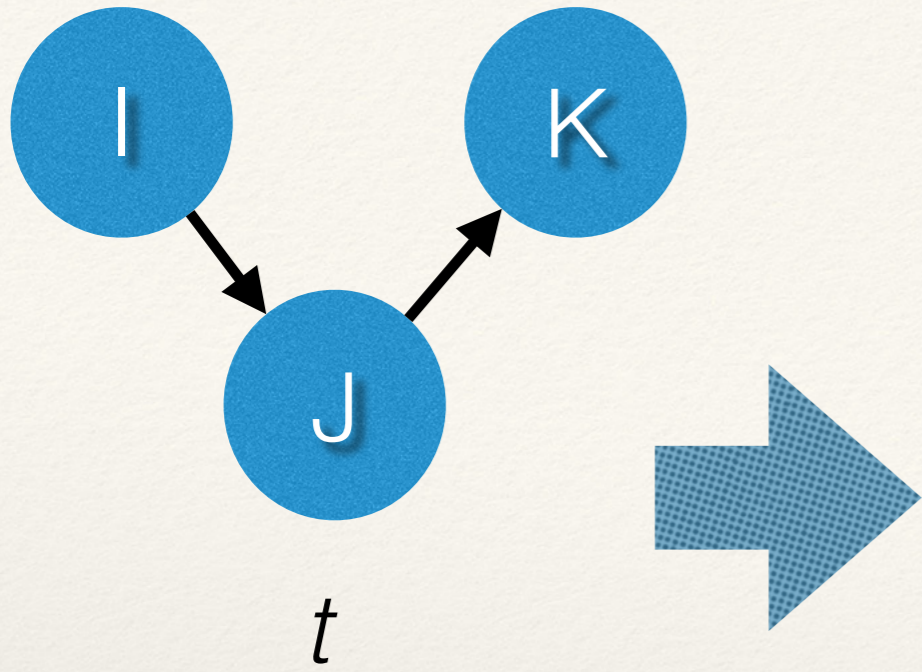


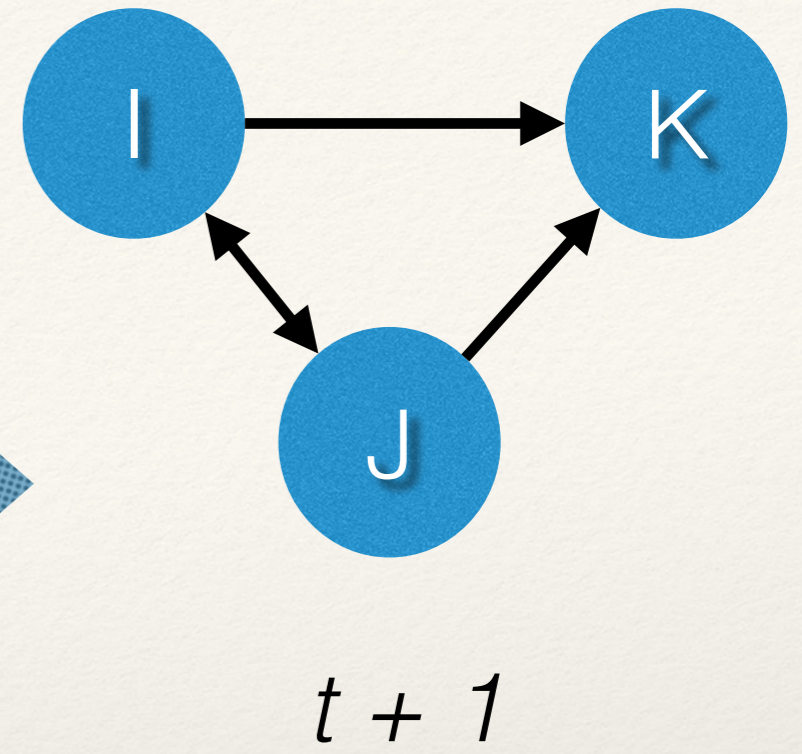
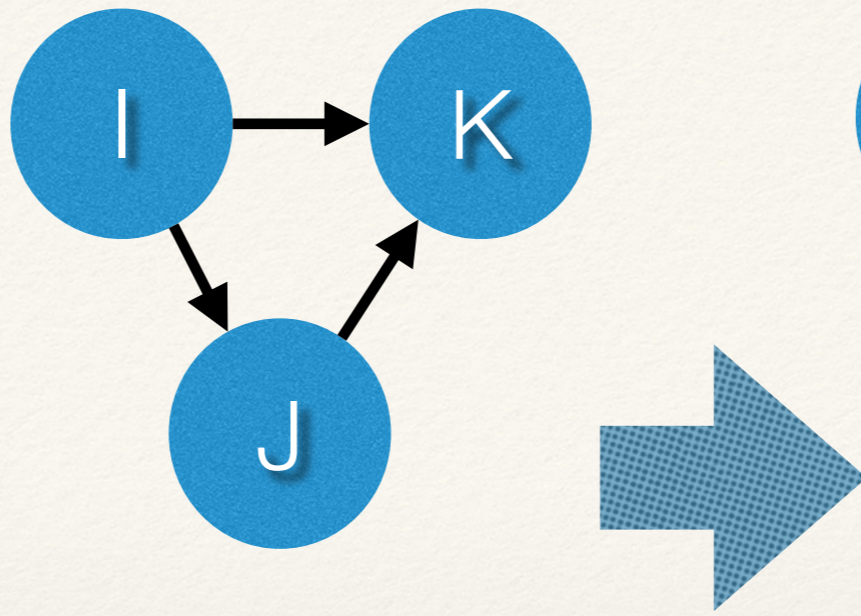
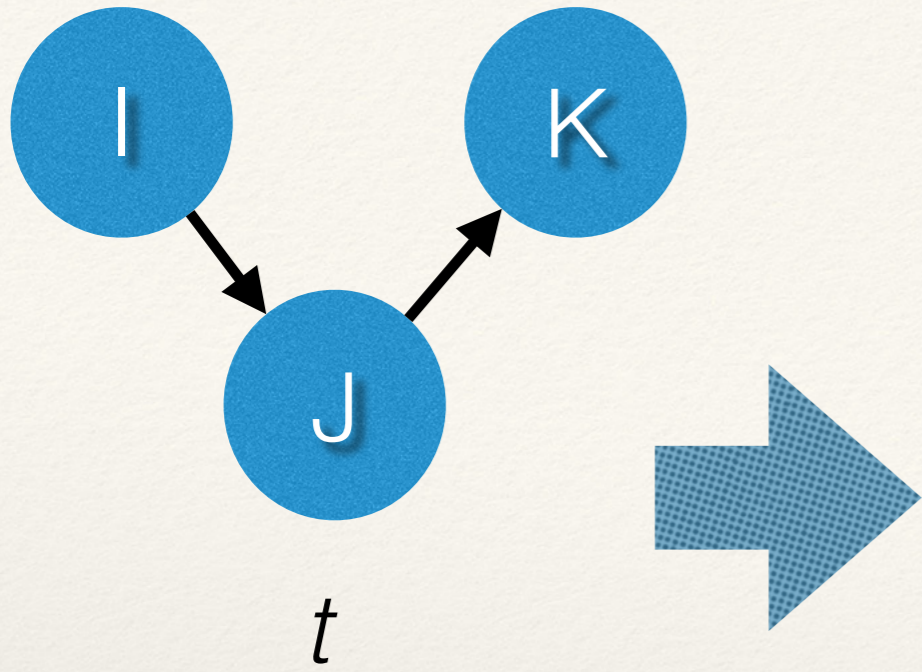
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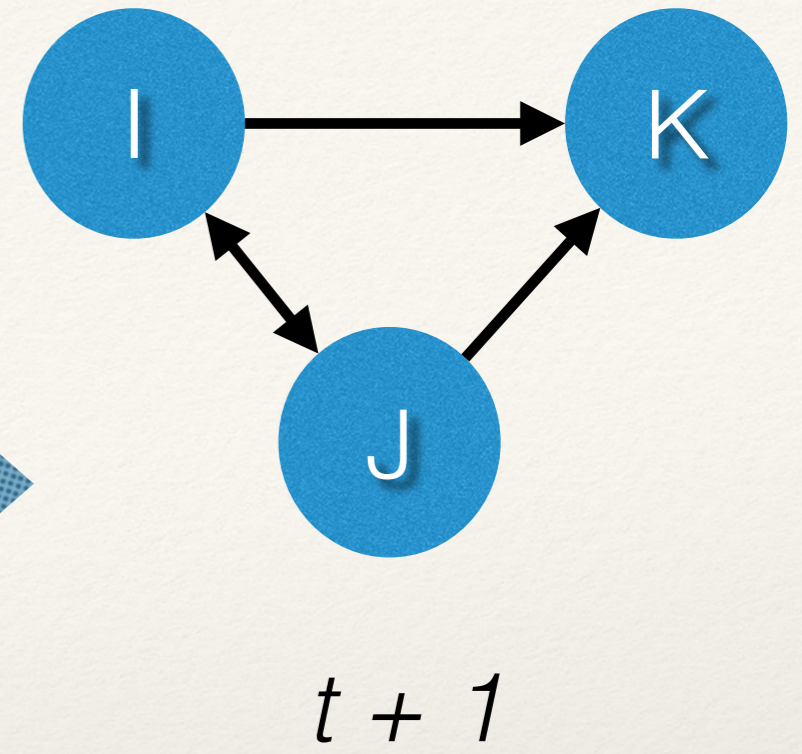
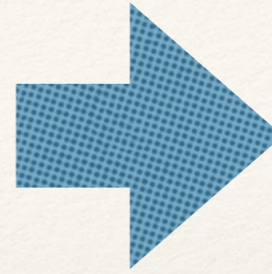
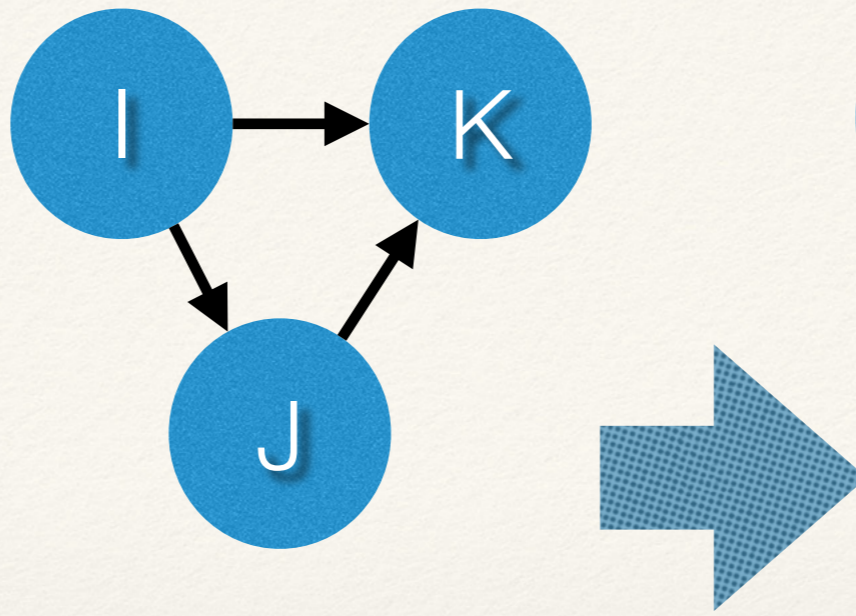
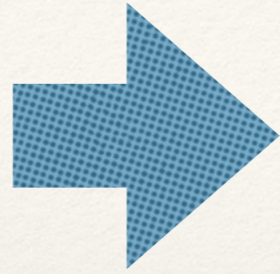
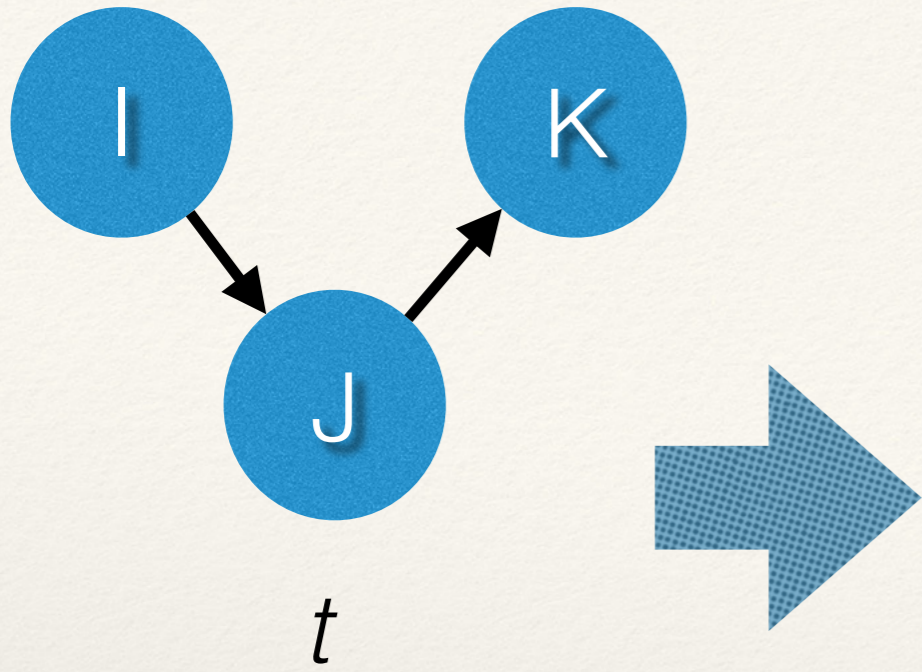
*How did those
changes happen?*



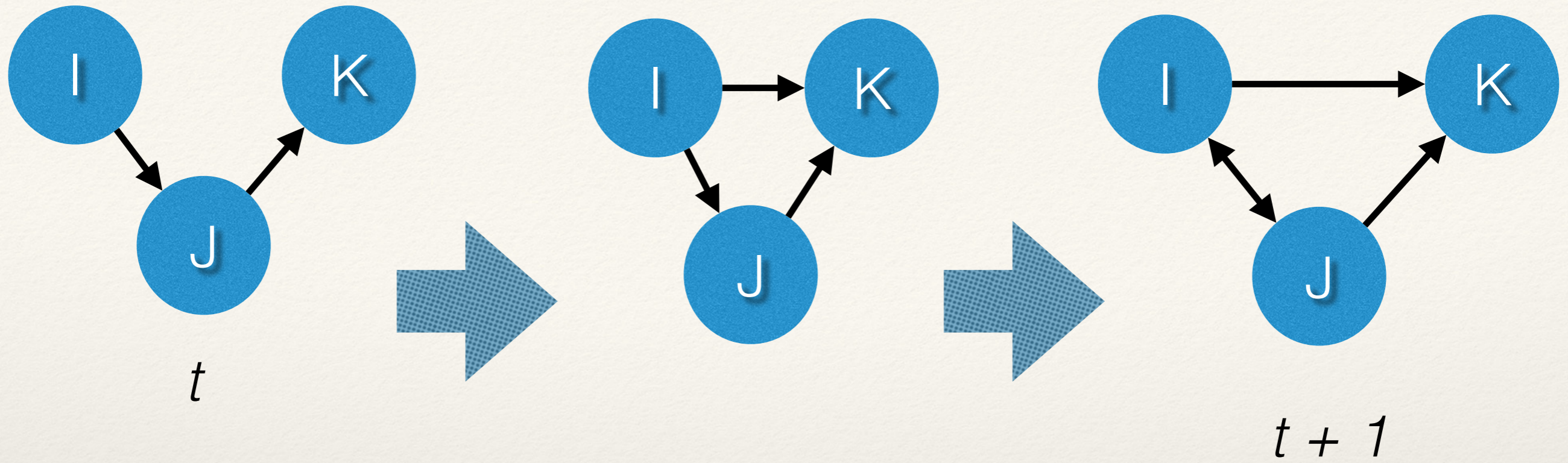
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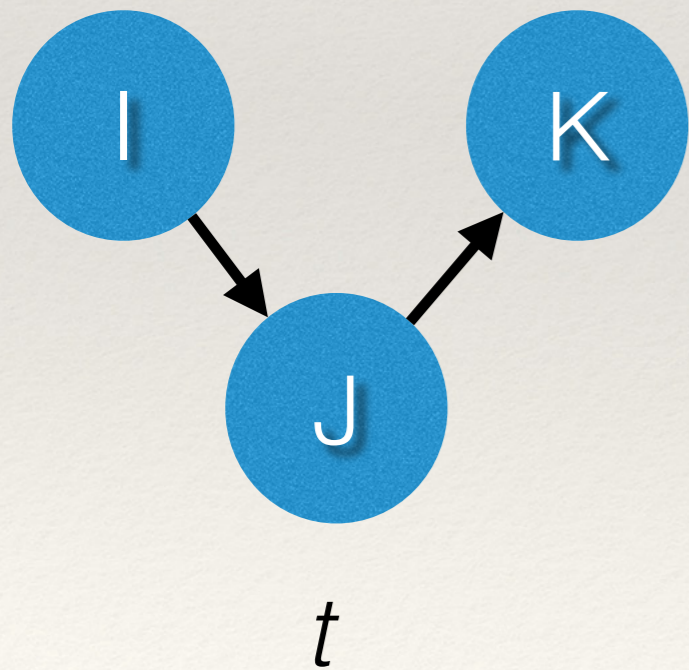


