

# Opinion dynamics in complex networks: threshold $q$ -voter

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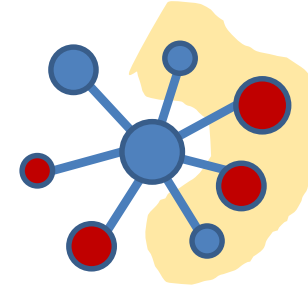
ICTP-SAIFR, São Paulo, Brazil

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# SUMMARY

- Opinion dynamics

- Usually pairwise interactions
- interactions with group of influence
- noisy version (independence)



- How order/disorder transitions are affected by

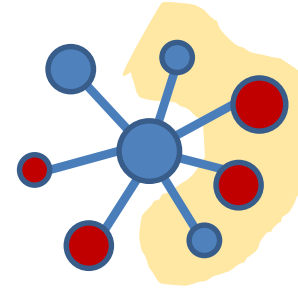
- composition of the group
- network structure

- Approach

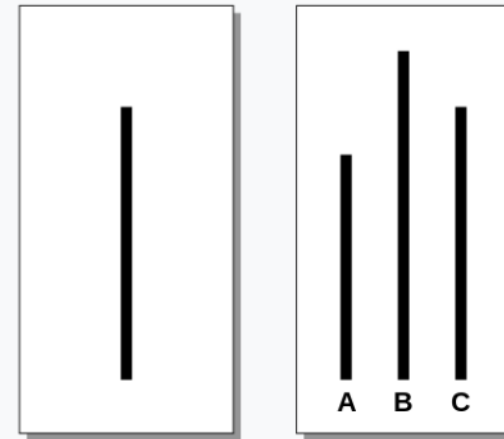
- Agent-based simulations in networks
- Theoretically: pair-approximation

# MOTIVATIONS

**Social pressure:** Individuals interacting with other individuals may change their own opinions and beliefs (even against rationality)



## Asch conformity experiments



→ Strong tendency to go along with the rest

- Asch, S. *Opinions and social pressure*, *Sci. Am.* **193**, 31–35 (1955)
- Latane, B. *The psychology of social impact*, *Am. Psychol.* **36**, 343–356 (1981)
- Cialdini, RB, Goldstein, NJ. *Social influence: conformity and compliance*, *Annu. Rev. Psychol.* **55**, 591–621 (2004)

# (NOISY) VOTER MODELS

State variable: opinion  $s_i \in \{-1, +1\}$

Opinions change through mechanisms that act stochastically:

- with probability  $1-p$

**social influence** (by interaction in conformity with other agents) occurs according to a social rule

- with probability  $p$

**idiosyncratic changes** (independently of other agents)

a random opinion is adopted

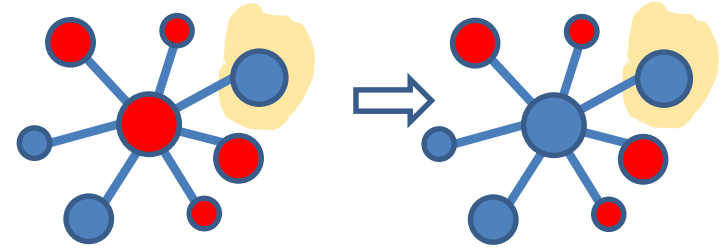
# (NOISY) VOTER MODELS: social rules

## VOTER:

a neighbor  $j$  is randomly selected

If  $j$  has opposite opinion, then  $s_i \rightarrow -s_i$ .

Liggett, Thomas M. *Annals of Probability* 25 (1997)



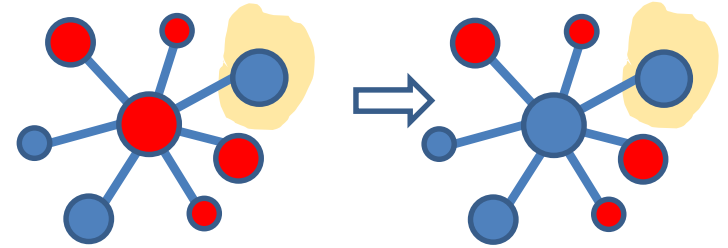
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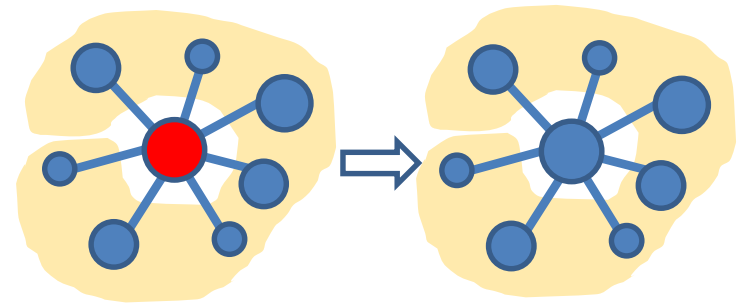


