

Lecture Feb 12

# Dynamical phenomena in single and interacting networks

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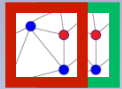
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# Outline



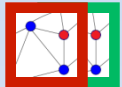
1. Introduction: failures & recoveries



2.1 **Single networks** phase diagram



2.2 Finite size effects



3.1 **Interacting networks** phase diagram



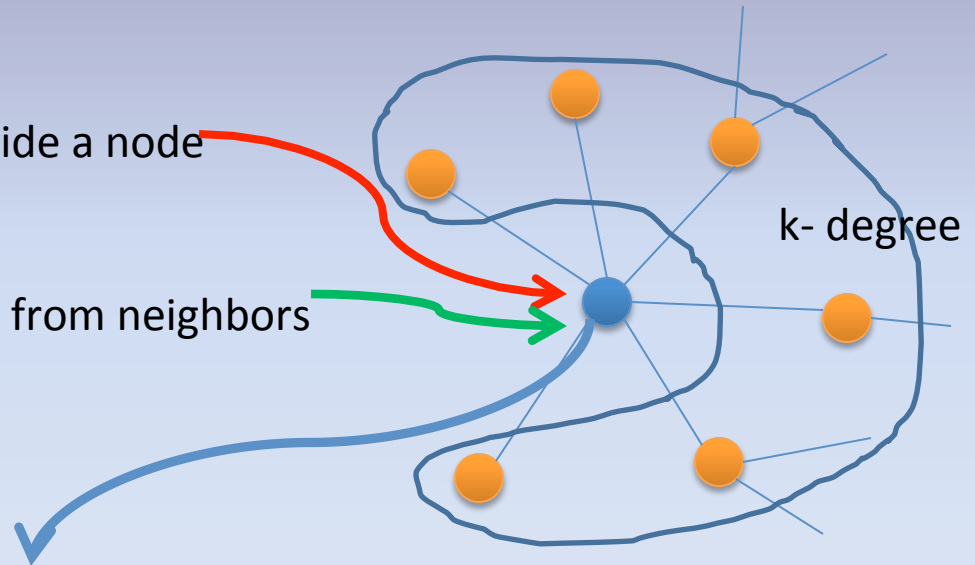
3.2 Finite size effects

# A network model with failures and recoveries.

- Each node in a network can be **active** or **failed**.
- We suppose there are **TWO possible reasons for the nodes' failures**: INTERNAL and EXTERNAL.

1. **INTERNAL failure**: intrinsic reasons inside a node

2. **EXTERNAL failure**: damage “imported” from neighbors



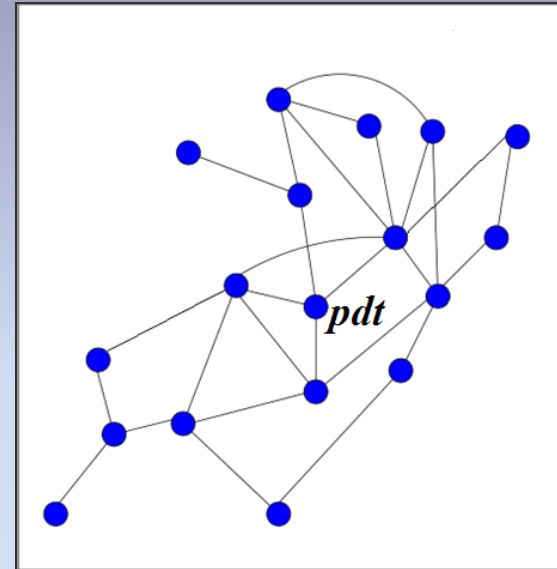
**RECOVERY**: A node can also **recover** from each kind of failure.

LET'S SPECIFY/MODEL THE RULES.

# 1. INTERNAL FAILURES

**p**- rate of internal failures (per unit time, for each node).  
During interval  $dt$ , there is probability  $pdt$  that the node fails.

**Recovery:** A node *recovers* from an internal failure after a time period  $\tau$ .



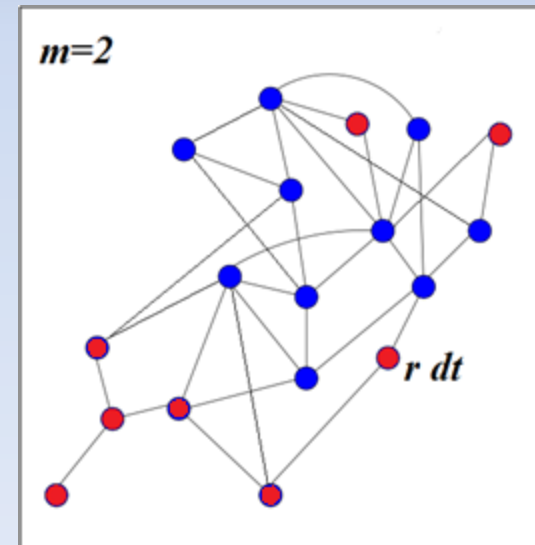
## 2. EXTERNAL FAILURES – if the neighborhood of a node is too damaged

IF: “CRITICALLY DAMAGED neighborhood”: **less than or equal to  $m$  active neighbors**,  
where  **$m$**  is a fixed treshold parameter.

THEN: There is a probability  $r dt$  that the node will experience externally-induced failure during  $dt$ .

$r$  - external failure rate

*A node recovers from an external failure after time  $\tau'$ .*



FAILURE TYPE	RULE	RECOVERY
Internal failure	With rate $p$ on each node	After time $\tau$
External failure	IF( $\leq m$ active neighbors) THEN Extra rate $r$ on each node	After time $\tau'$

Out of these 5 parameters, we fix three of them:  
 $m=4$ ,  $\tau=100$  and  $\tau'=1$ .

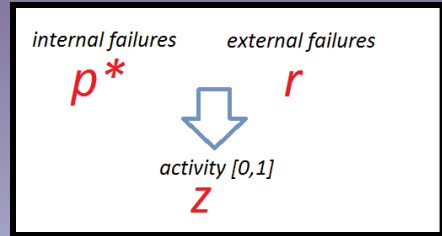
We let  $(p,r)$  to vary.

It turns out it is convenient to define  $p^*=exp(-p\tau)$ .  
 So we use  $(p^*,r)$  instead of  $(p,r)$ .