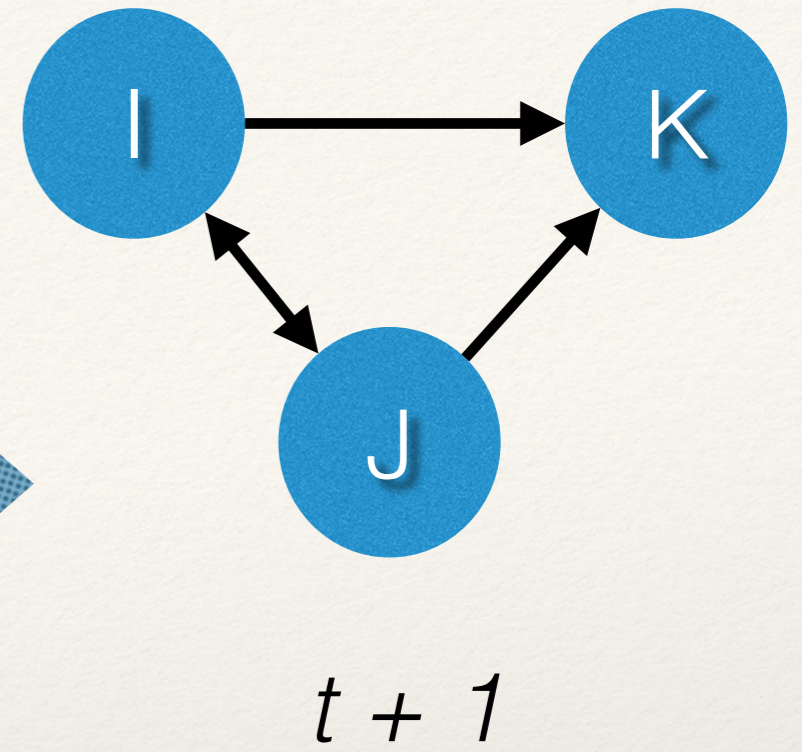
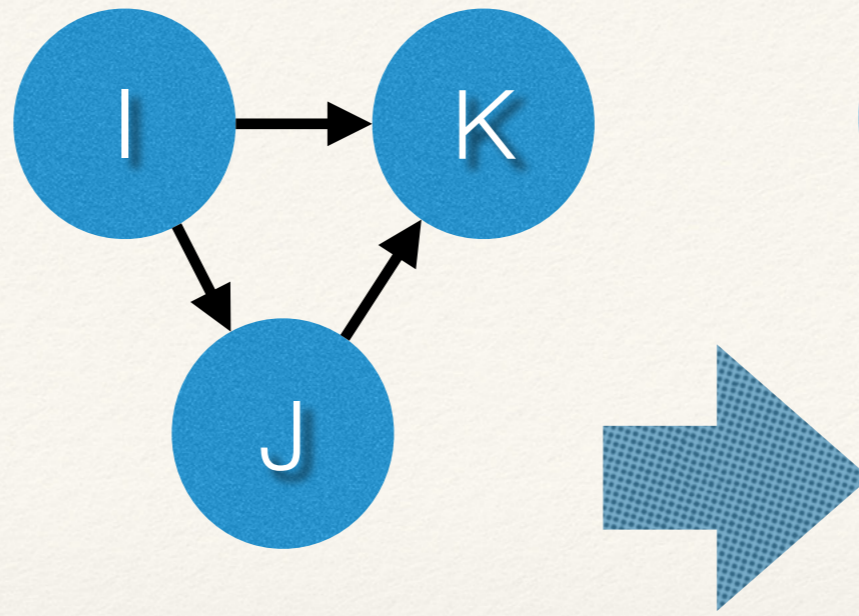
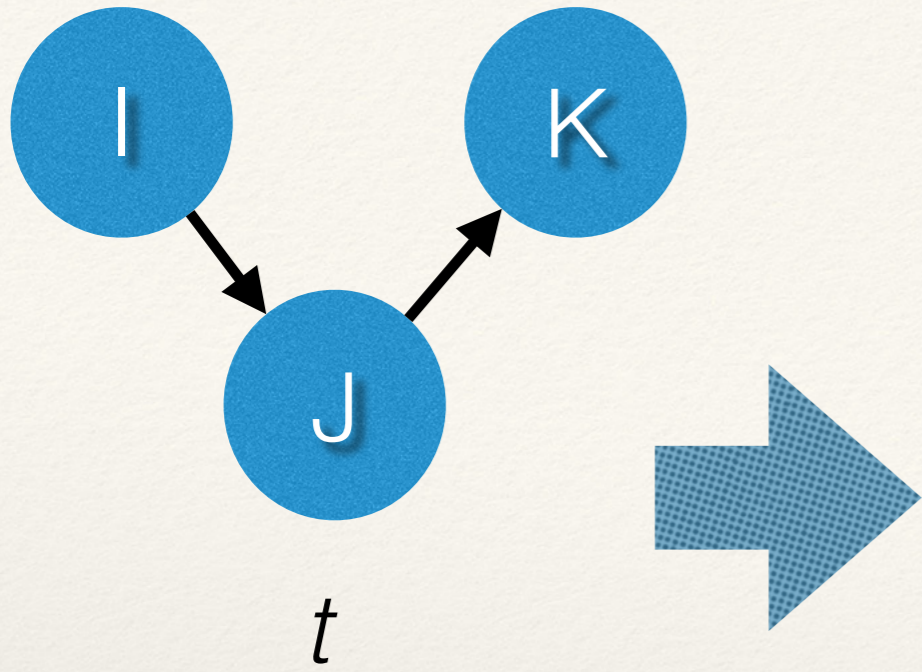


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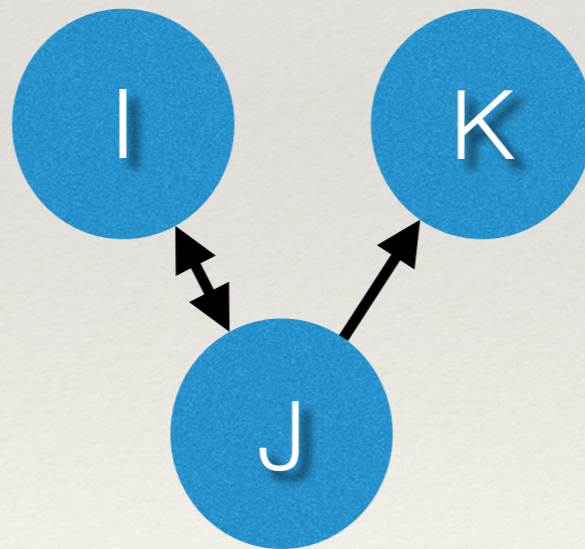
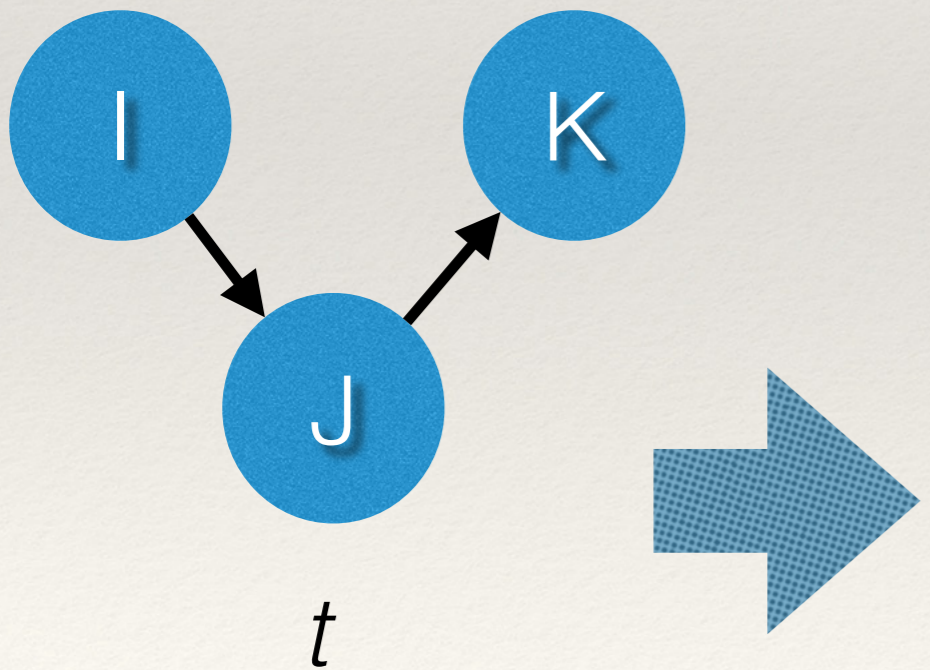


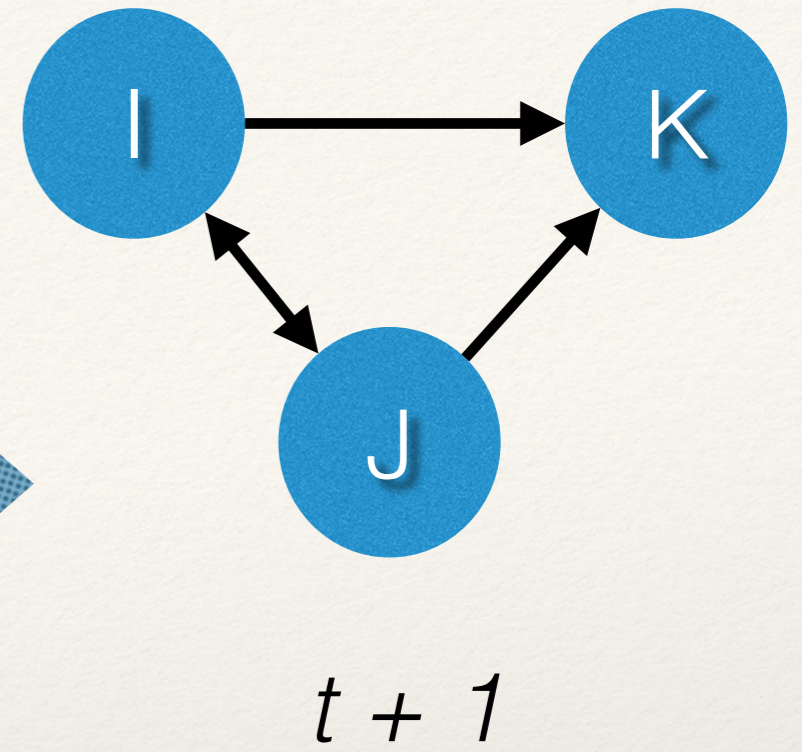
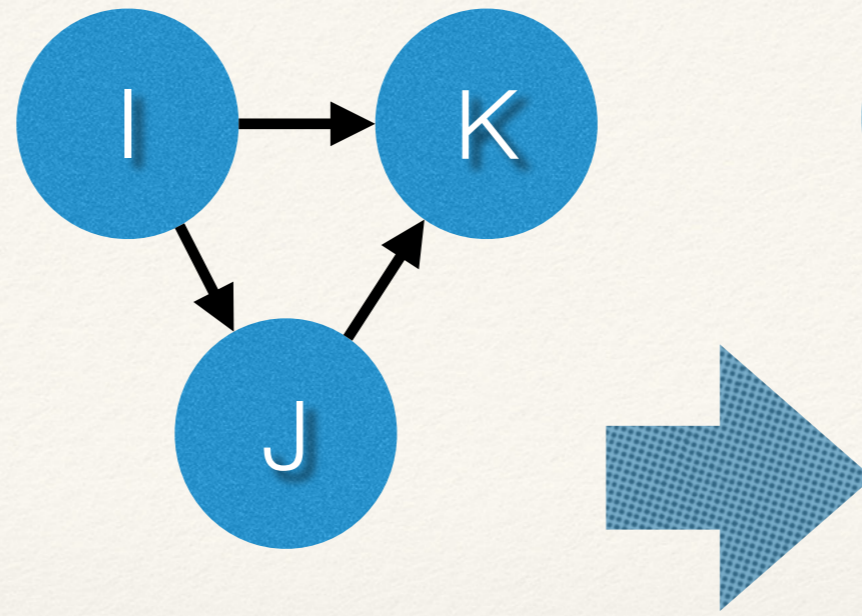
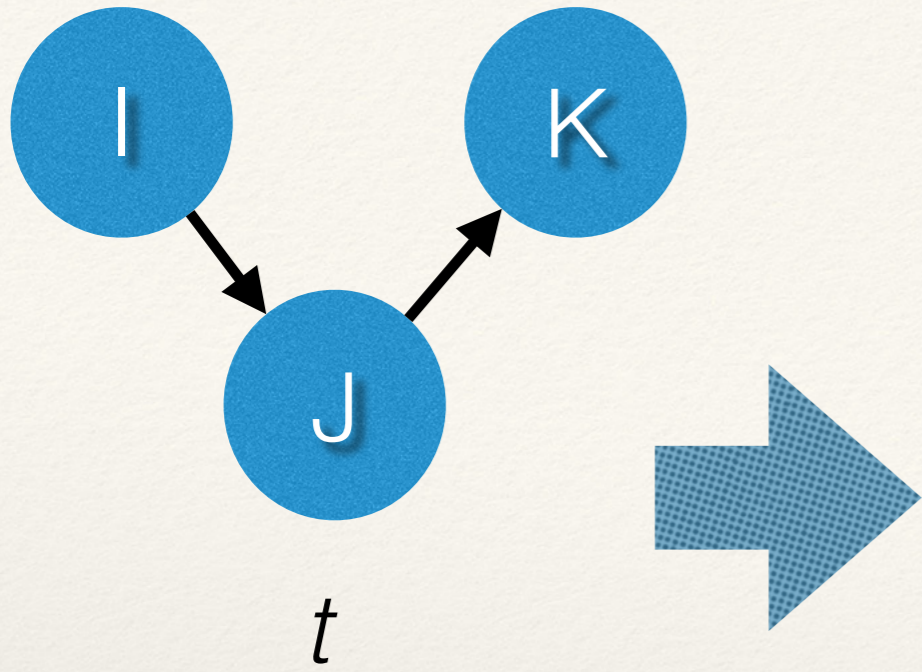
Or...



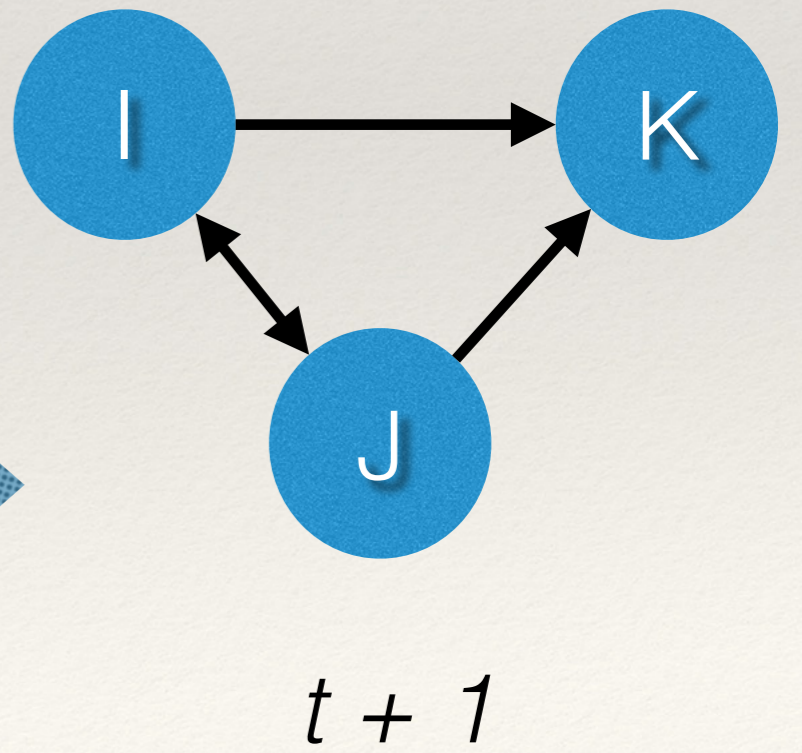
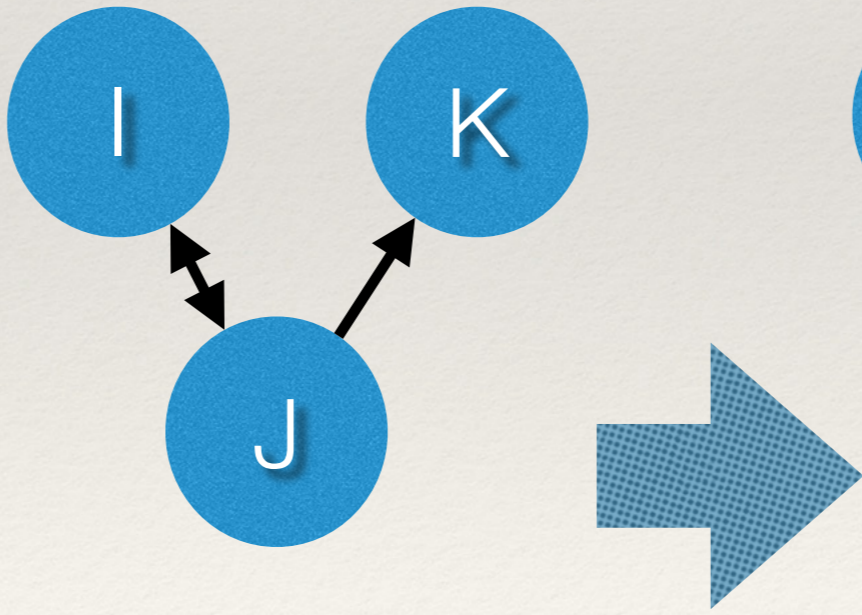
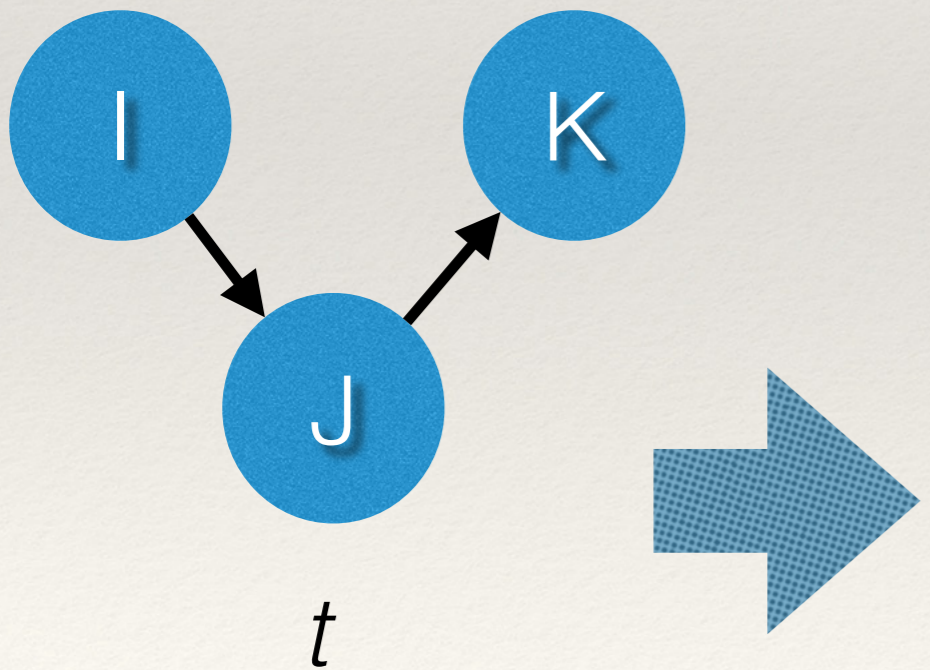