## q-VOTER:

$q$  neighbors of agent  $i$  are analyzed

If all  $q$  share the opposite opinion:  $s_i \rightarrow -s_i$ .

Castellano, Muñoz, Pastor-Satorras, *PRE* 80, 041129 (2009)



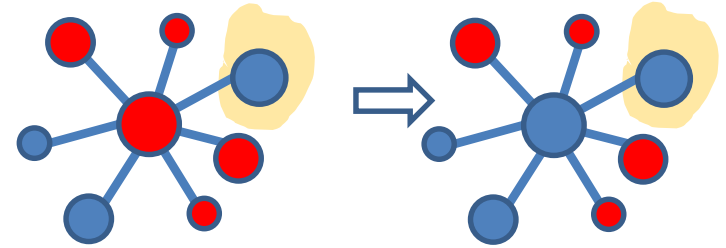
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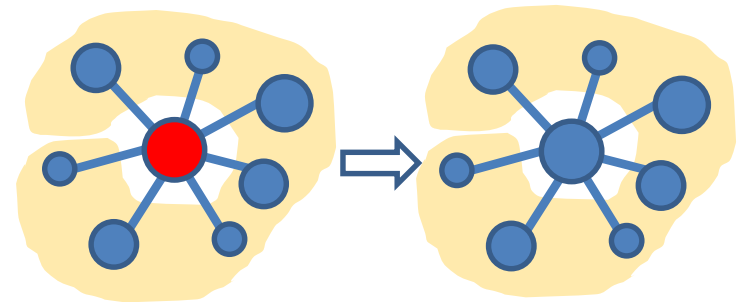


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## Threshold q-VOTER:

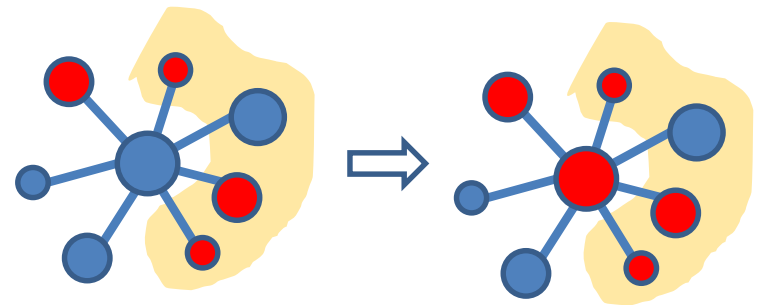
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Vieira, Anteneodo, *PRE* 97, 052106 (2018)

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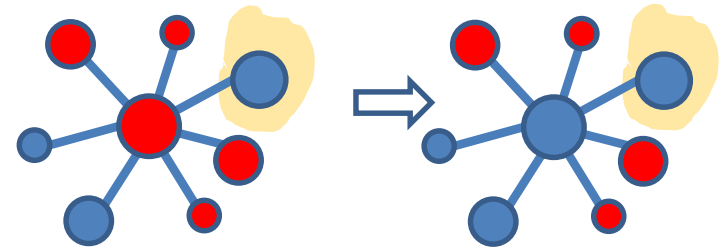
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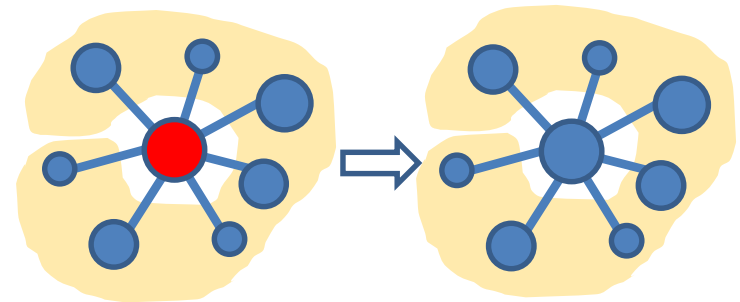