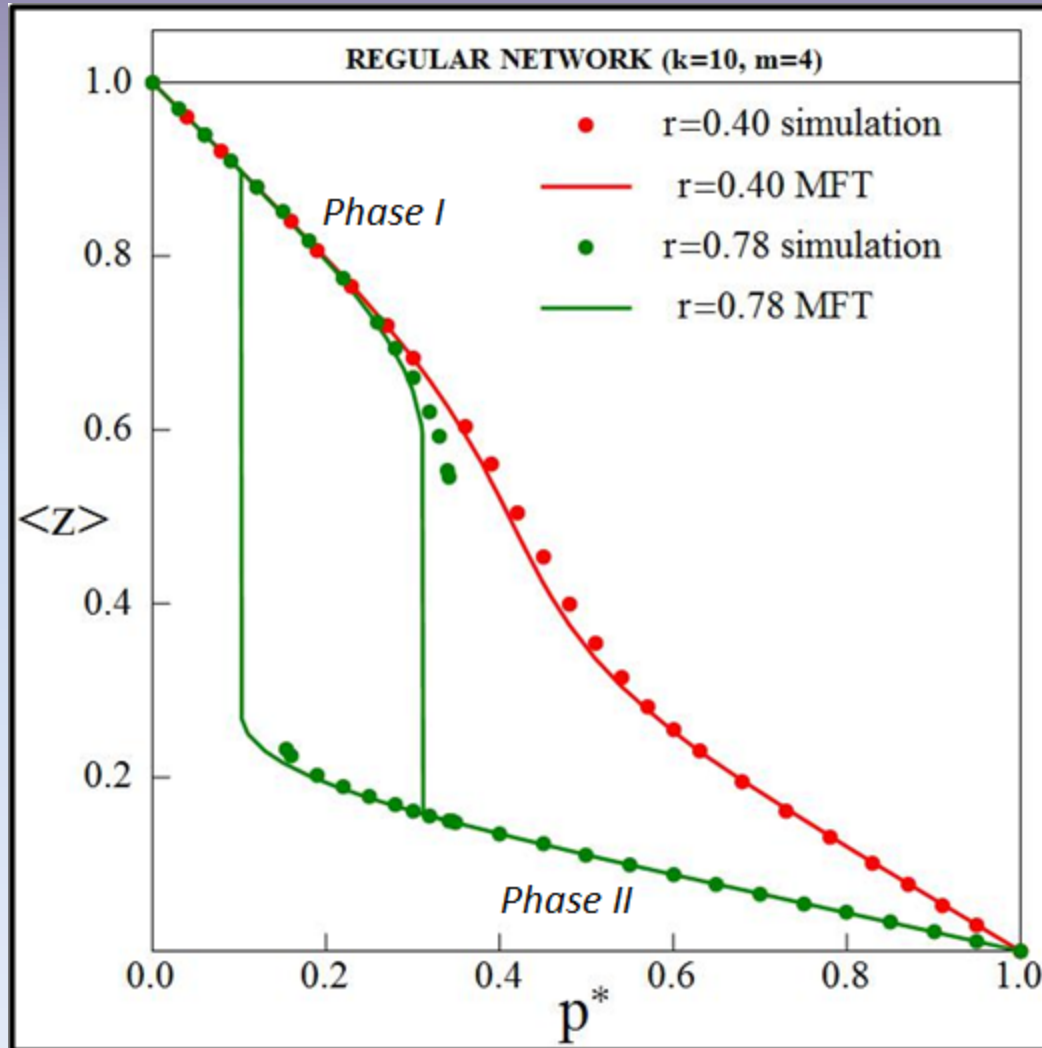
We measure activity  $Z$  of the network as a function of  $(p^*,r)$ .



# Model simulation [Random regular networks]



$\langle Z \rangle$  - average fraction of active nodes ( $Z$  fluctuates)