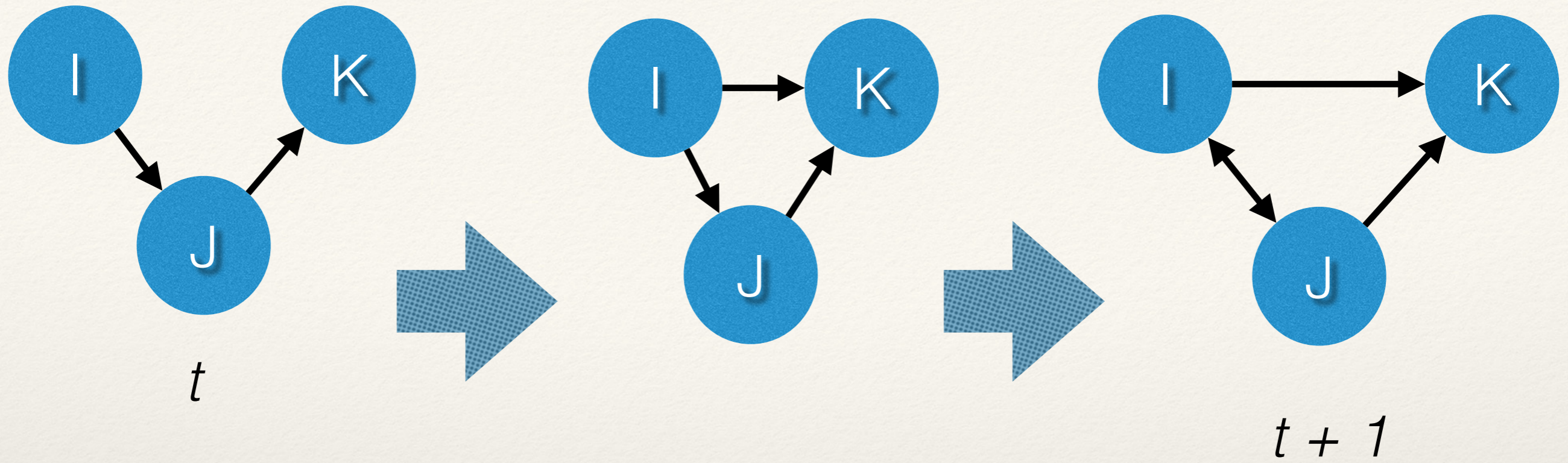
Or...



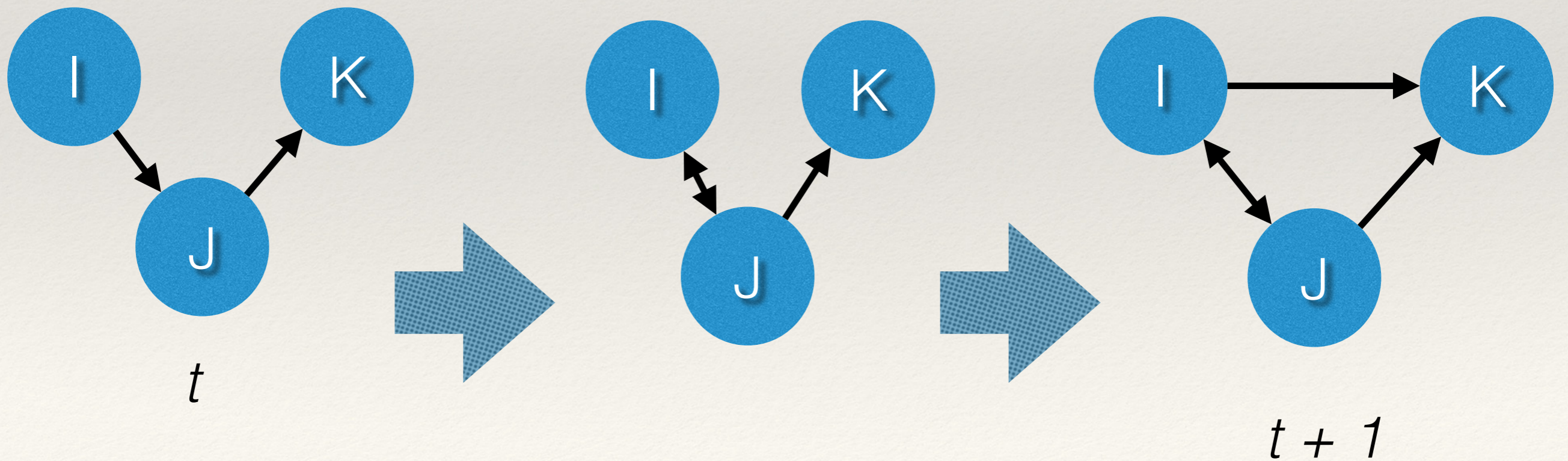


Or...

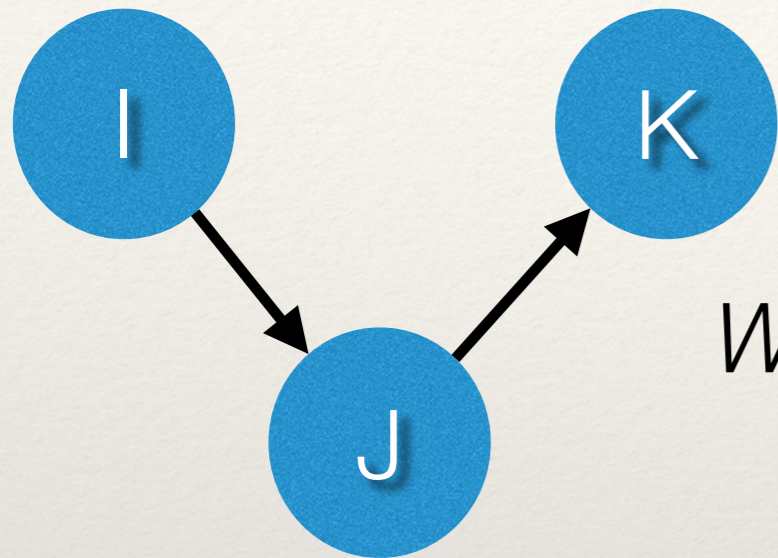




Both are different sequences.

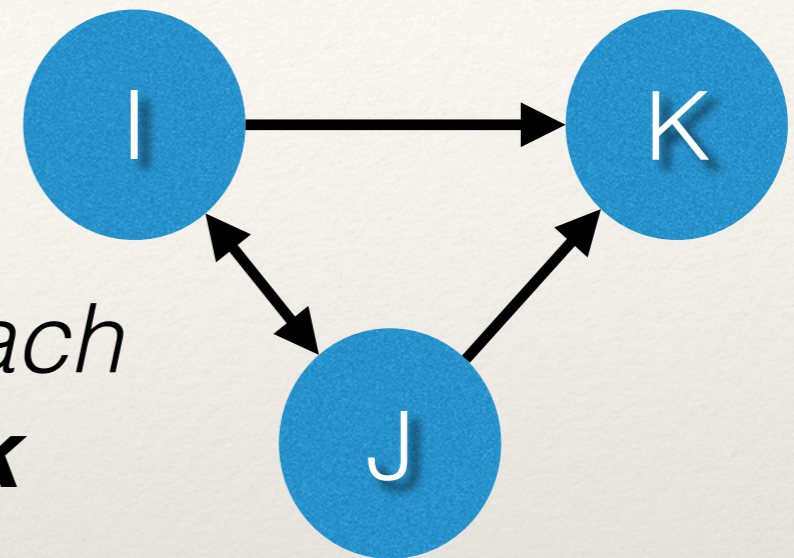


Network Dynamics



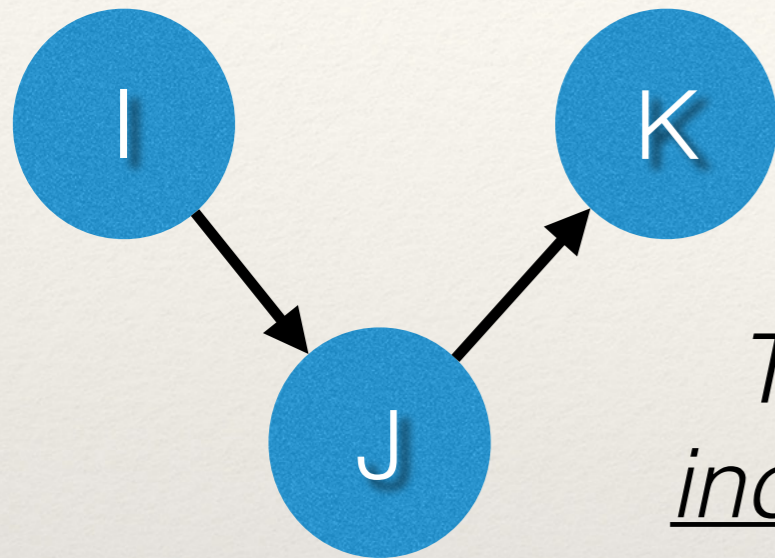
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*We don't observe each
step with **network
panel data***



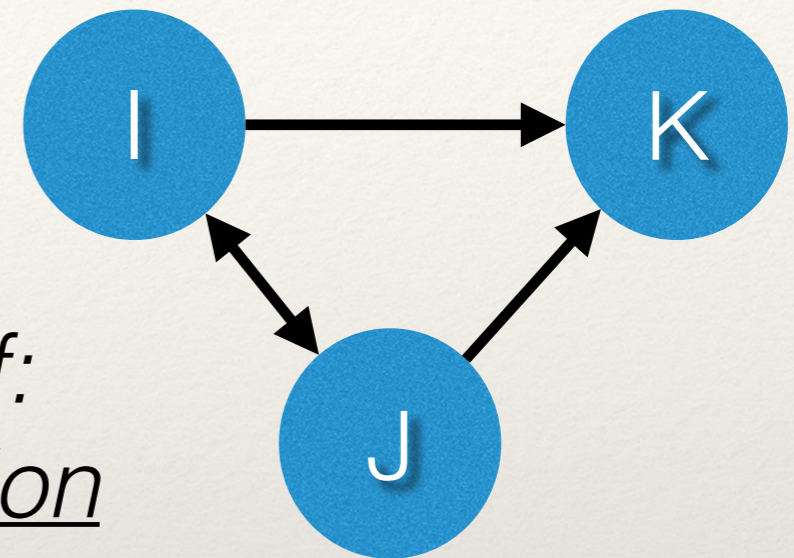
$t + 1$

Network Dynamics



t

*This is a problem of:
incomplete observation
of change*

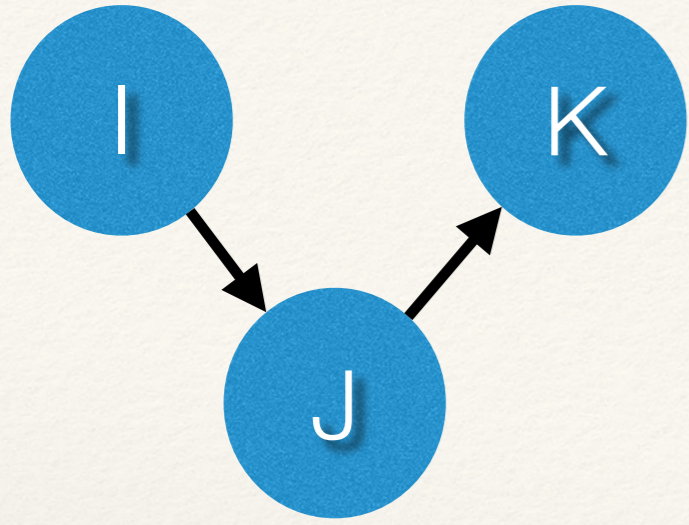


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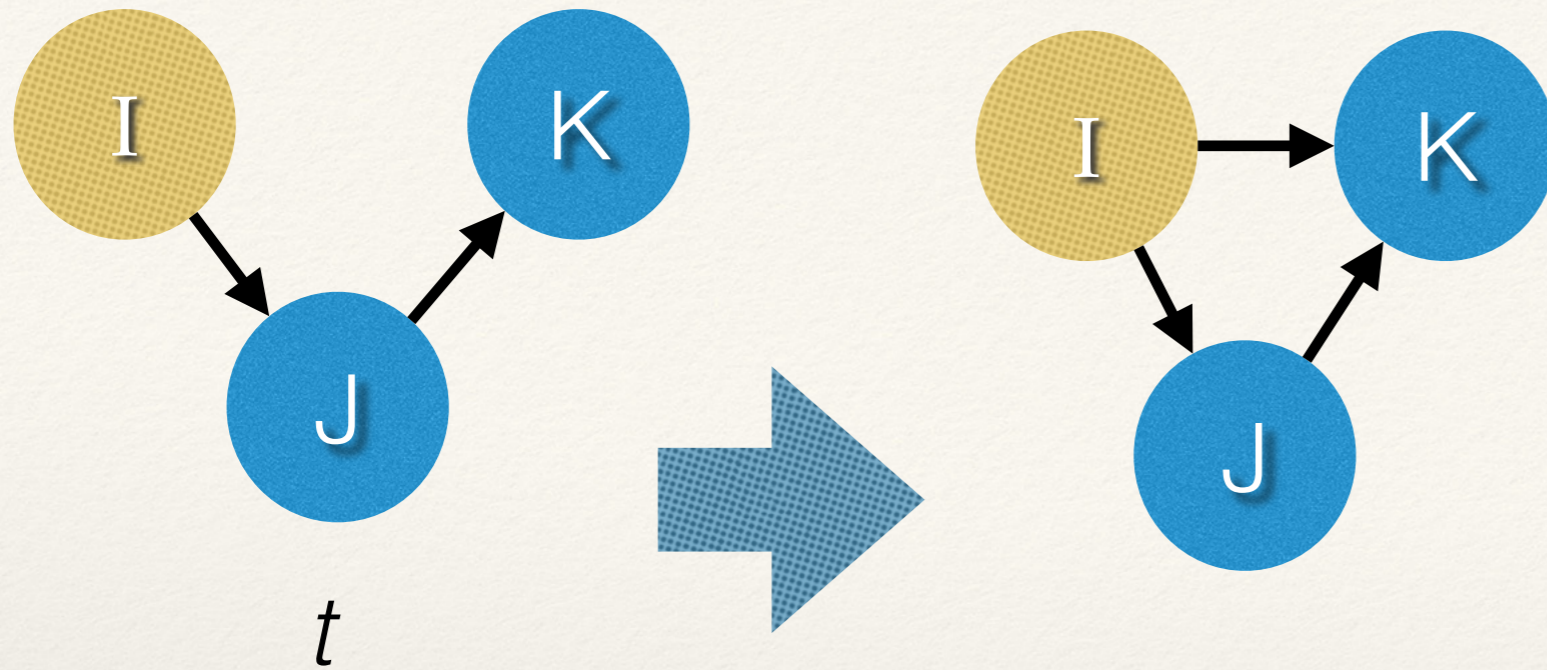
*We want to model
these dynamics as
micro steps*

What is a “micro-step”?

- ❖ Uniquely identify actors
 - ❖ Actors control and decide about the tie variable
 - ❖ (*Note the difference from an ERGM*):
 - ❖ “actor-based” vs. “edge-based”

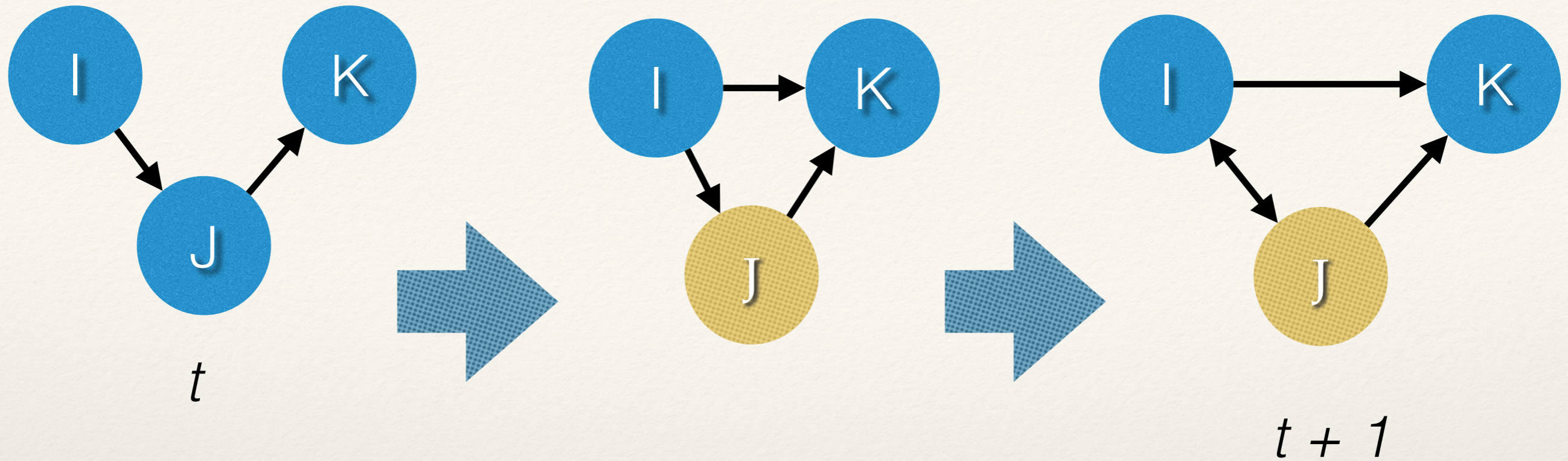


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*In the first micro-step,
the yellow node makes
a decision.*

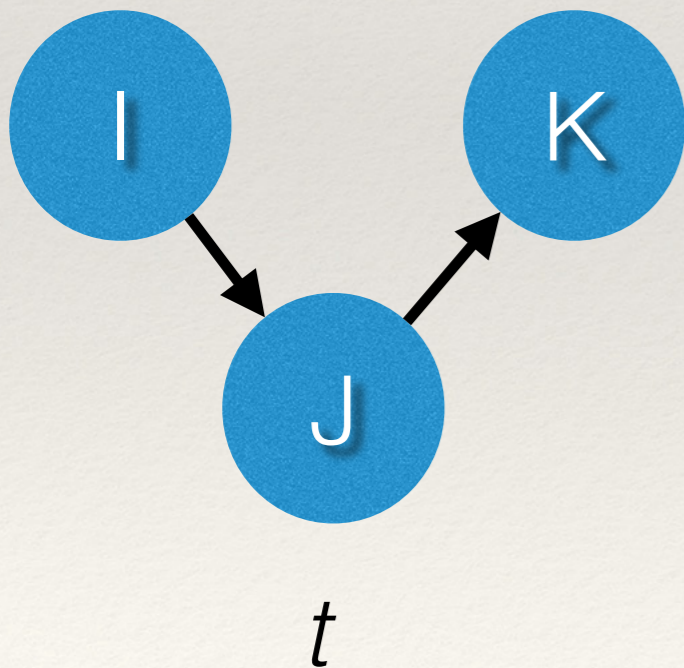
Specifically, add a tie.