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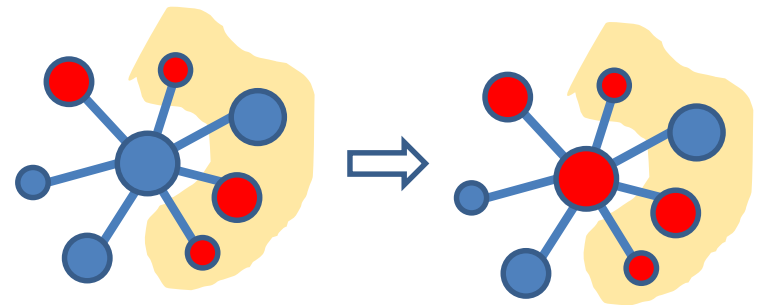
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$q_0 = q = 1$ : standard voter

$q_0 = q$ :  $q$ -voter, unanimity voter



# THRESHOLD $q$ -VOTER

## ALL-TO-ALL (MEAN-FIELD)

Transition rates, using  $n = n_+$ ,  $x = n/N$

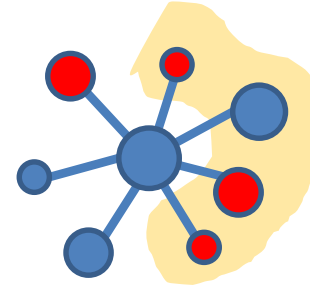
$$w(n \rightarrow n - 1) = nG(1 - x),$$

$$w(n \rightarrow n + 1) = (N - n)G(x),$$

Conditional probability that a sorted agent with opinion  $-1$  flips its opinion.

$$G(x; q, q_0, p) = (1 - p)g_1(x; q, q_0) + p/2,$$

$$g_1(x; q, q_0) = \sum_{j=q_0}^q \binom{q}{j} x^j (1 - x)^{q-j}$$



# THRESHOLD q-VOTER

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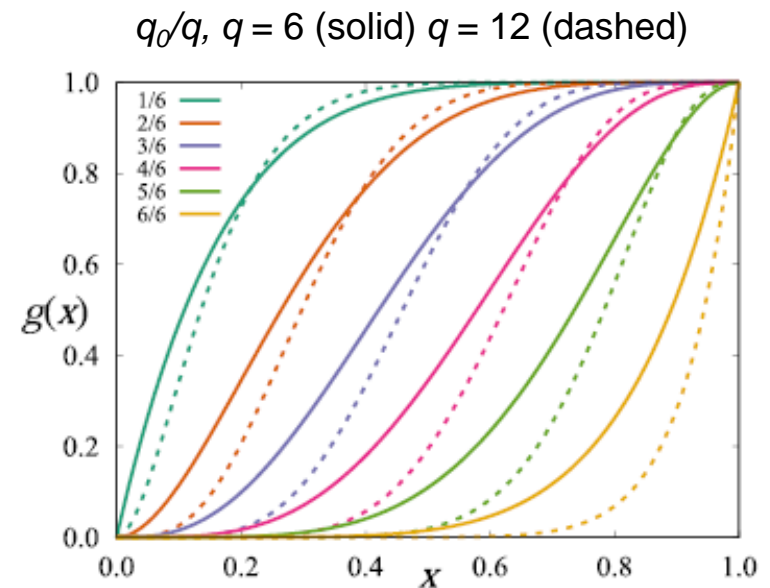
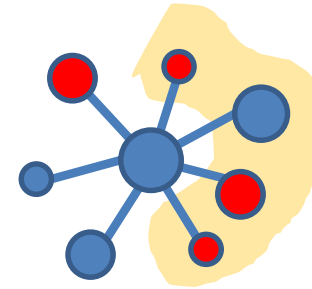
Conditional probability that a sorted agent with opinion -1 flips its opinion.