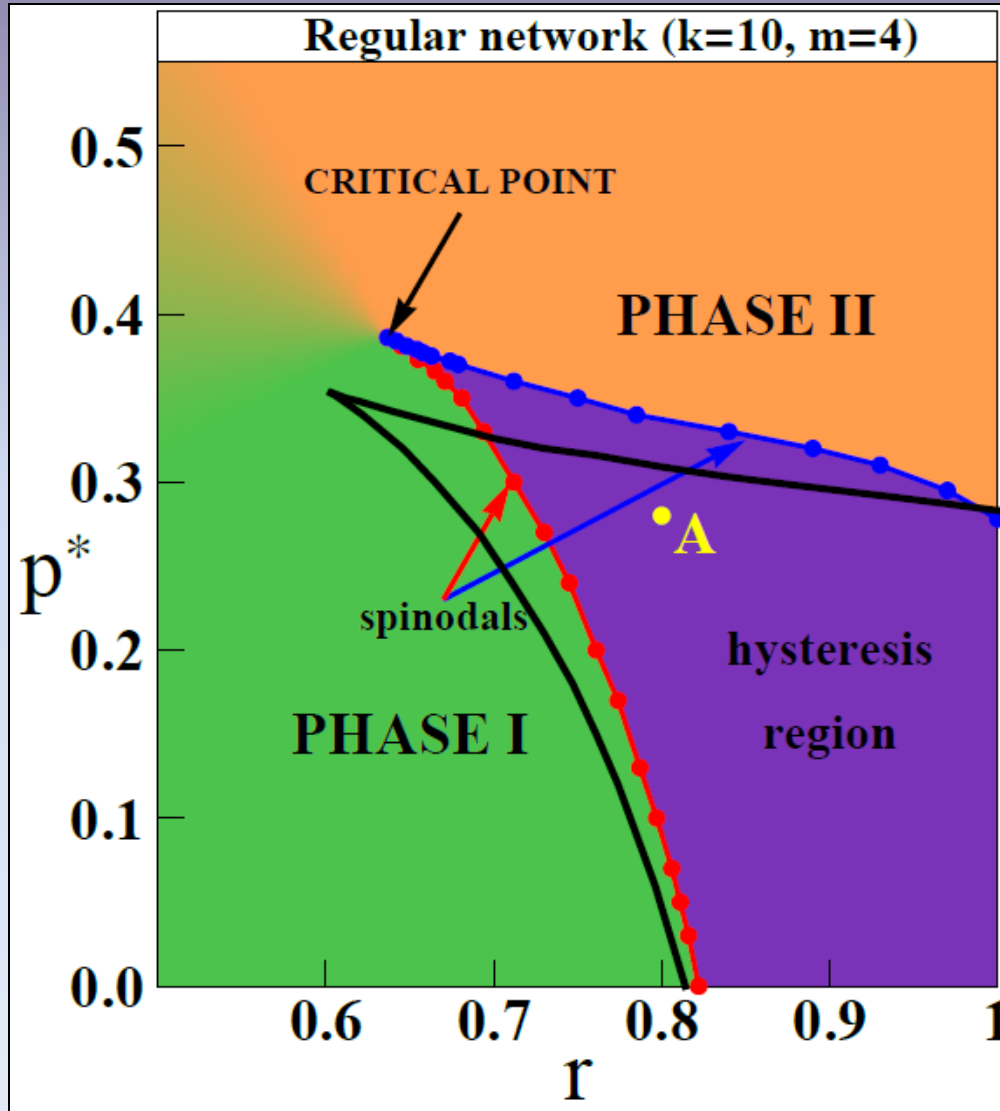


We fix  $r$ , and measure  $\langle z \rangle(p^*)$

For some values of  $r$  we have a hysteresis loop.



# Phase diagram (single network, random regular)



GREEN; High activity Z  
ORANGE: Low activity Z

In the hysteresis region both phases exist, depending on the initial conditions or the memory/past of the system.