In the second micro-step, the yellow node makes a decision.

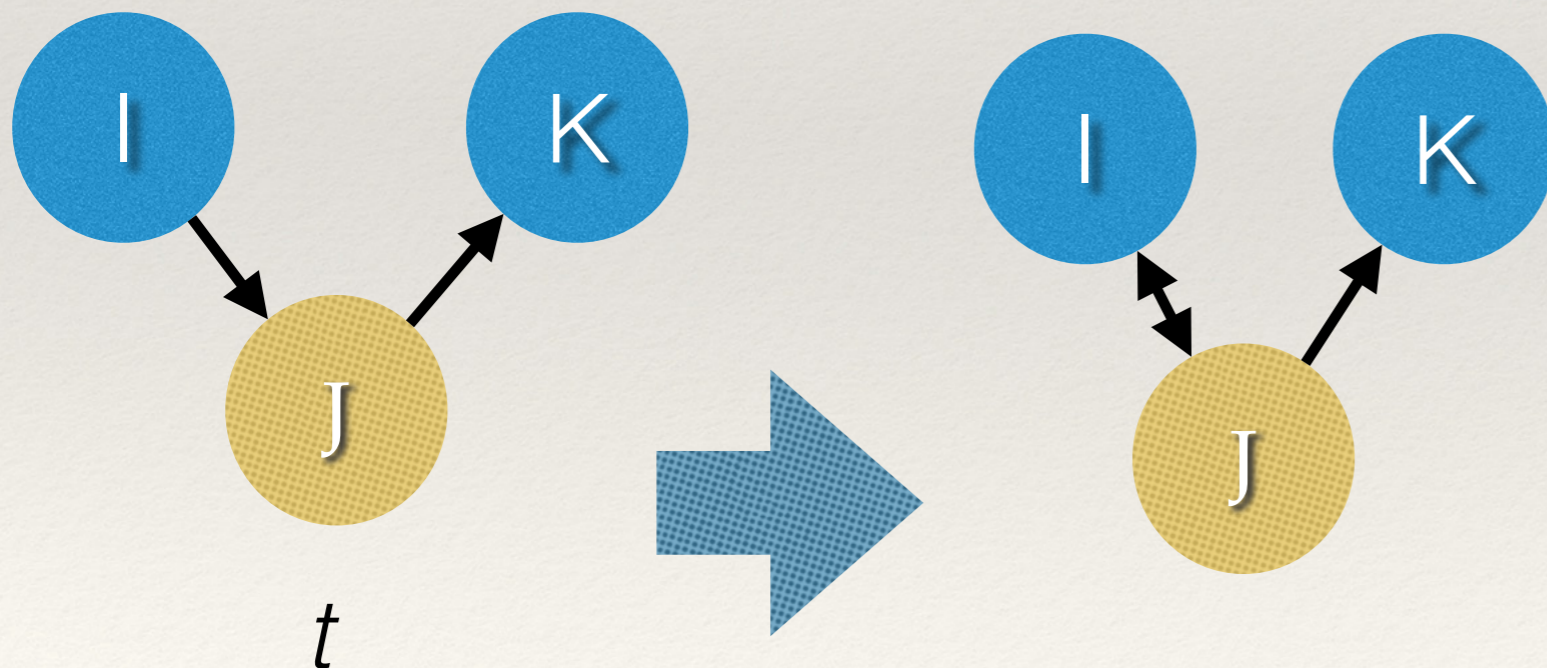
Specifically, reciprocate a tie.

*But, there are different
sequences of
decisions that occur by
which we would reach
the second network.*



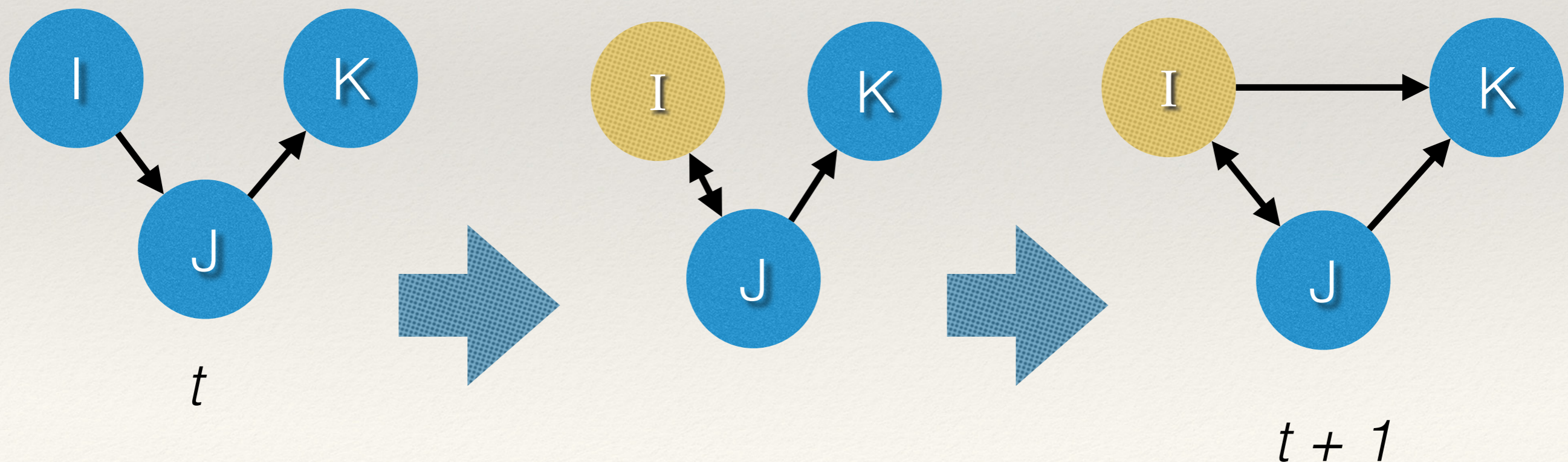
*In the first micro-step,
the yellow node makes
a decision.*

*Specifically,
reciprocate a tie.*



In the second micro-step, the yellow node makes a decision.

Specifically, add a tie.



Why 'actor-based'?

- ❖ Actors make decisions which drive change in the network (i.e. actor-driven or agent-based model).
- ❖ ERGMs are *edge*-based models in that we parameterize the configurations of edges that characterize the network.
 - ❖ With the SABM, we parameterize what the actor is doing with their ties.

Why 'actor-based'?

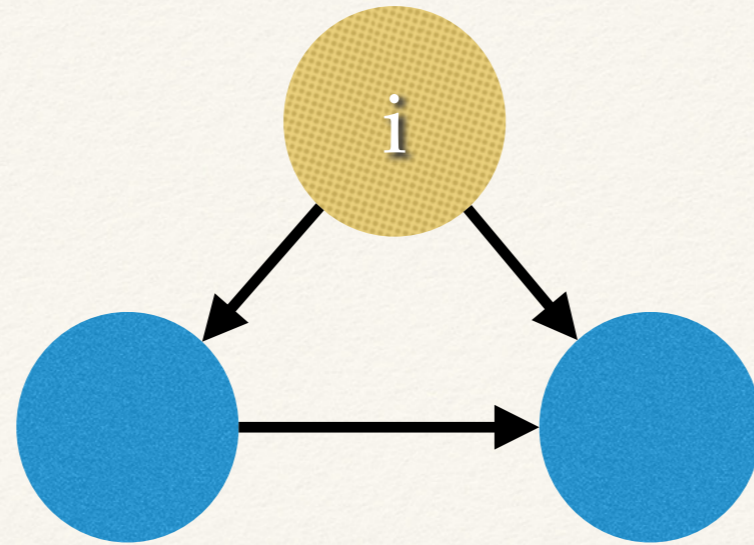
- ❖ The SABM has 2 sub-models (called **functions**):
 - ❖ When can actor i make a decision? (**rate**)
 - ❖ Which decision does actor i make? (**objective**)

Simulating Network Evolution

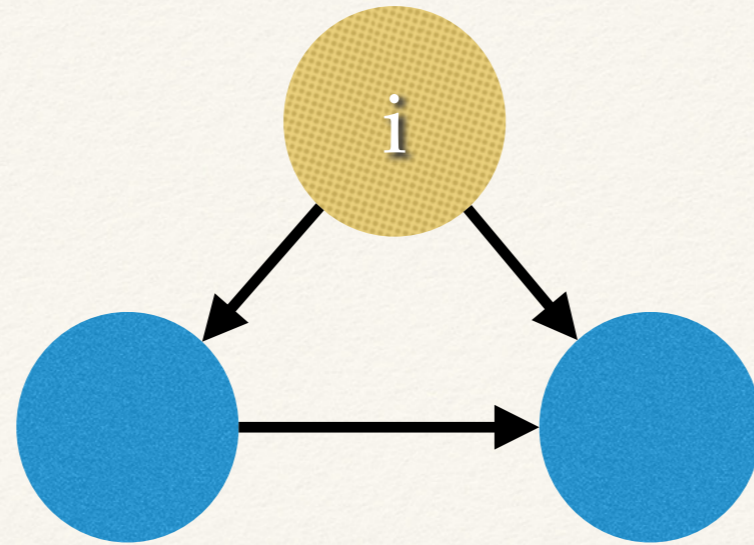
- ❖ The SABM logic goes like this:
 - ❖ Start with a network at t_0 and run an algorithm to t_1 .
 - ❖ For all actors, a *waiting time* is sampled according to the *rate function*.
 - ❖ Take the actor with the shortest waiting time and allow the actor to set a *micro step*.

Objective Function

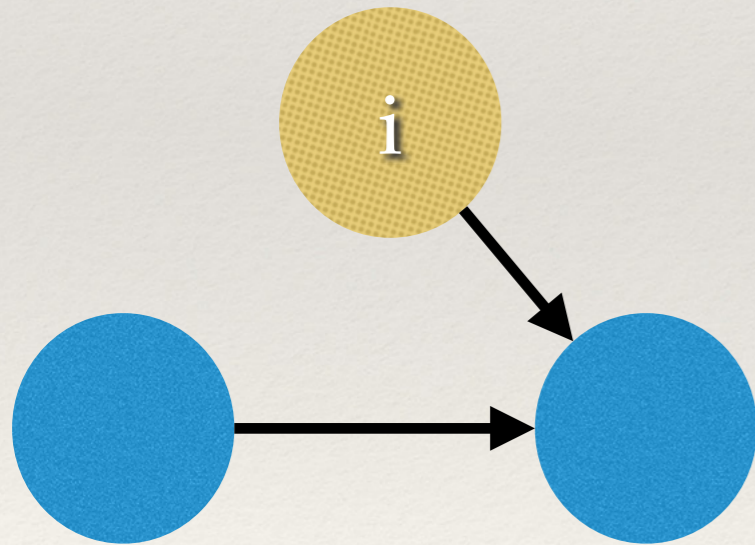
- ❖ The **rate** function determines how many decisions actors make.
- ❖ The **objective** function expresses how likely it is for an actor to change his/her network in a particular way.
- ❖ Represents the short-term *objectives* of the actor (hence the name).
 - ❖ “defined on the set of possible states of the network, as perceived from the point of view of the focal actor” (Snijders et al. 2010).



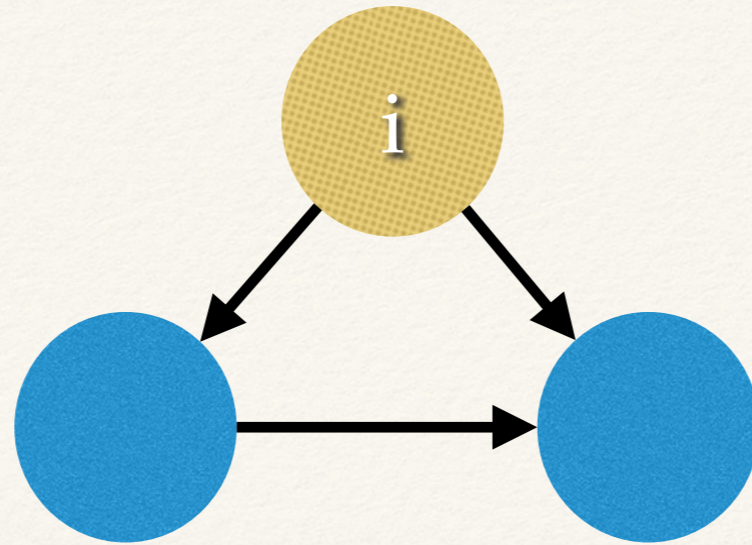
What can i do?



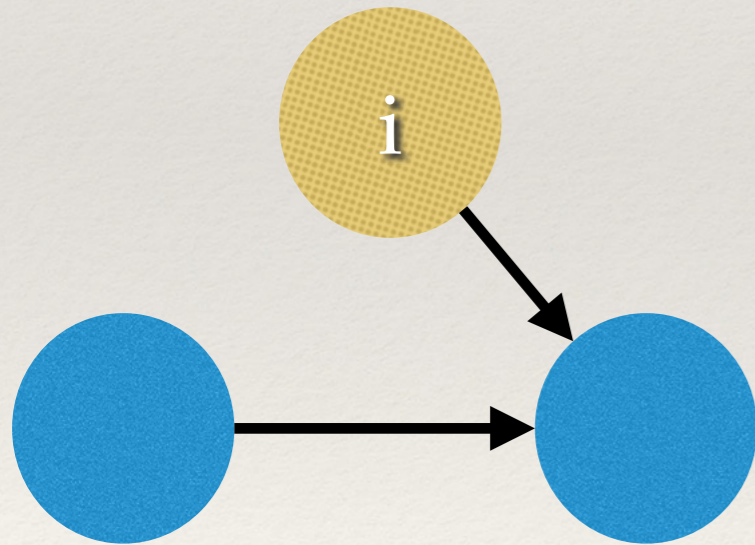
What can *i* do?



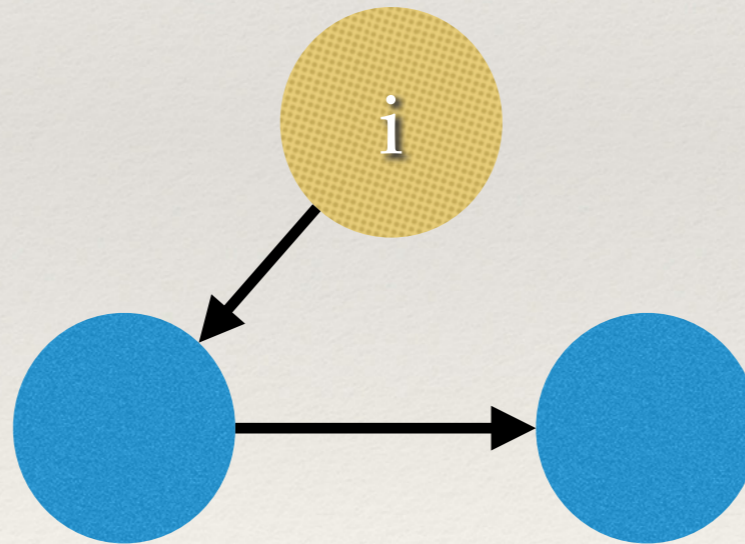
Drop an
existing tie



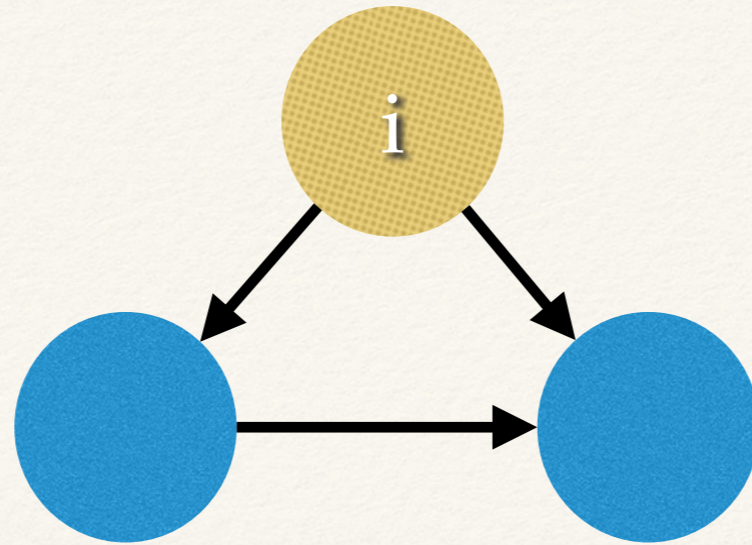
What can *i* do?