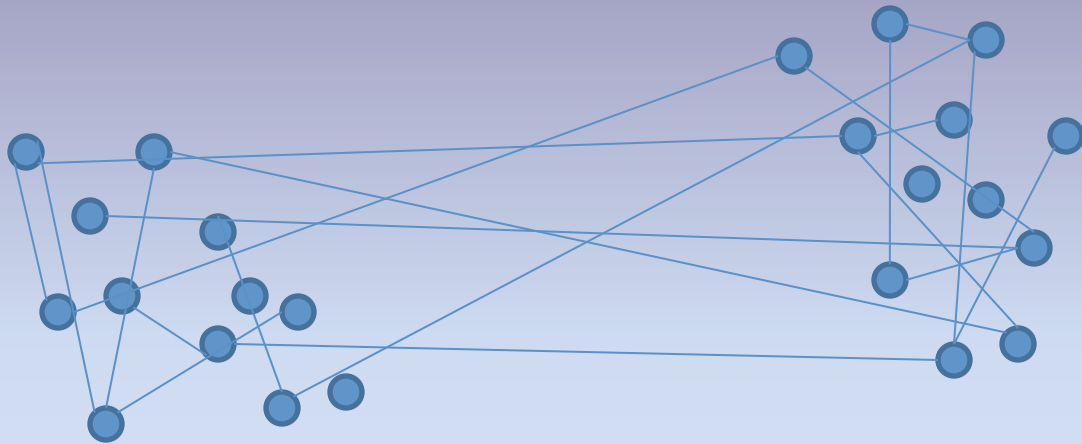
Blue line: critical line (spinodal) for the abrupt transition  $I \rightarrow II$

Red line: critical line (spinodal) for the abrupt transition  $II \rightarrow I$





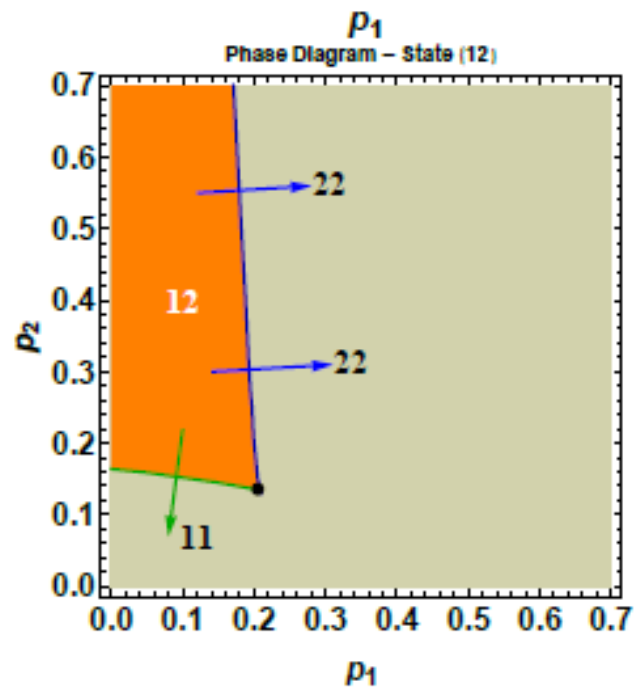
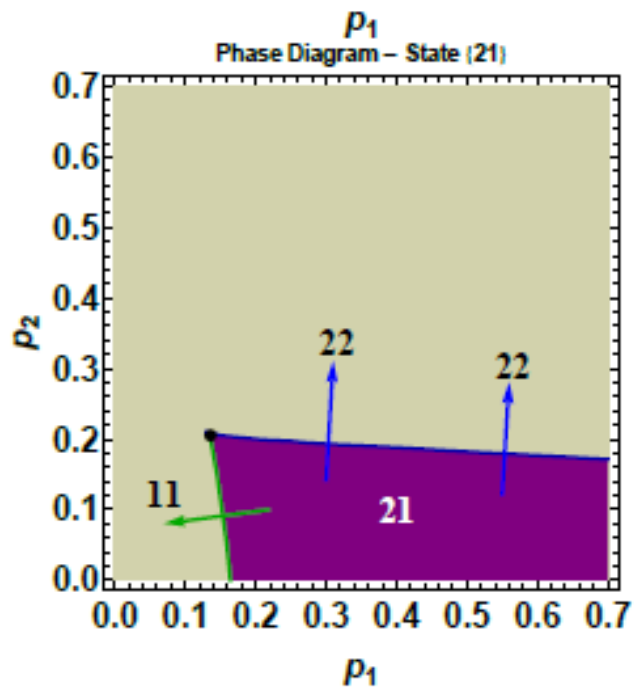
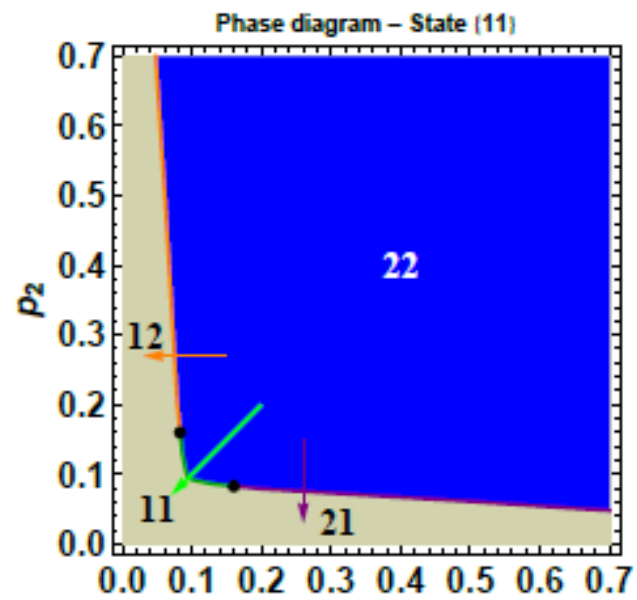
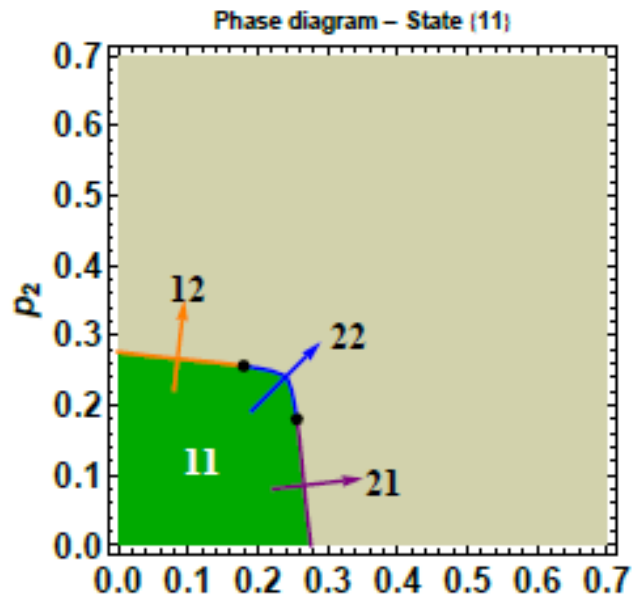
# 3.1. Interacting networks