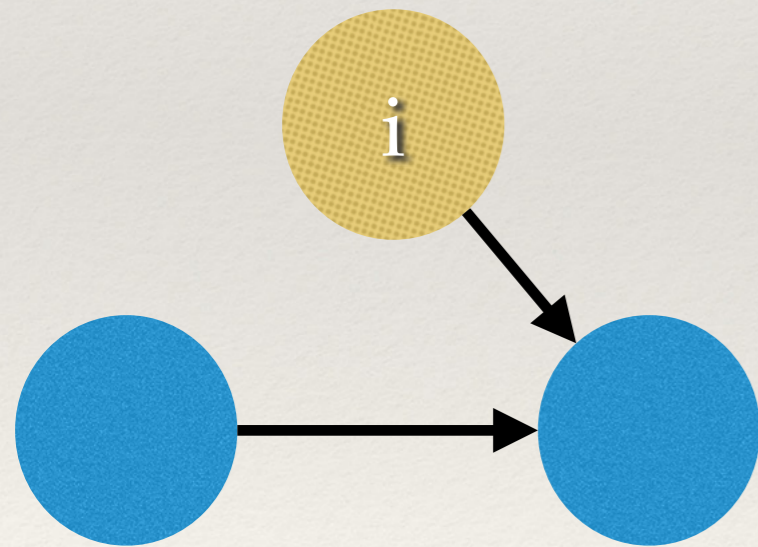
Drop an existing tie



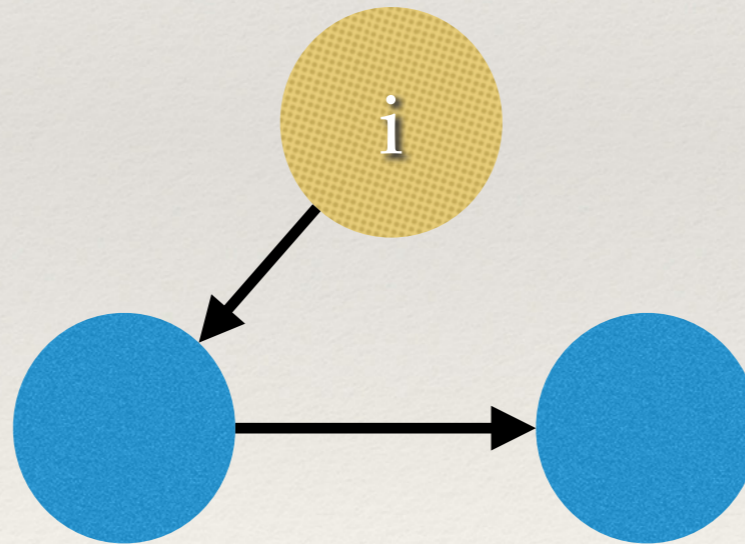
Or... drop a different tie



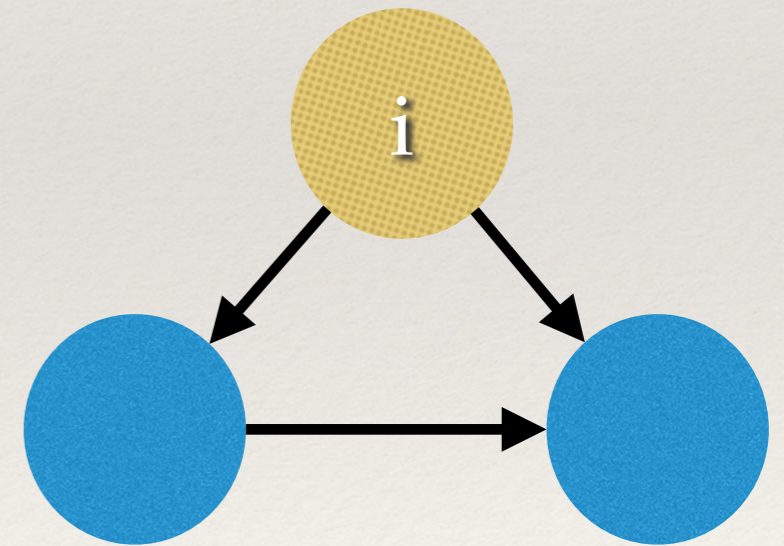
What can *i* do?



Drop an existing tie



Or... drop a different tie



Or... stay the same

Objective Function

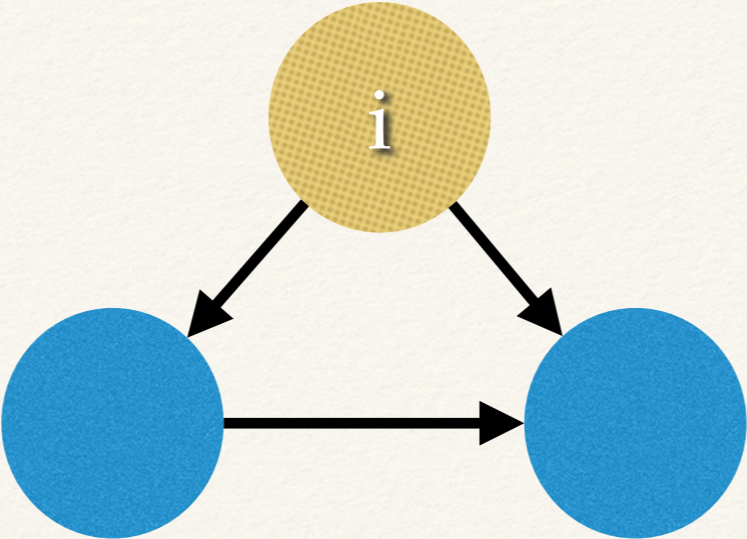
- ❖ The SABM simulates networks and compares them to the observed network.
- ❖ Basically, optimizing a random utility function.
 - ❖ Take an actor, evaluate what he/she can do, determine which is most likely.

Objective Function

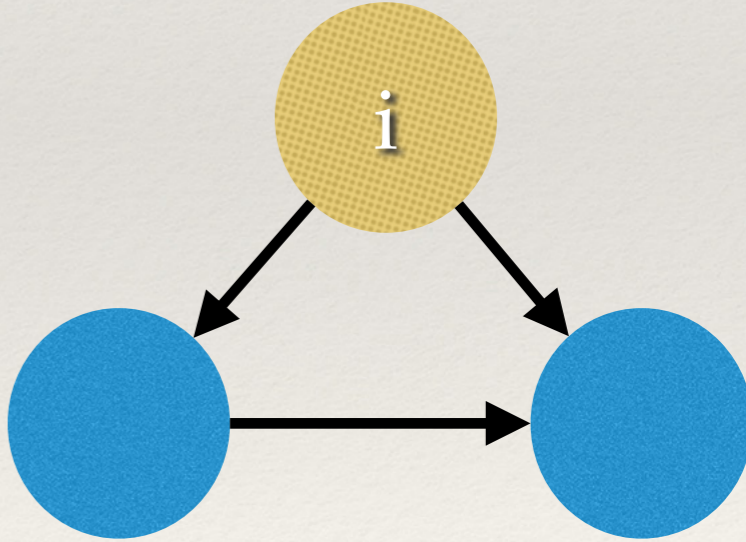
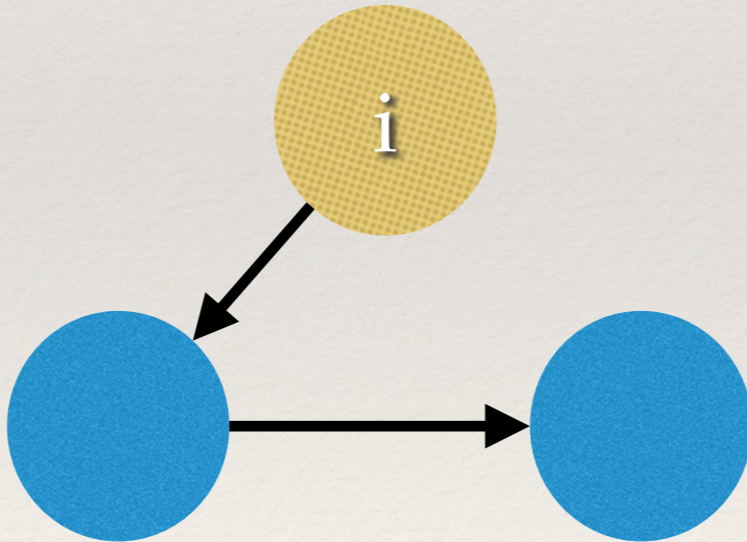
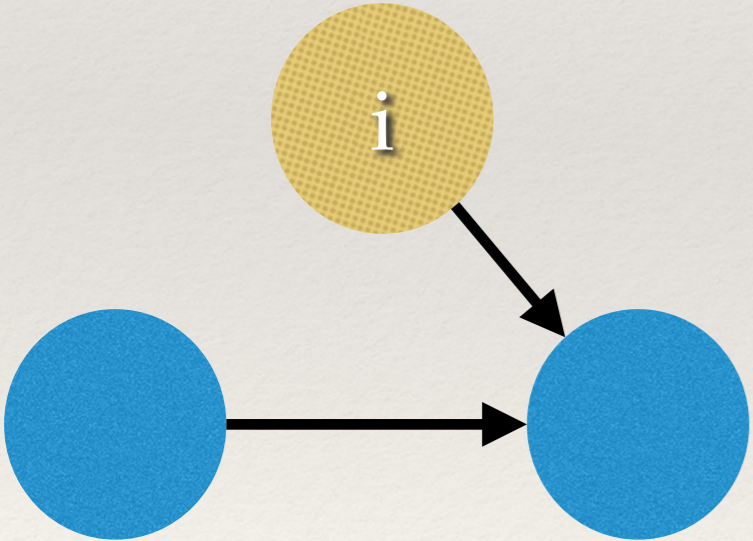
- ❖ The estimated model then yields estimates about actors' preferences
- ❖ **Positive** effects that are large and significantly different from zero indicate a preference over available alternatives for that particular configuration.
- ❖ The opposite for **negative** effects that are large and significantly different from zero.

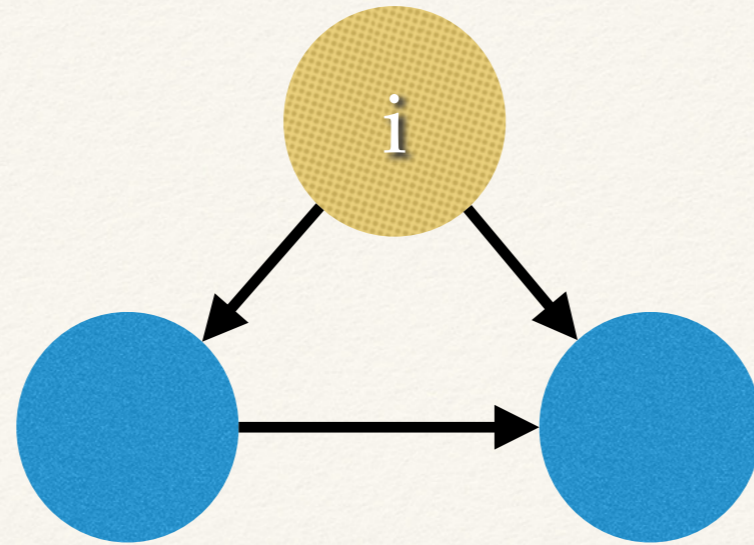
Objective Function

- ❖ As with ERGMs, network configurations operationalize the process in which we are interested.
- ❖ Thus, we can test hypotheses regarding actor-based mechanisms.
- ❖ Again, the difference is about what actors “want to do” or “prefer” (which is different from ERGMs).

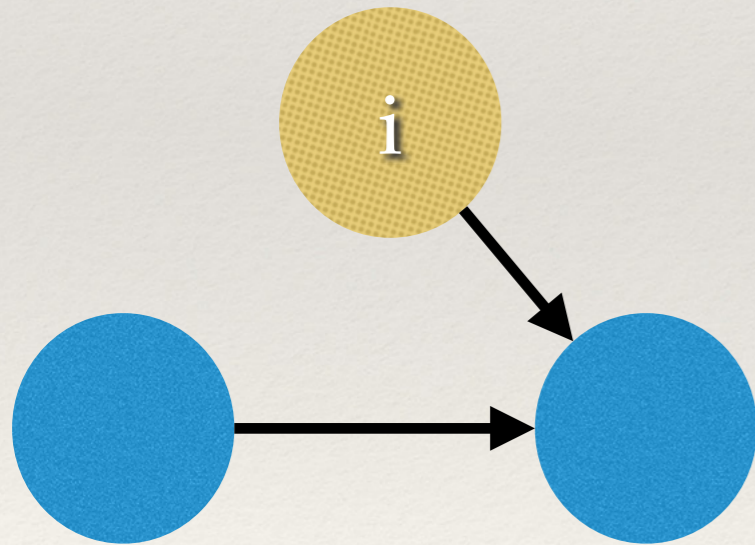


What can *i* do?

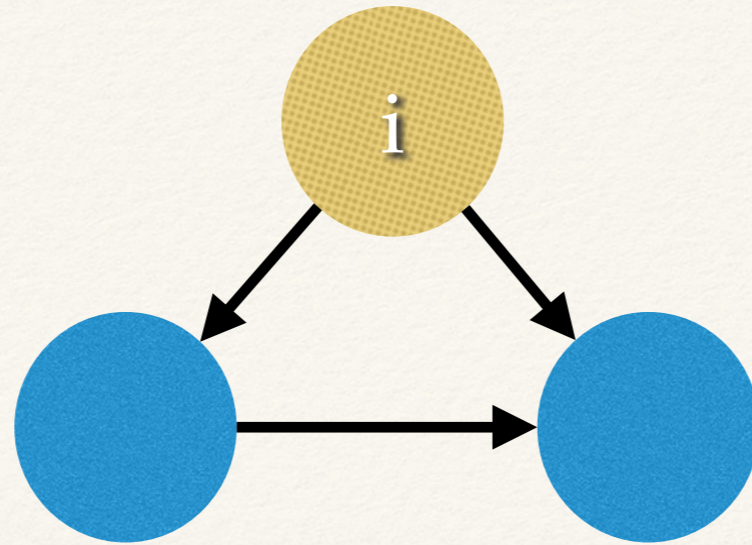




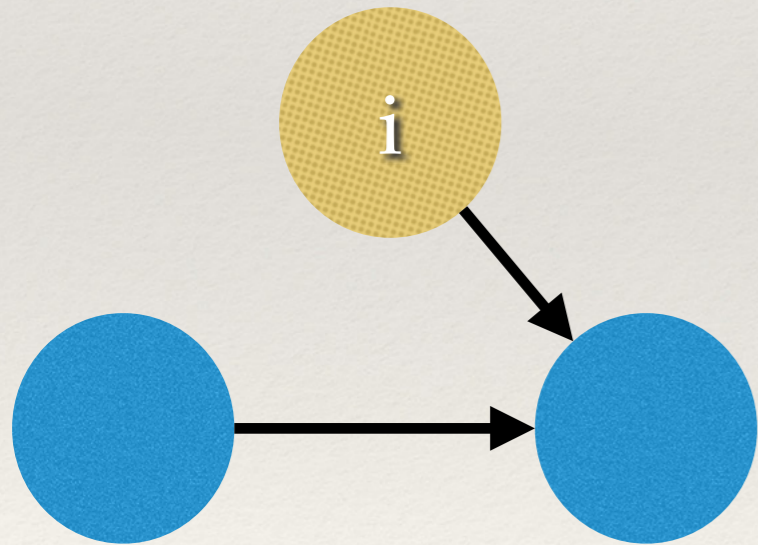
What can i do?



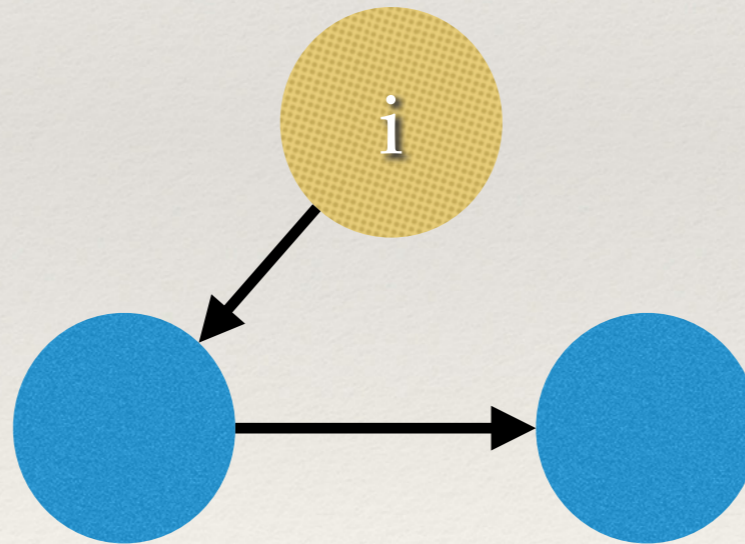
Drop ties to
unpopular others



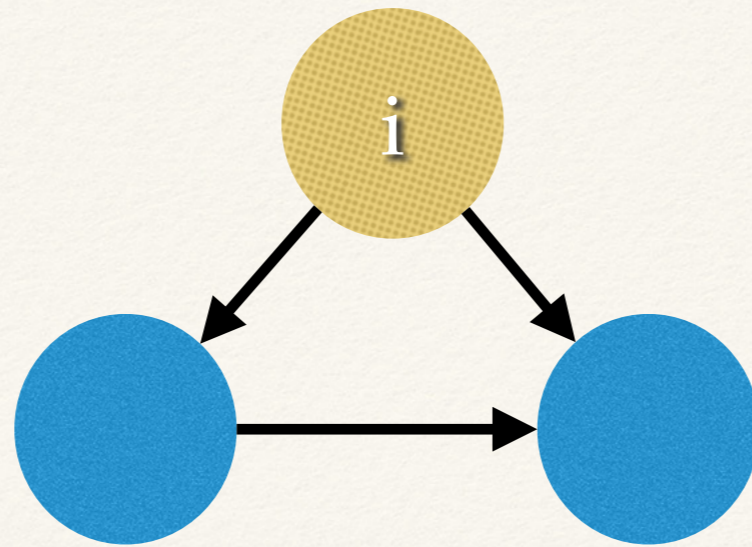
What can i do?



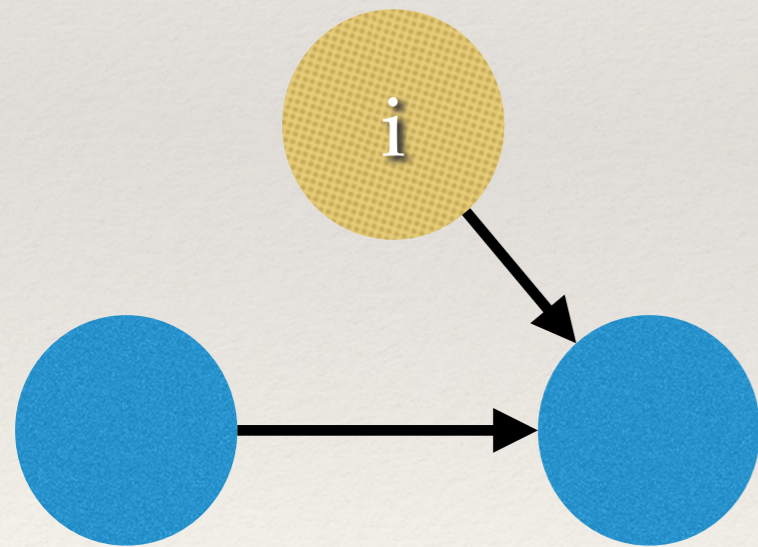
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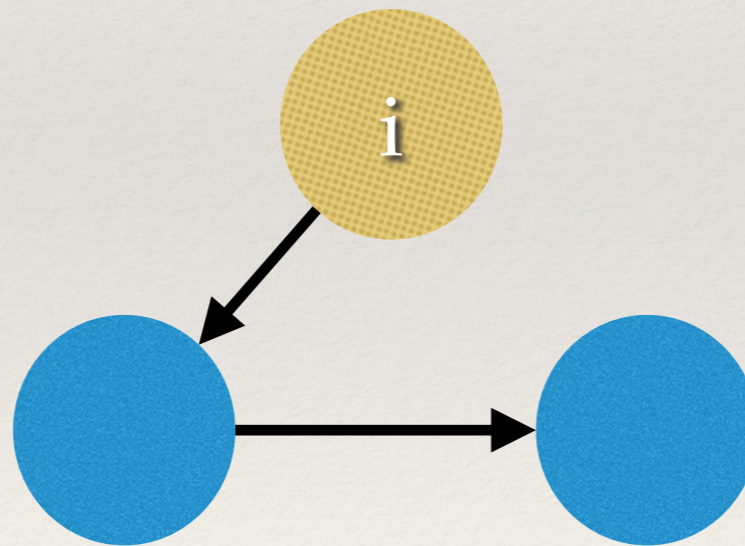
Drop ties to
popular others



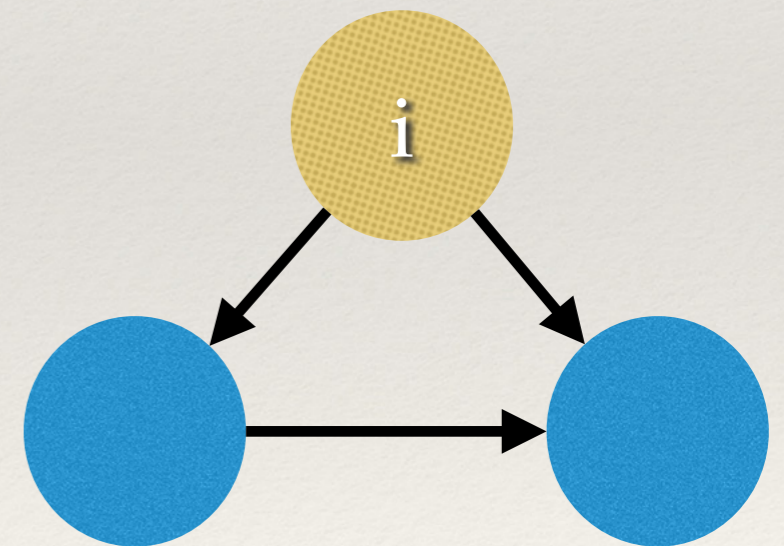
What can *i* do?



Drop ties to
unpopular others



Drop ties to
popular others



Be happy with 2
friends :)

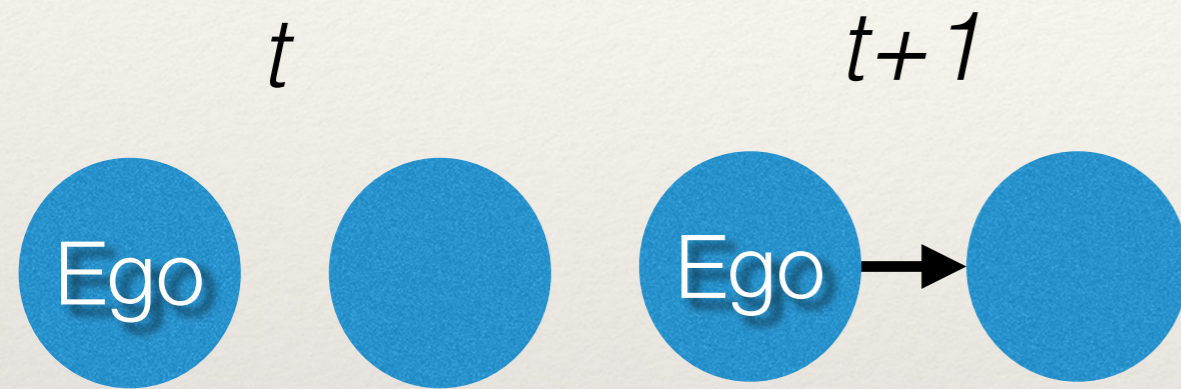
Basic Effects

t	$t+1$	<u>Effect</u> (RSiena term)	<u>Preference</u>
-----	-------	--------------------------------	-------------------

Basic Effects



Basic Effects



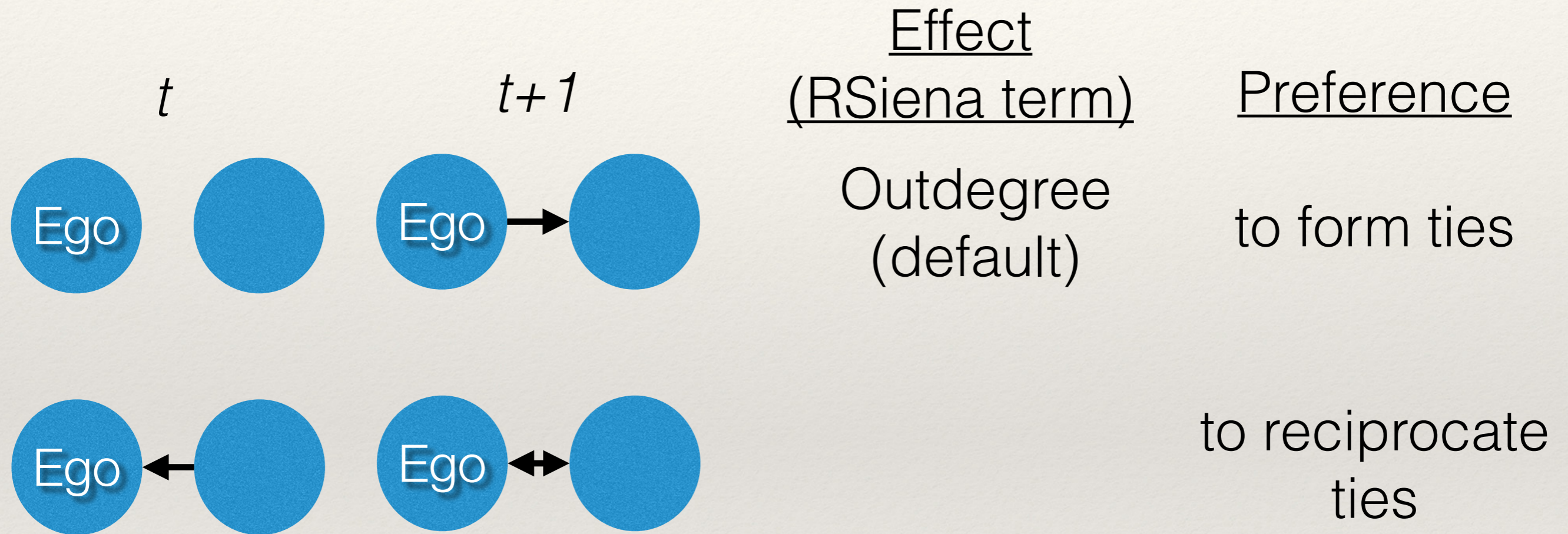
Effect
(RSiena term)

Outdegree
(default)

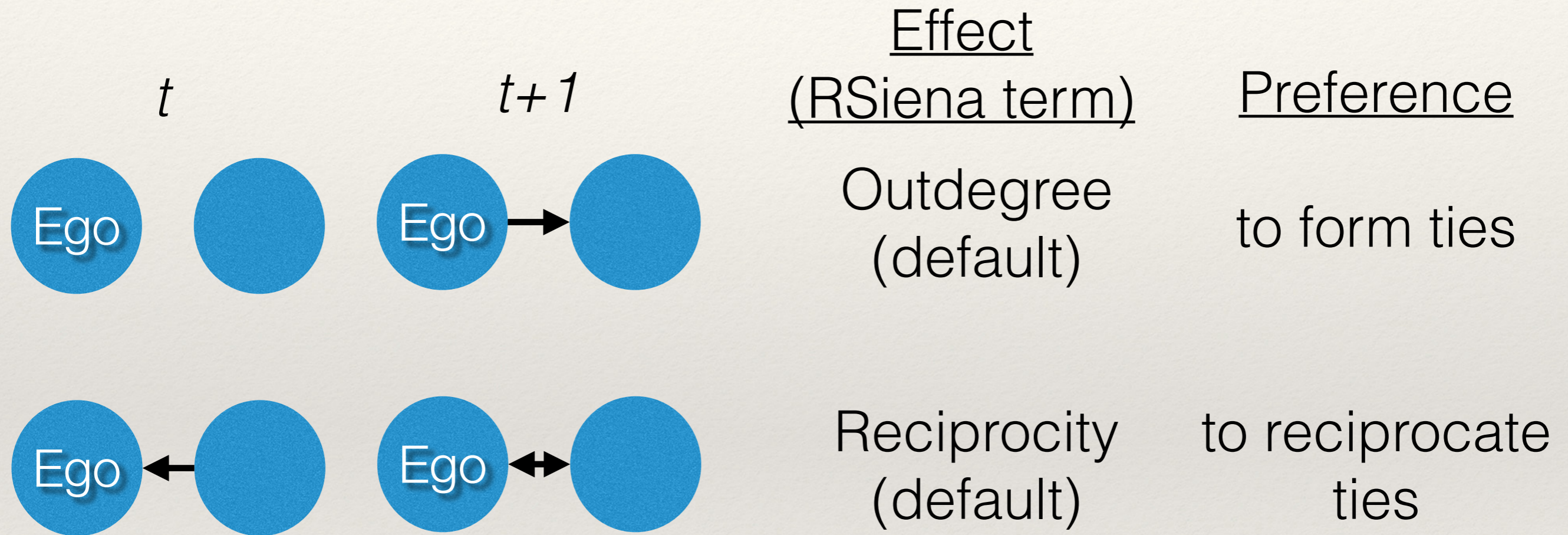
Preference

to form ties

Basic Effects



Basic Effects

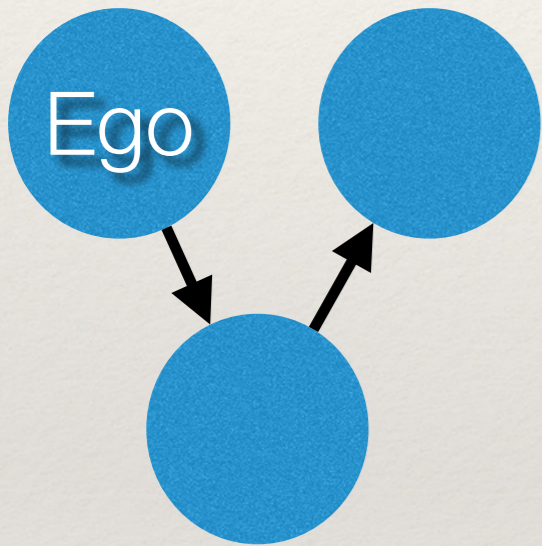


Additional Effects

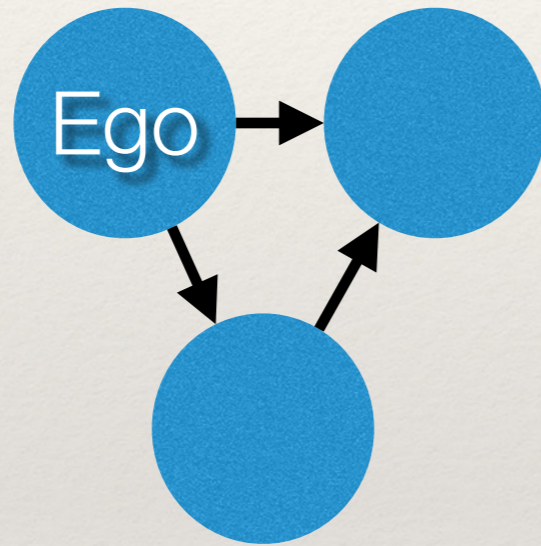
t	$t+1$	<u>Effect</u> (RSiena term)	<u>Preference</u>
-----	-------	--------------------------------	-------------------

Additional Effects

t



$t+1$



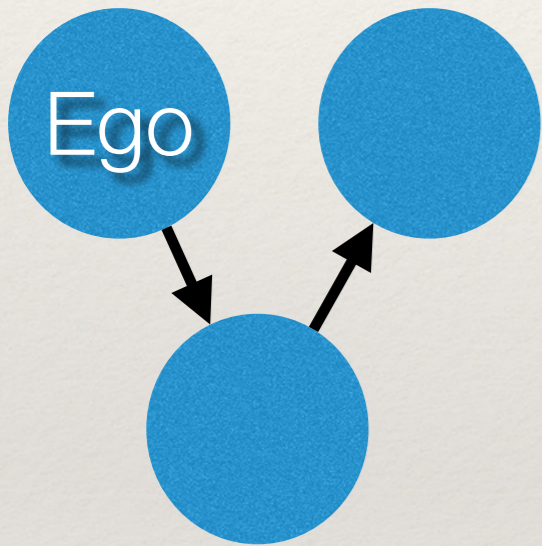
Effect
(RSiena term)

Preference

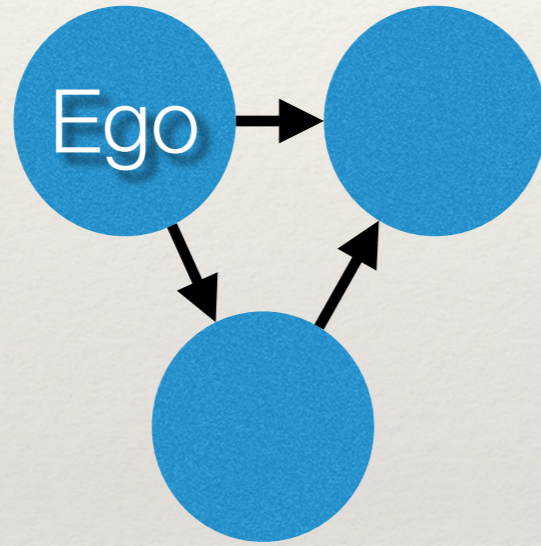
for being
friend of the
friend's friends

Additional Effects

t



$t+1$



Effect
(RSiena term)

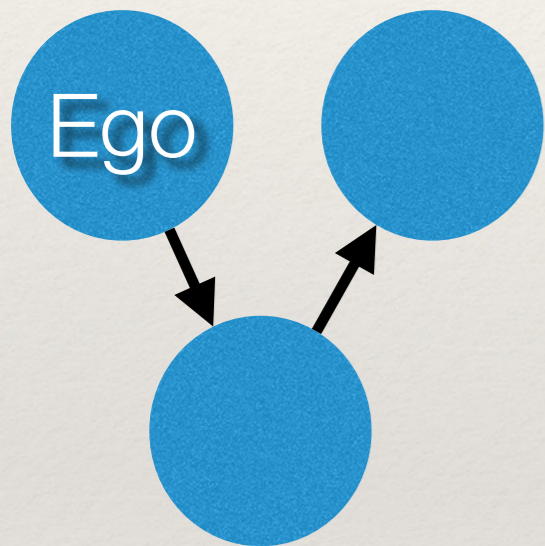
Transitive
Triplets
(transTrip)

Preference

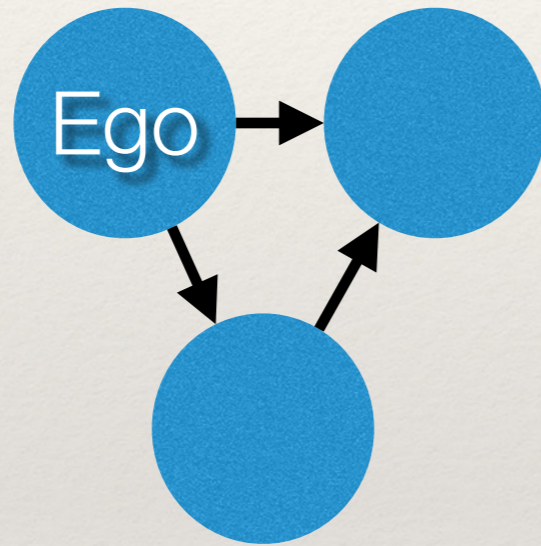
for being
friend of the
friend's friends