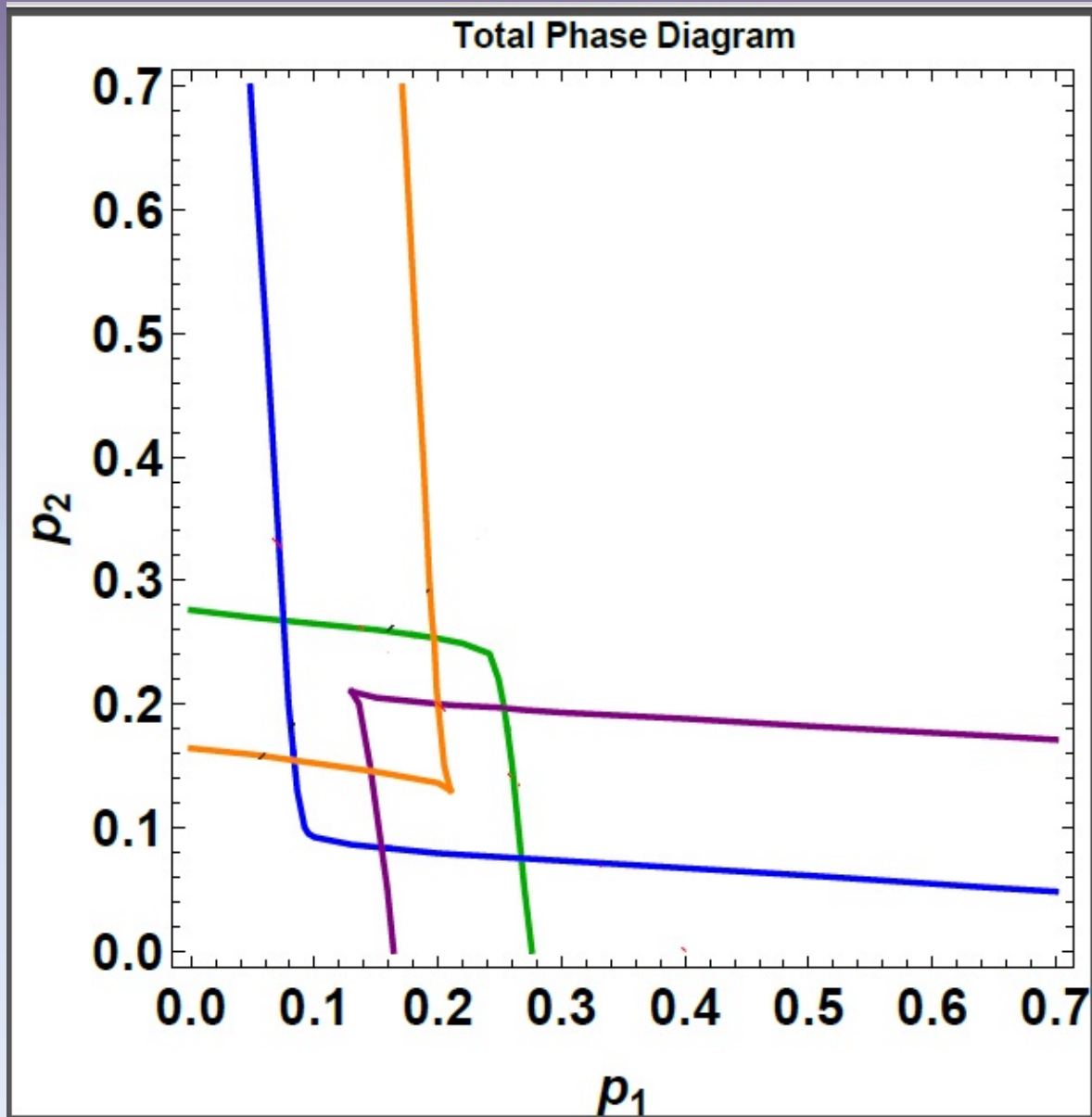


Network A

Network B

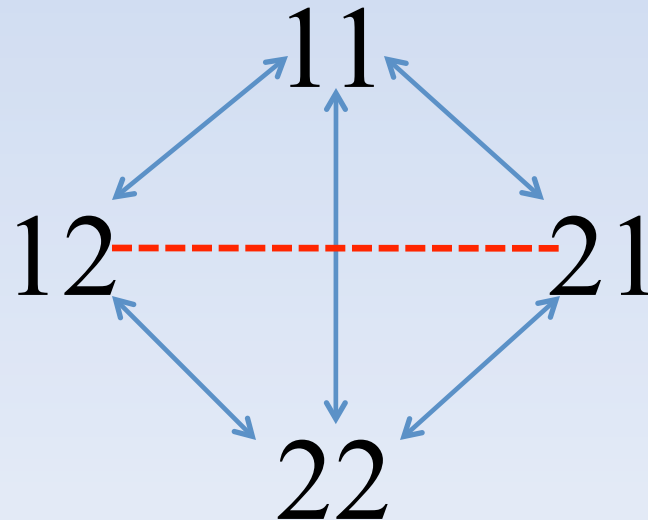
FAILURE TYPE	RULE	RECOVERY
Internal failure	With rate $p$ on each node	After time $\tau$
External failure	IF( $\leq m$ active neighbors) THEN Extra failure rate $r$	After time $\tau'$
Dependency failure	IF(companion node from the opposite network failed) THEN Extra failure rate $r_d$	After time $\tau''$





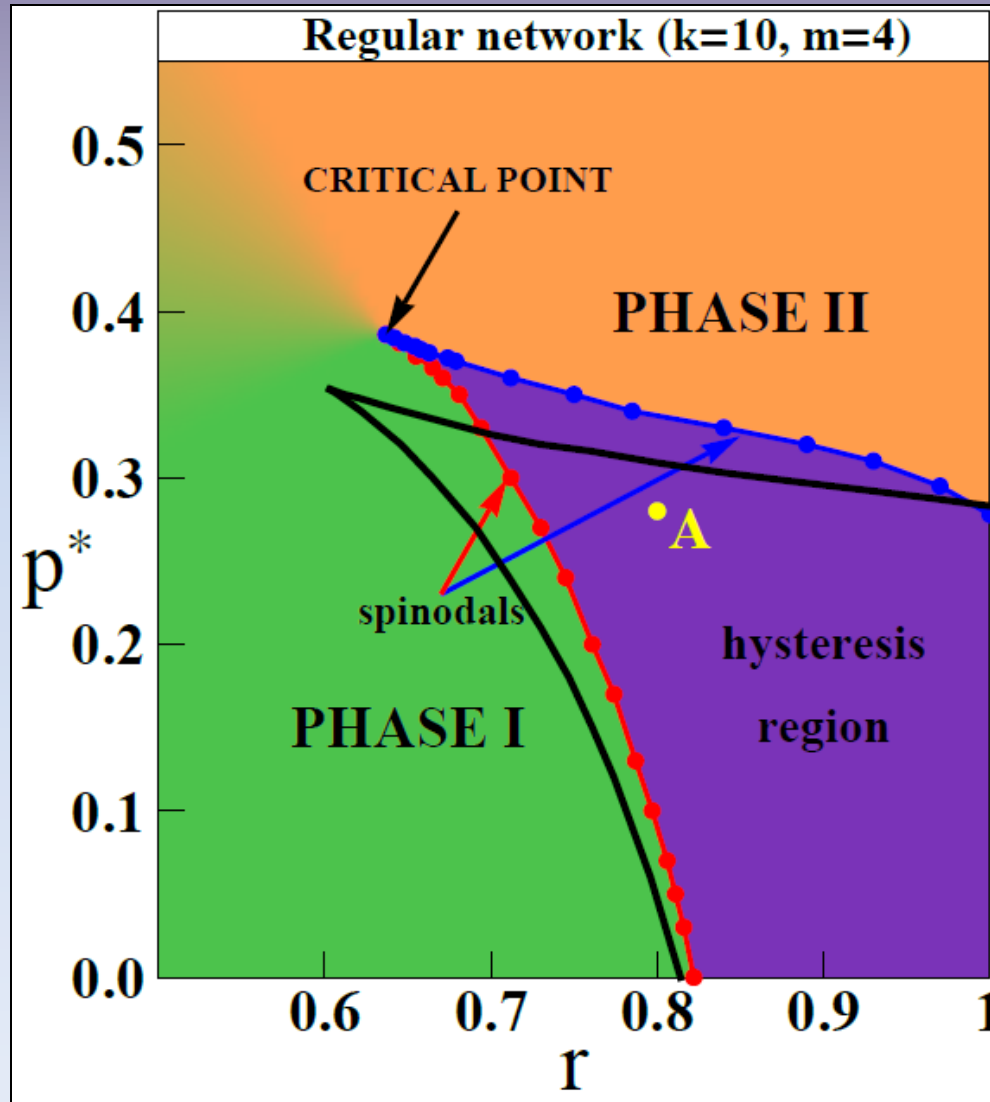
# Elements of the phase diagram

- 2 critical points
- 4 triple points
- 10 allowed transitions
- 2 forbidden transitions

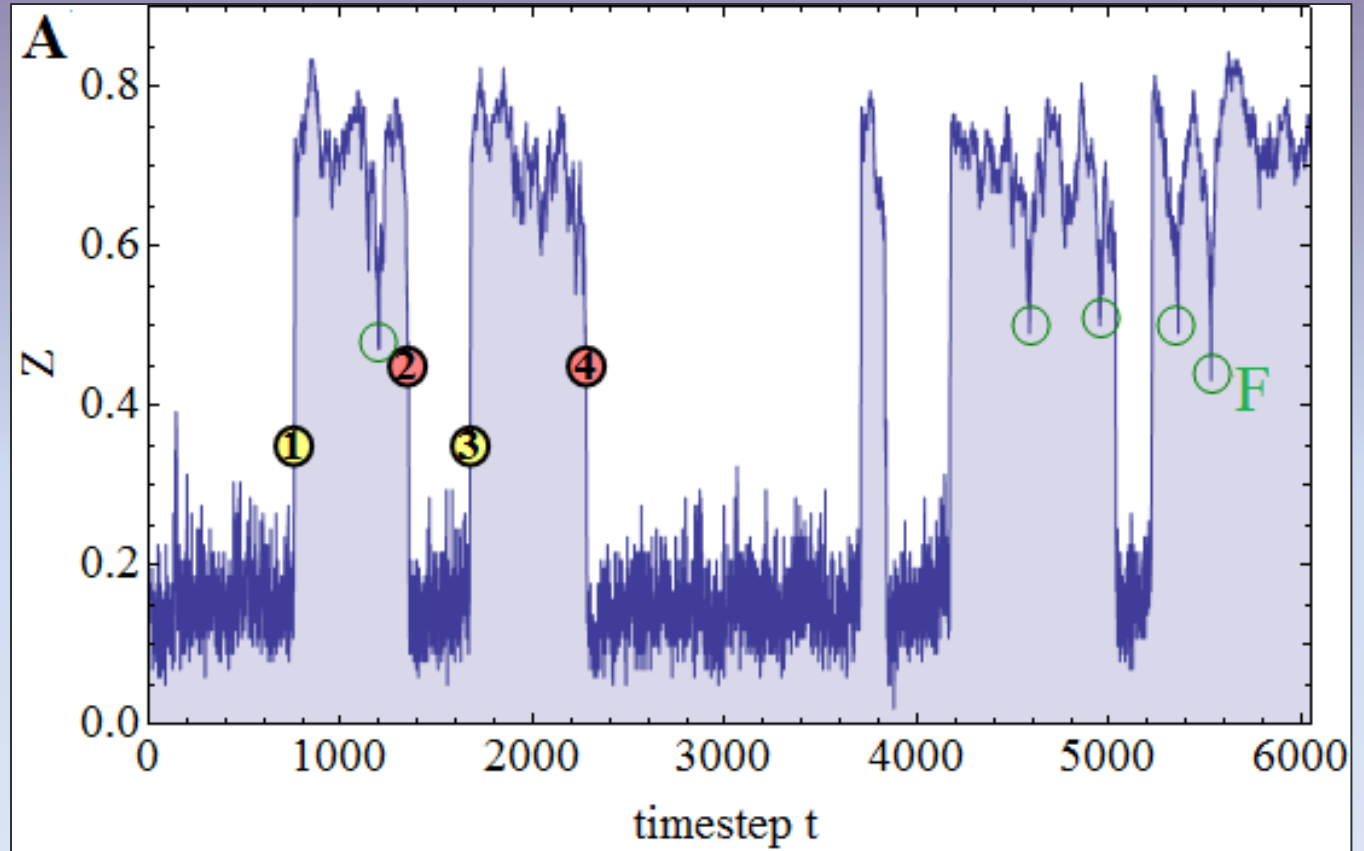


# Using network representation to model phenomena in finance

Let's pick point A, take a small system  $N=100$ , and run the simulation



# Finite size effects



Sudden transitions

Is there any  
forewarning?

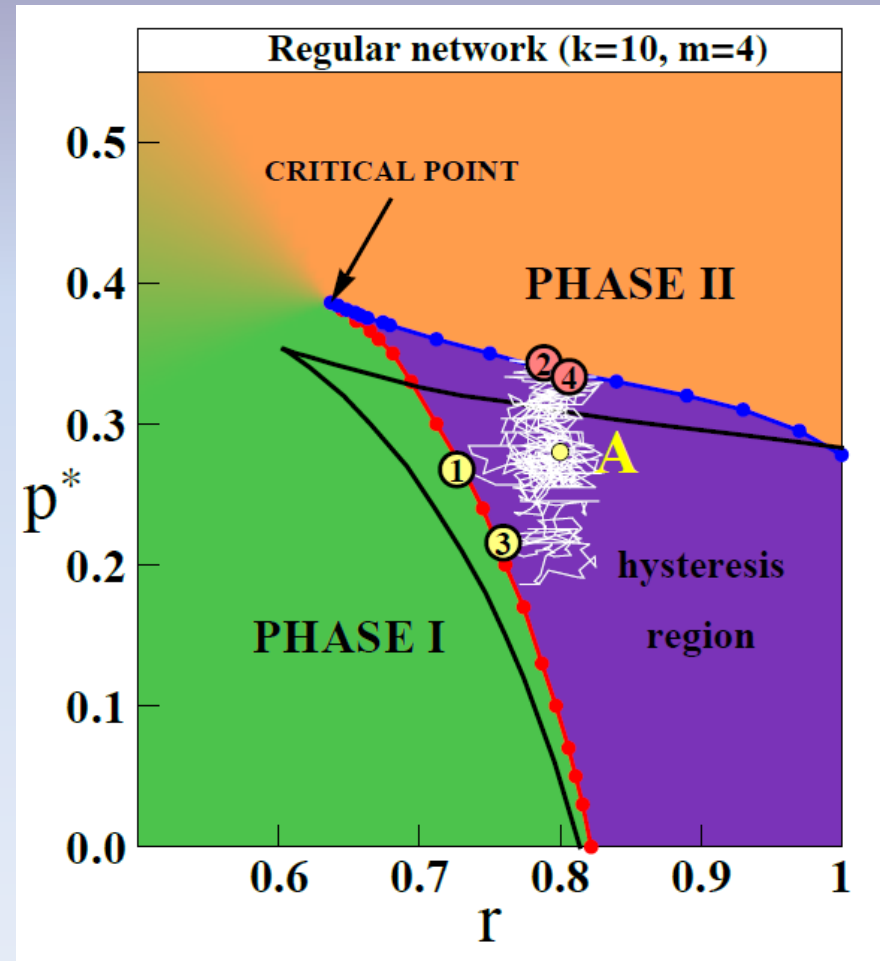
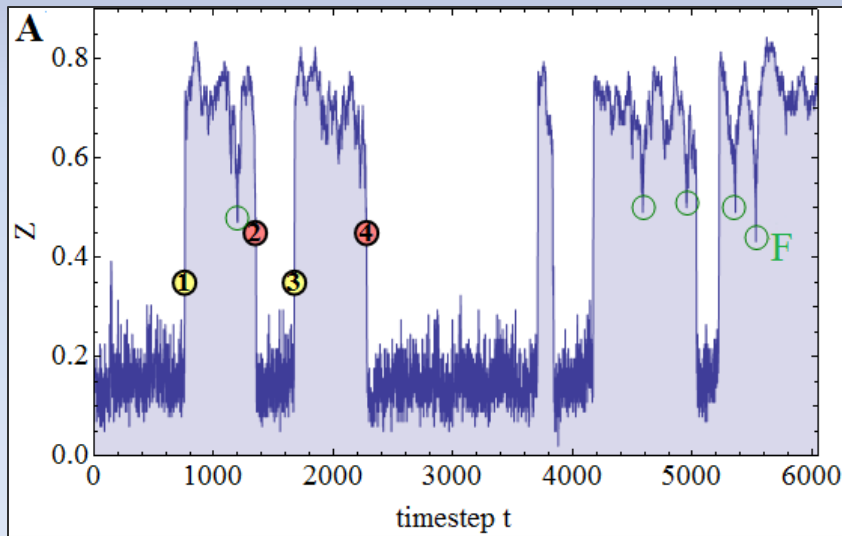
( Remember :  $Z$  = Fraction of active nodes )