Additional Effects

t



$t+1$

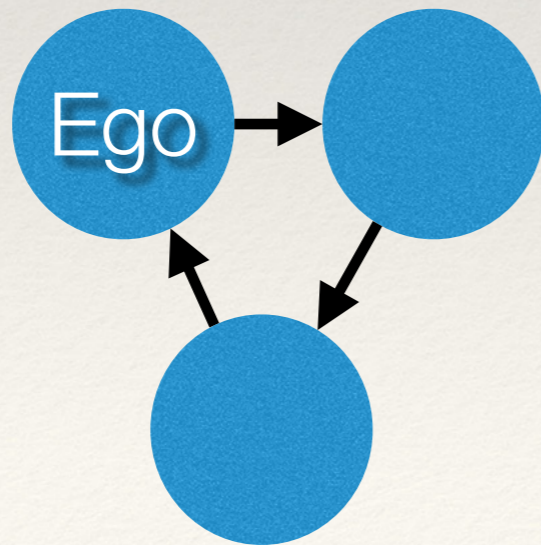
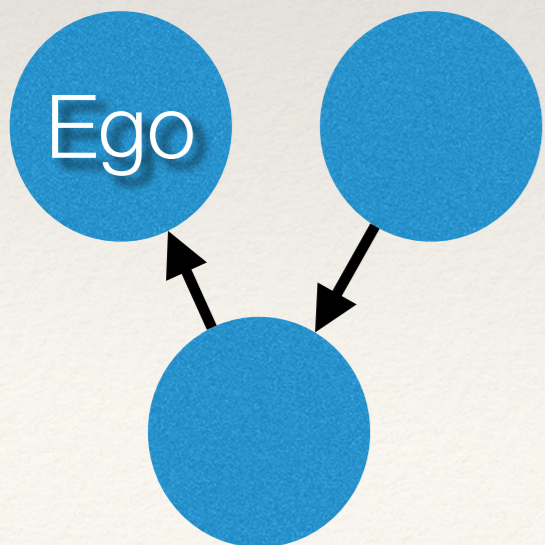


Effect
(RSiena term)

Transitive
Triplets
(transTrip)

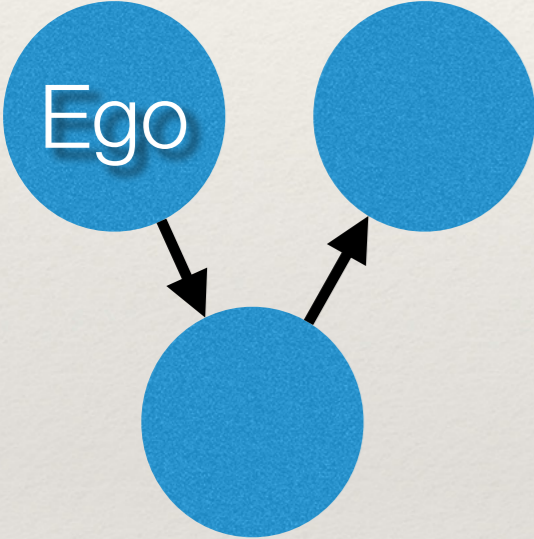
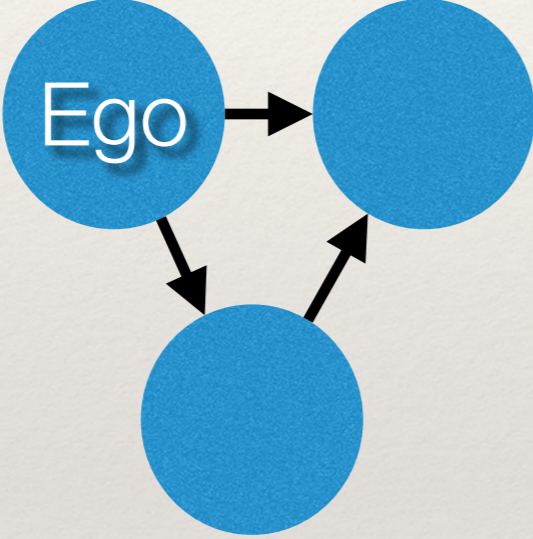
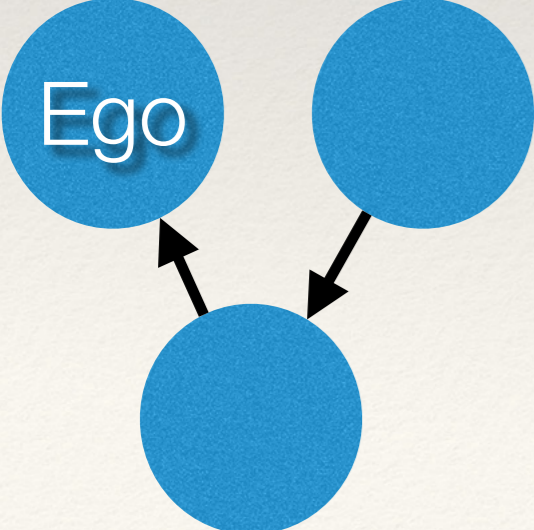
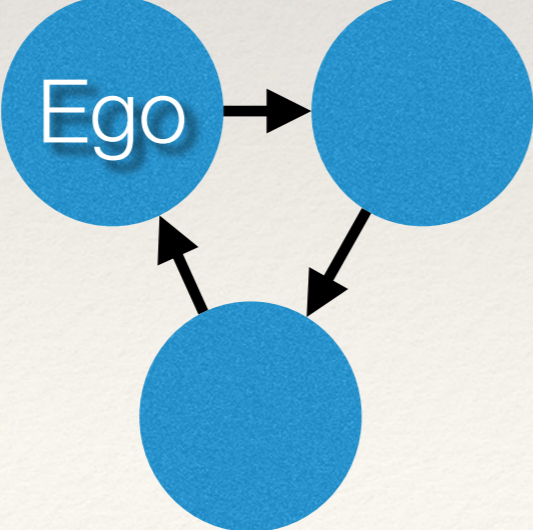
Preference

for being
friend of the
friend's friends



for forming
cycles

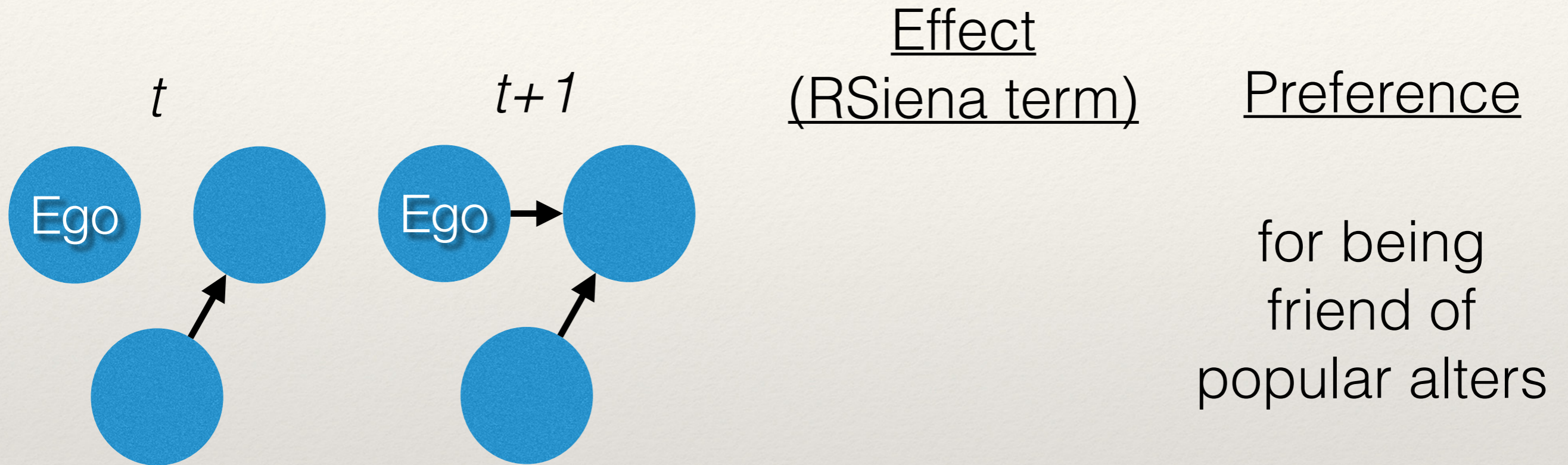
Additional Effects

		<u>Effect</u> (RSiena term)	<u>Preference</u>
t	$t+1$		
		Transitive Triplets (transTrip)	for being friend of the friend's friends
		Cyclical Triad (cycle3)	for forming cycles

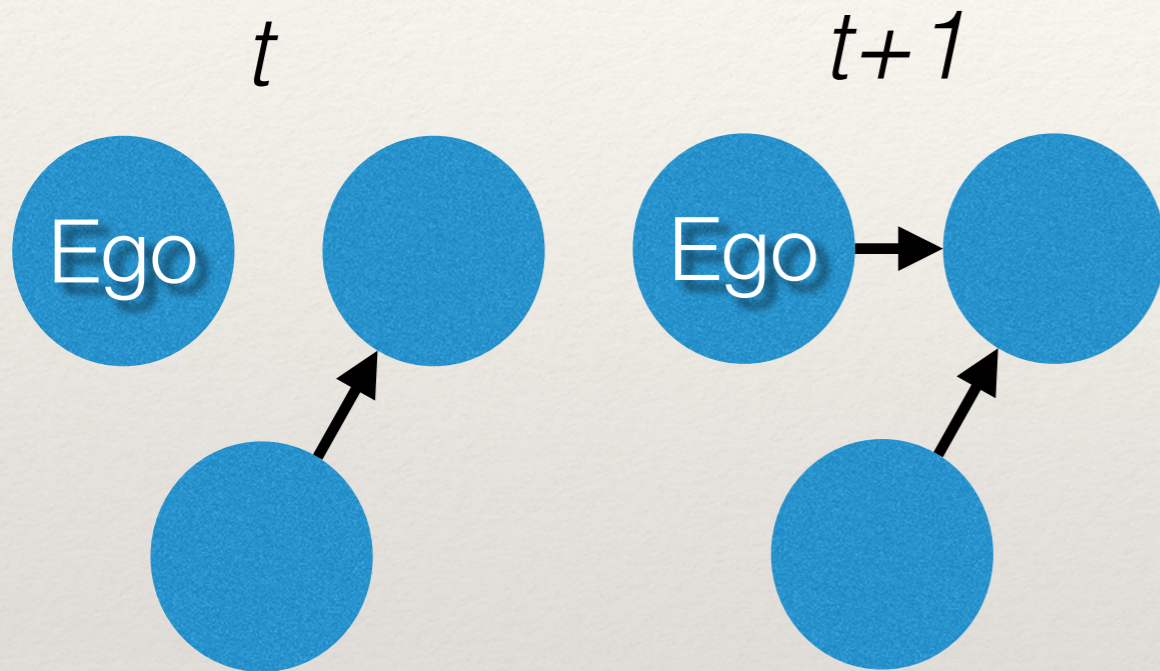
Additional Effects

t	$t+1$	<u>Effect</u> (RSiena term)	<u>Preference</u>
-----	-------	--------------------------------	-------------------

Additional Effects



Additional Effects



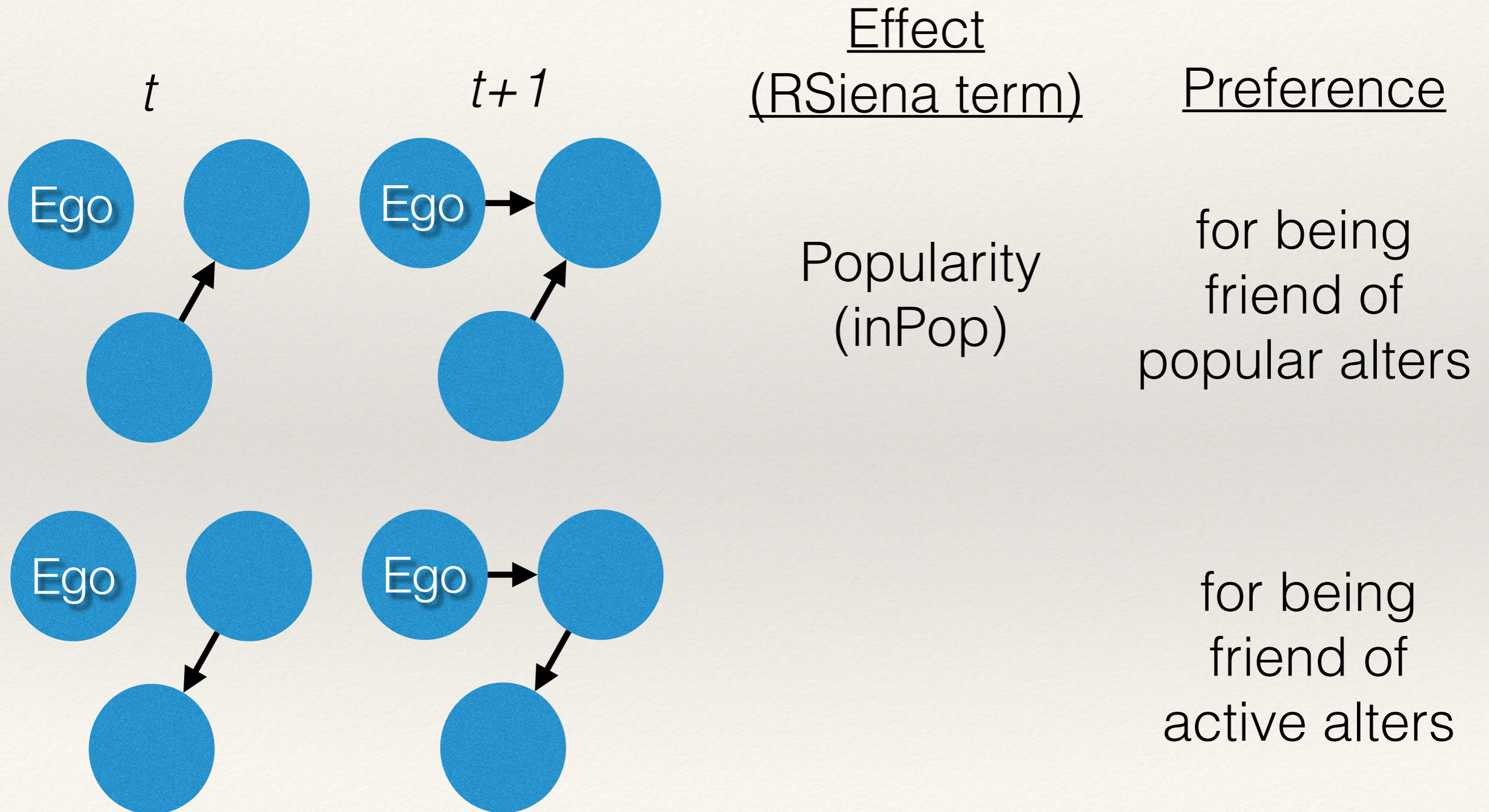
Effect
(RSiena term)

Popularity
(inPop)

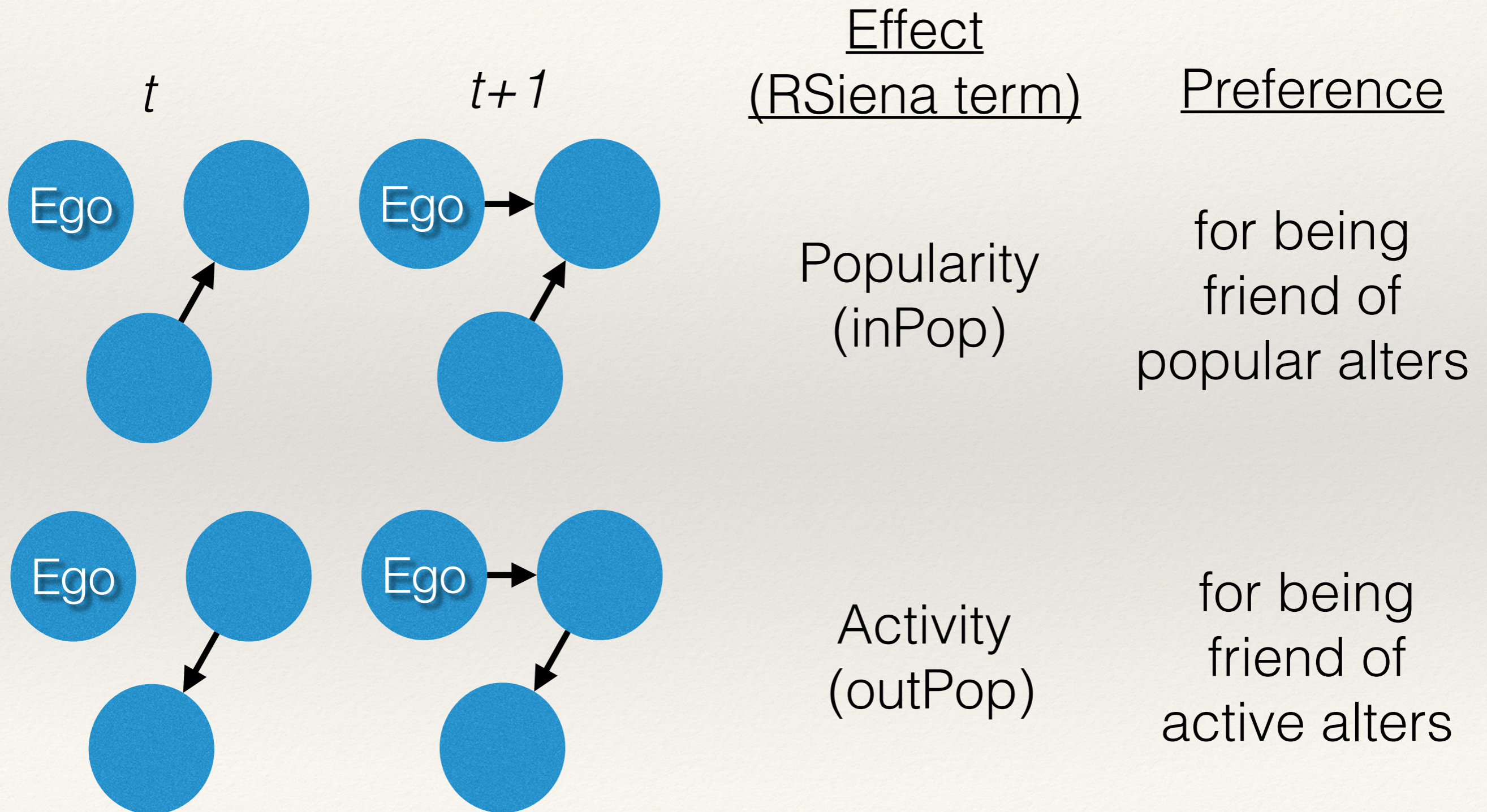
Preference

for being
friend of
popular alters

Additional Effects

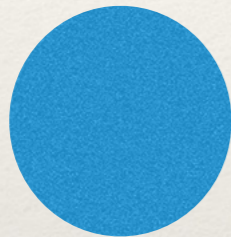


Additional Effects

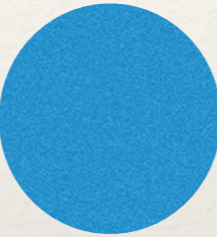


Interactions w/ Covariates

t



$t+1$



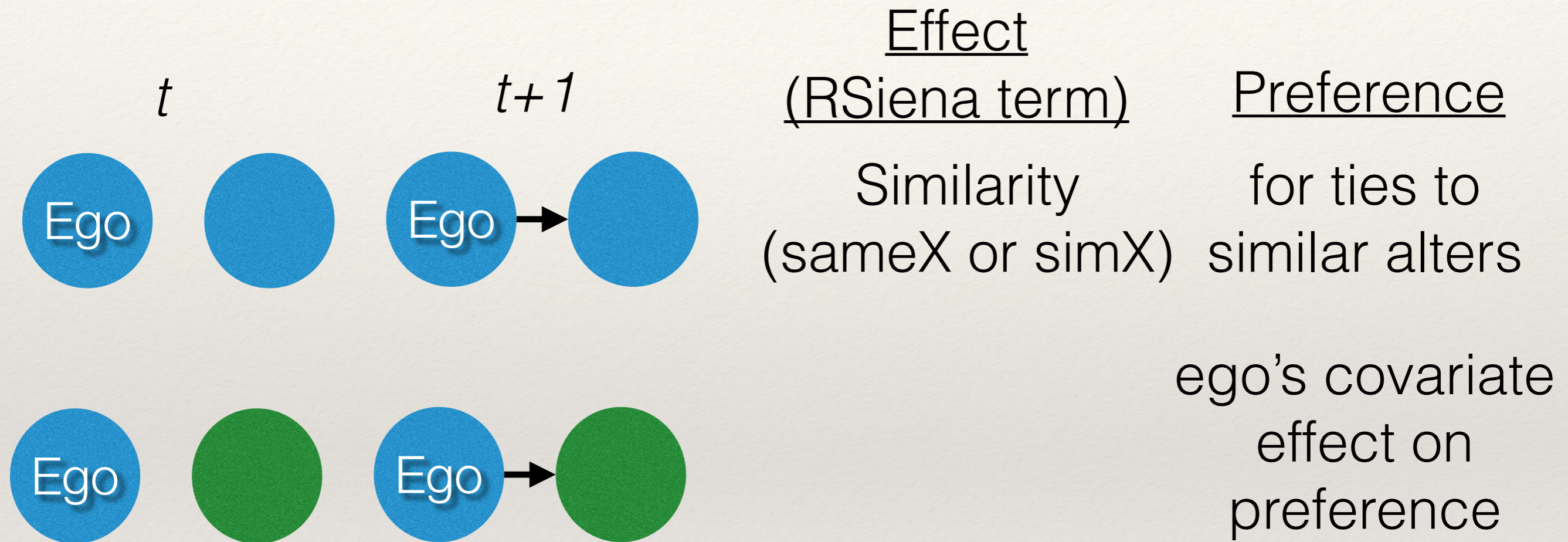
Effect
(RSiena term)

Preference
for ties to
similar alters

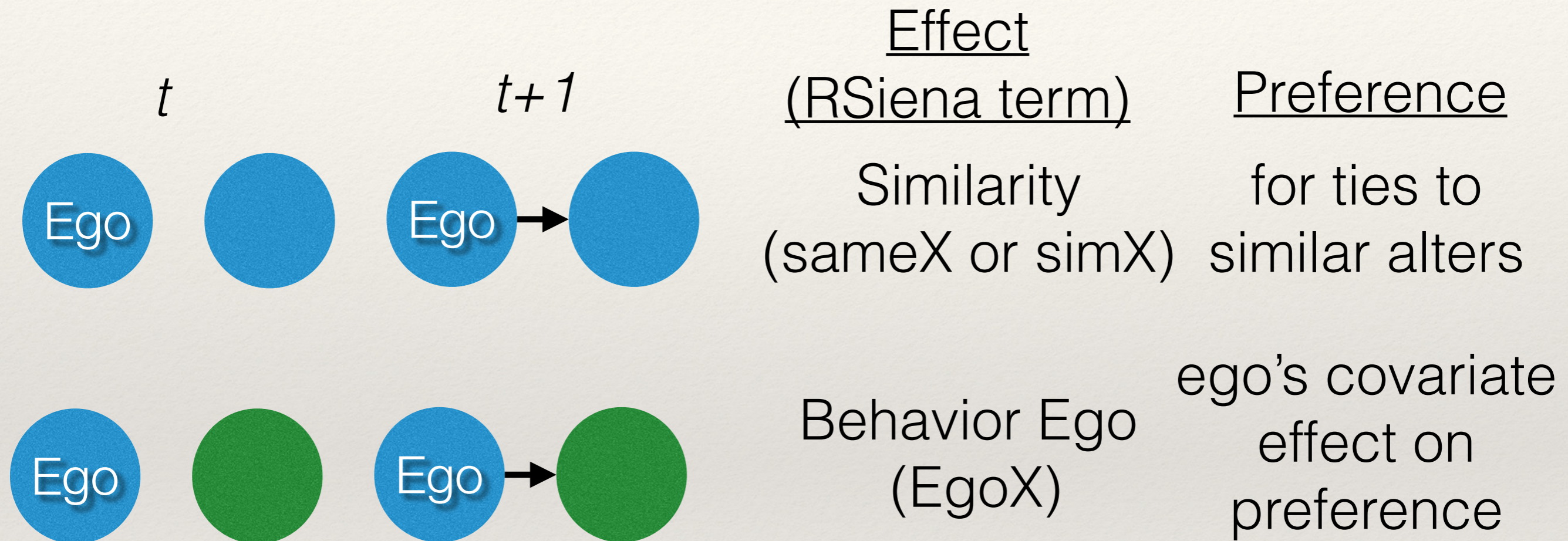
Interactions w/ Covariates



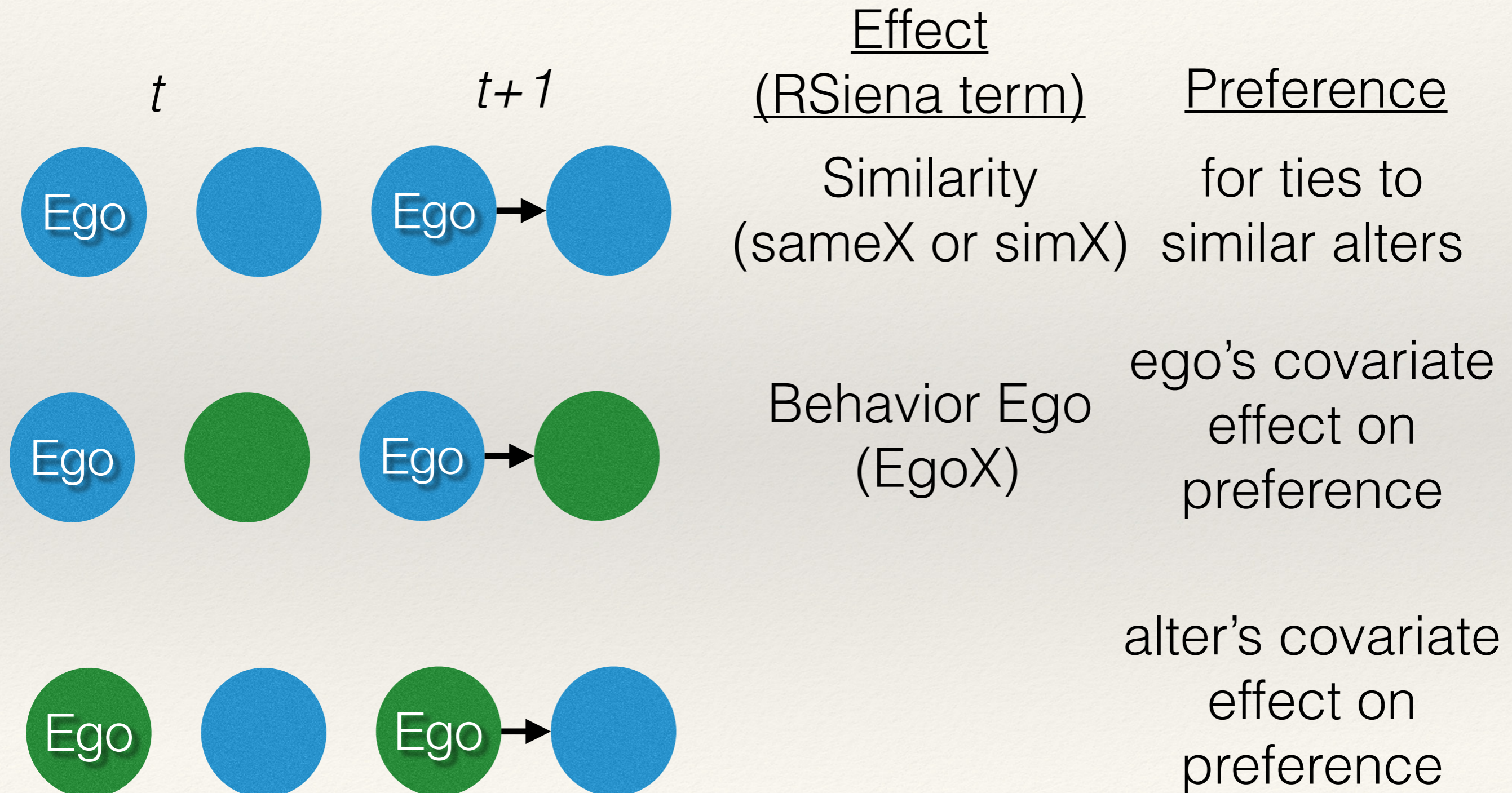
Interactions w/ Covariates



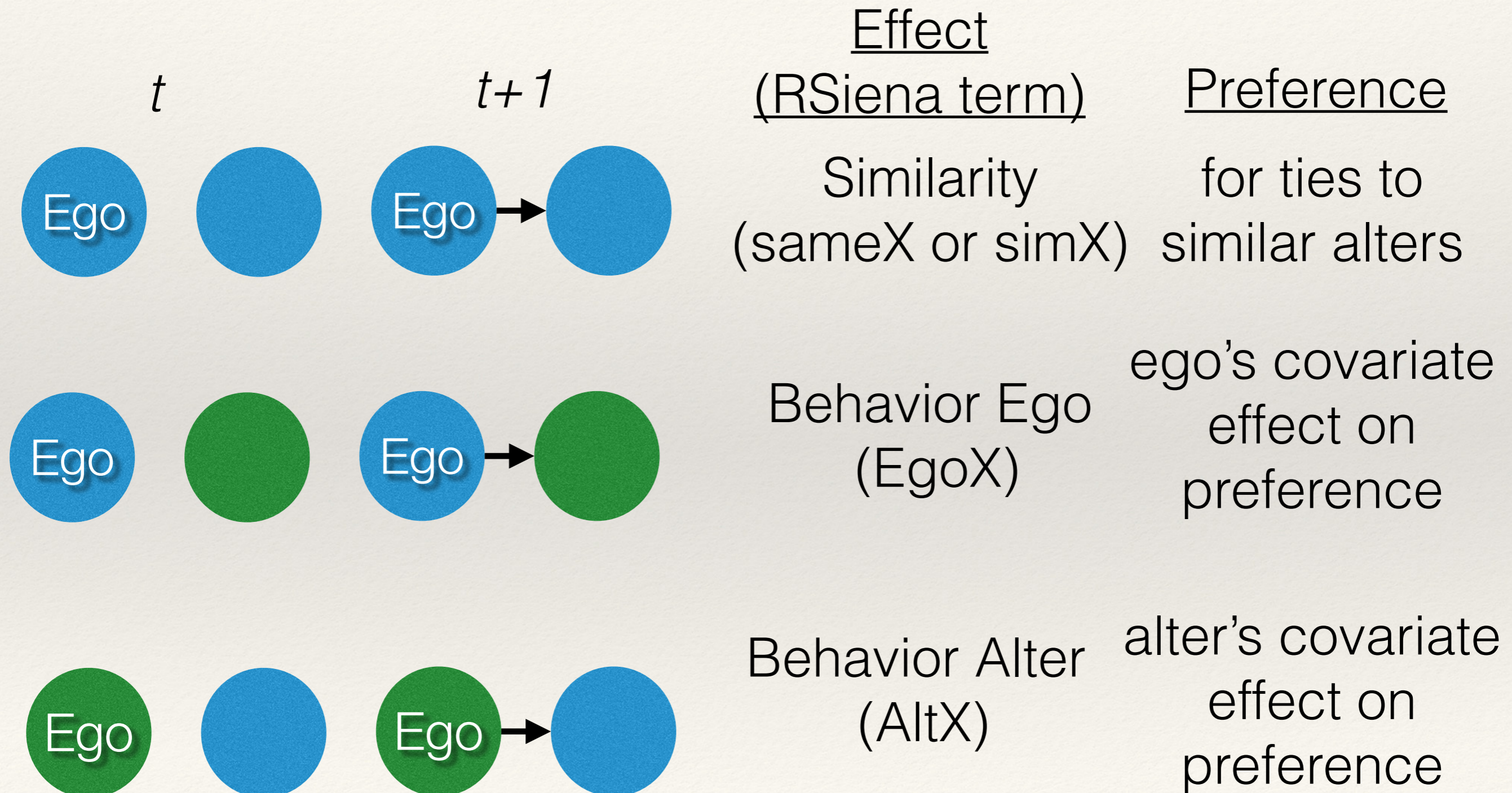
Interactions w/ Covariates



Interactions w/ Covariates



Interactions w/ Covariates



Motivating Example



CRIMINOLOGY

**VIOLENT OFFENDING AND VICTIMIZATION IN
ADOLESCENCE: SOCIAL NETWORK MECHANISMS
AND HOMOPHILY***

JILLIAN J. TURANOVIC¹ and JACOB T.N. YOUNG²

¹College of Criminology and Criminal Justice, Florida State University

²School of Criminology and Criminal Justice, Arizona State University

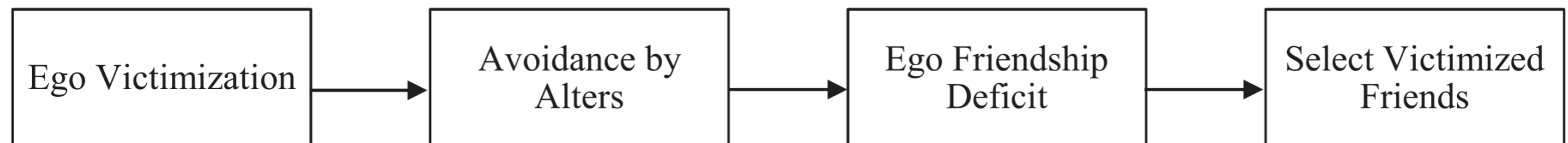
Motivating Example

❖ What are the findings?

a Violence homophily through preference for similarity:

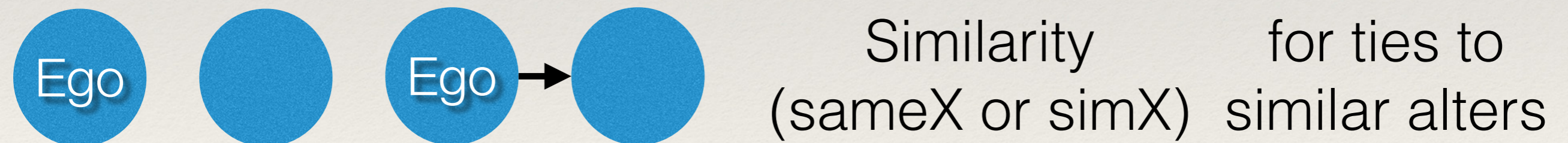


b Victimization homophily through avoidance:



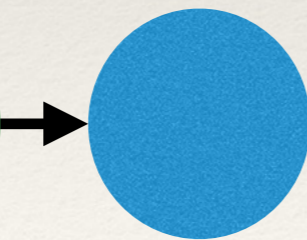
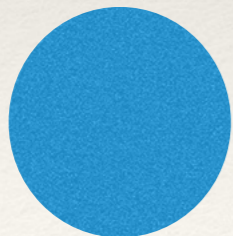
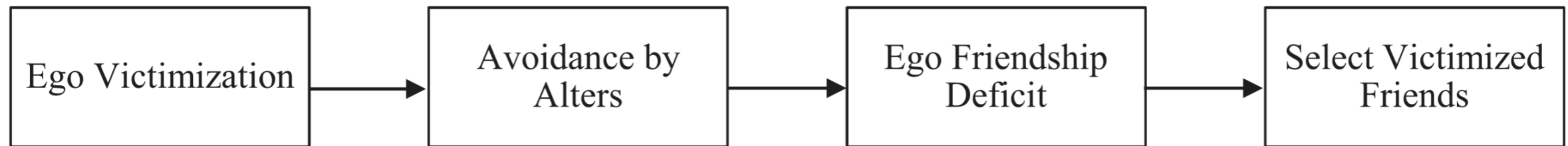
Motivating Example

a Violence homophily through preference for similarity:



Motivating Example

b Victimization homophily through avoidance:



Behavior Alter
(AltX)

alter's covariate
effect on
preference

Learning Goals

- ❖ By the end of this lecture, you should be able to answer these questions:
 - ❖ What is the difference between Exponential Random Graph Models (ERGMs) and Stochastic Actor-Based Models (SABMs)?
 - ❖ What is the logic of *micro-steps* and the simulation of networks using the **rate** function?
 - ❖ What is the logic of *preferences* and the simulation of networks using the **objective** function?

Questions?