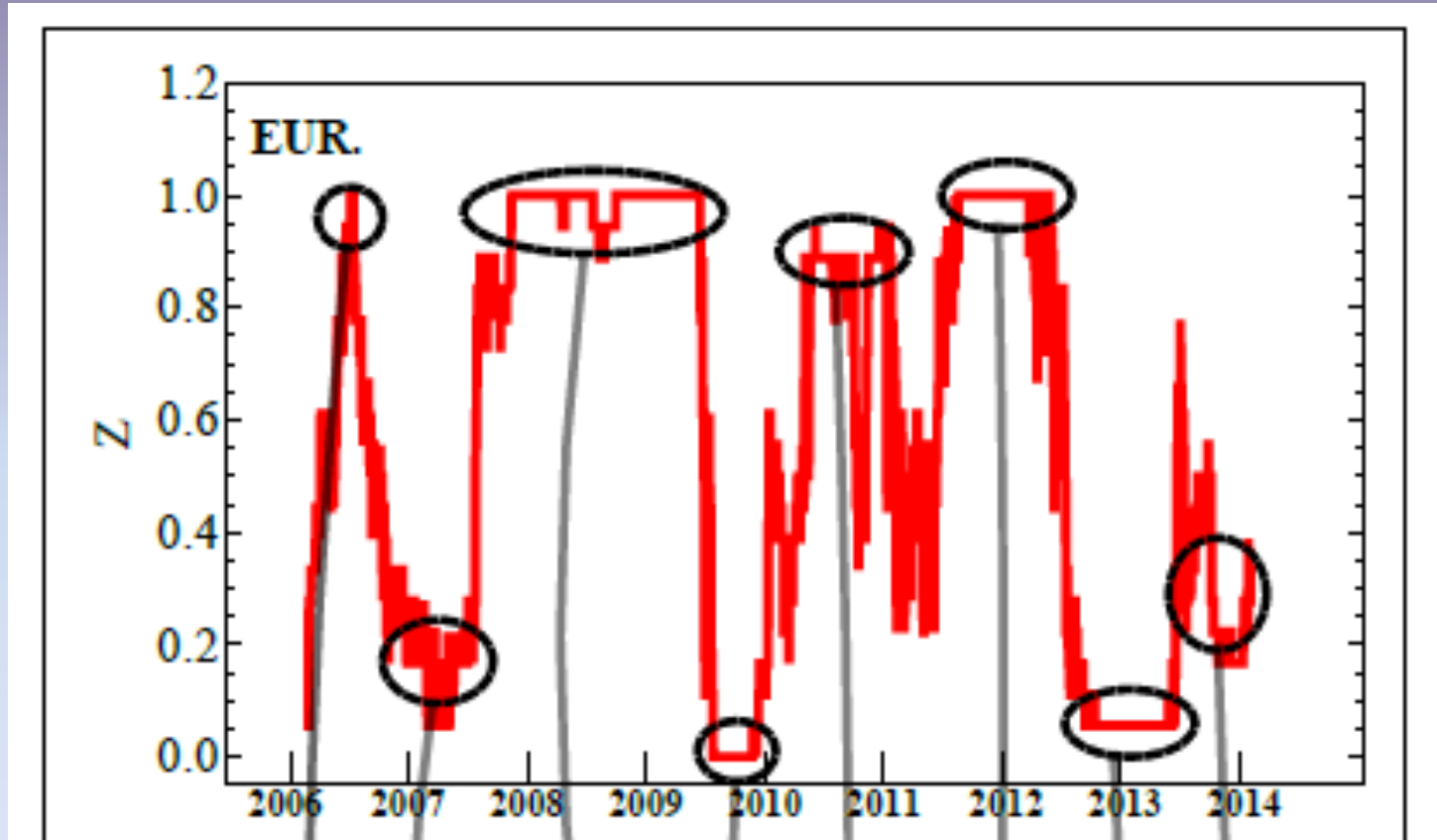
It turns out it can be predicted (to some extent).

Trajectory  $(r_\lambda(t), p_\lambda^*(t))$  in the phase diagram (white line, see below).

The trajectory crosses the spinodals (critical lines) interchangeably, and causes the phase flipping.

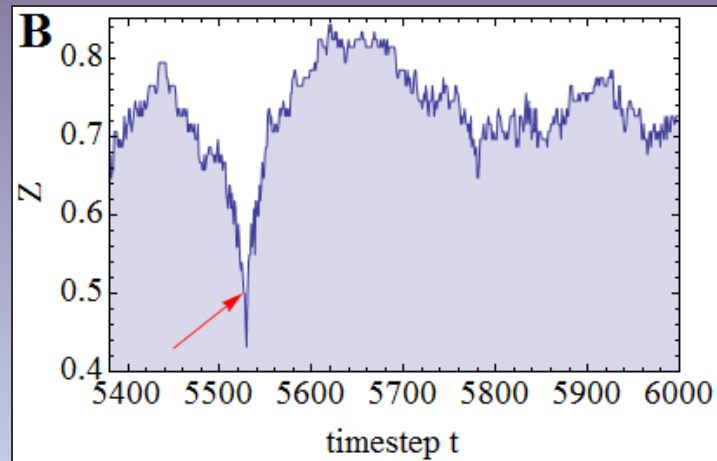


# CDS European network, in time (2006-2014)



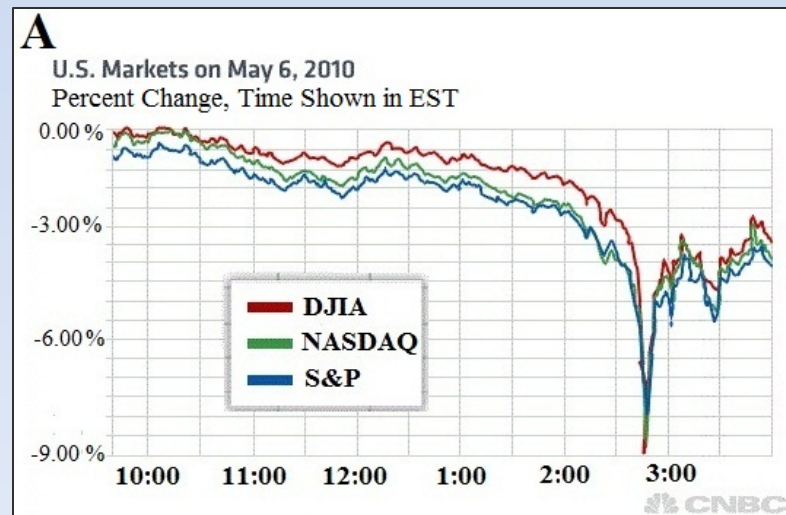
Model also predicts the existence of “flash crashes”.

Explanation: Unsuccessful transition to the lower state.



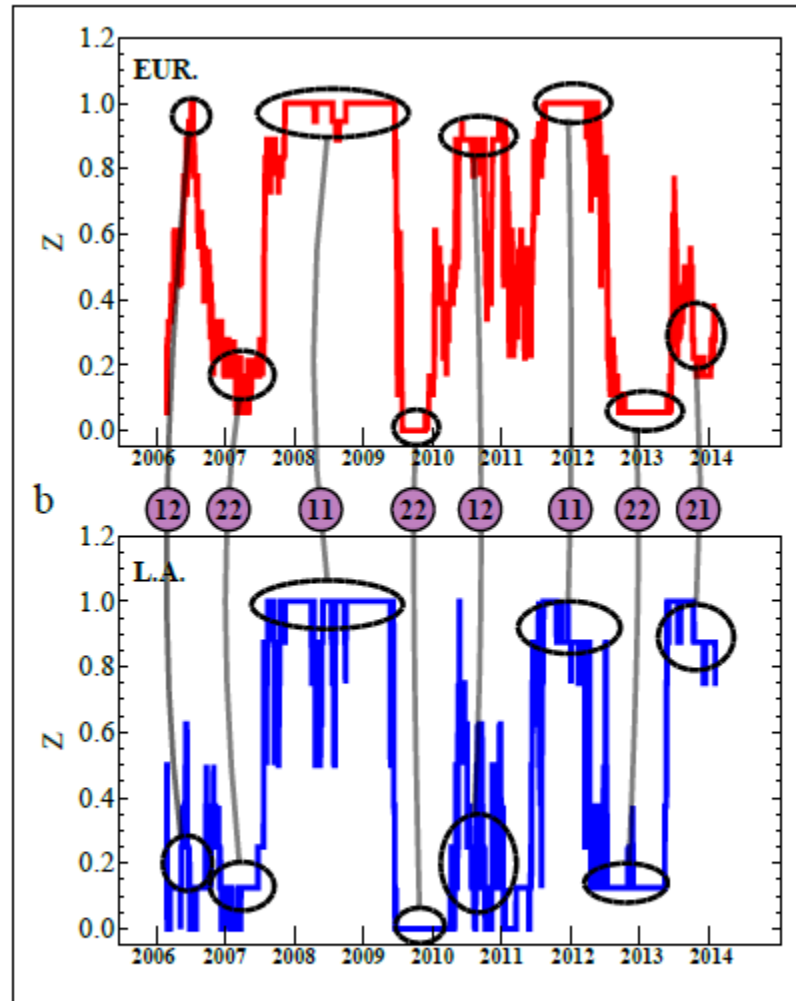
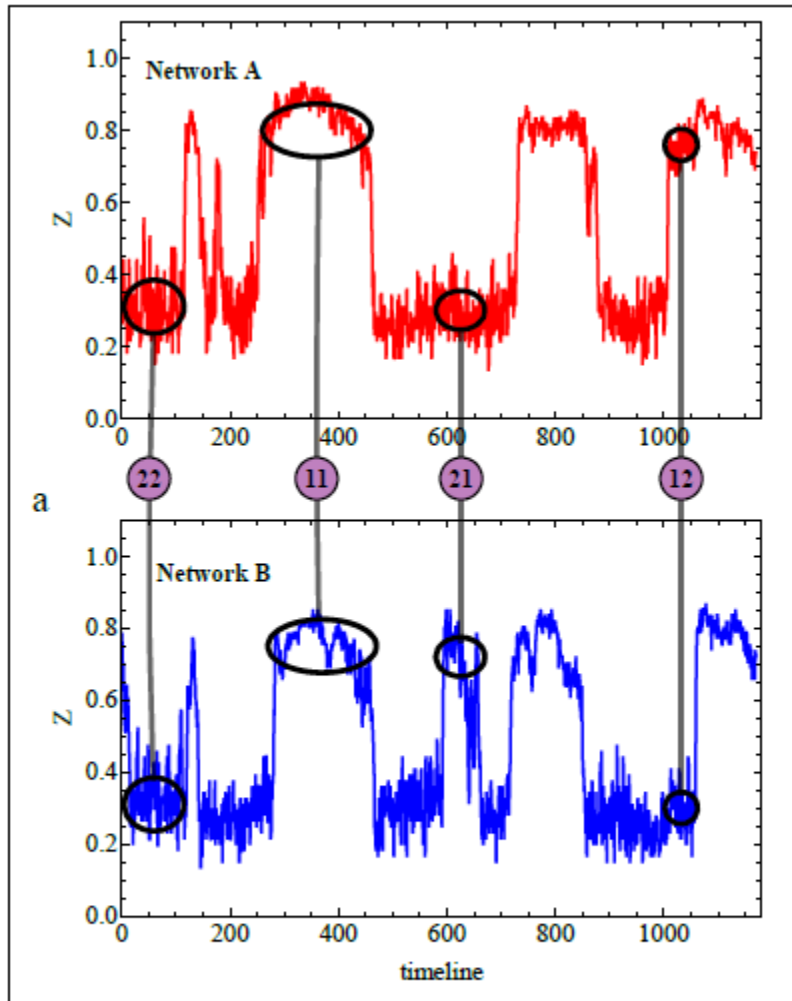
Real stock markets also show a similar phenomenon.

Q: Possible relation?



“Flash Crash 2010”

# Simulated and real interacting networks: CDS networks



6: Collective states in simulated and real interacting networks.