$$G(x; q, q_0, p) = (1 - p)g_1(x; q, q_0) + p/2,$$

$$g_1(x; q, q_0) = \sum_{j=q_0}^q \binom{q}{j} x^j (1 - x)^{q-j}$$

if  $q_0 = q$ :  $x^q$

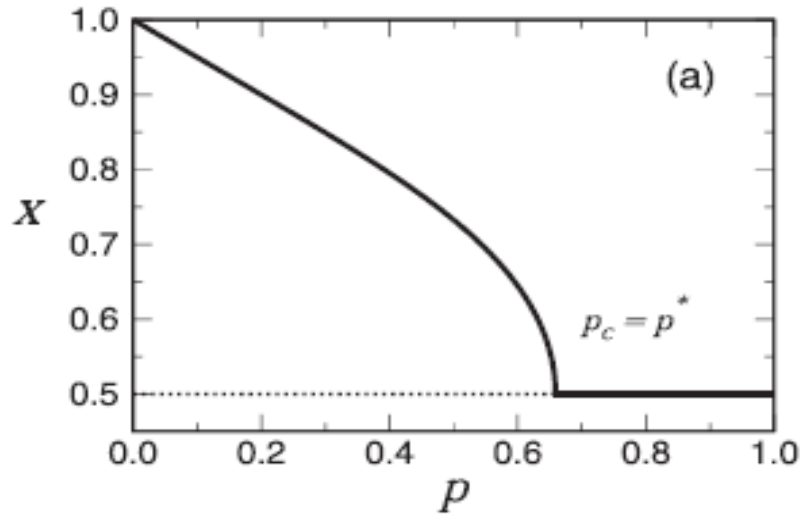
$$\frac{dx}{dt} = (1 - x)G(x) - xG(1 - x)$$



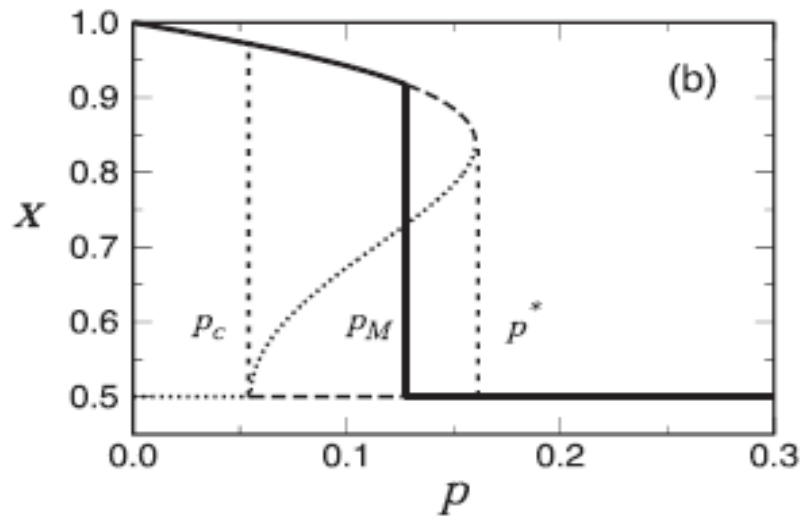
# THRESHOLD q-VOTER

## ALL-TO-ALL (MEAN-FIELD)

$$\frac{dx}{dt} = (1-x)G(x) - xG(1-x) \equiv v(x; p, q, q_0)$$



Continuous  $q_0^- < q_0 < q_0^+$

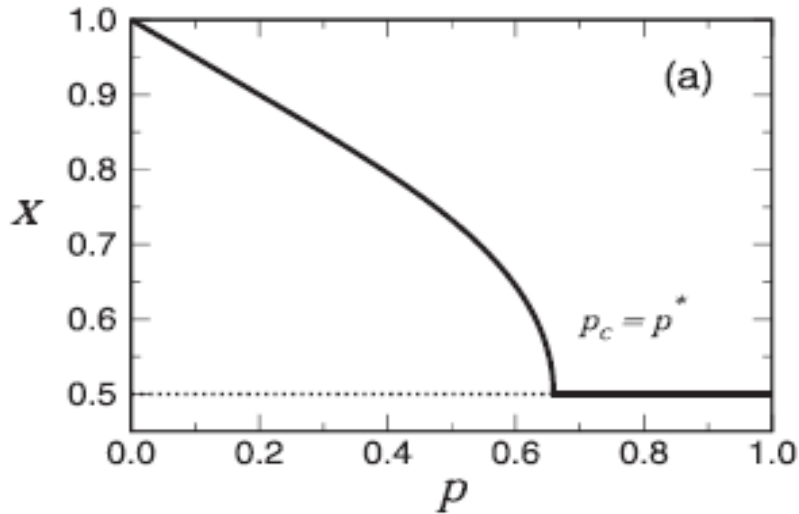


First-order  $q_0 < q_0^-$  and  $q_0 > q_0^+$

# THRESHOLD q-VOTER

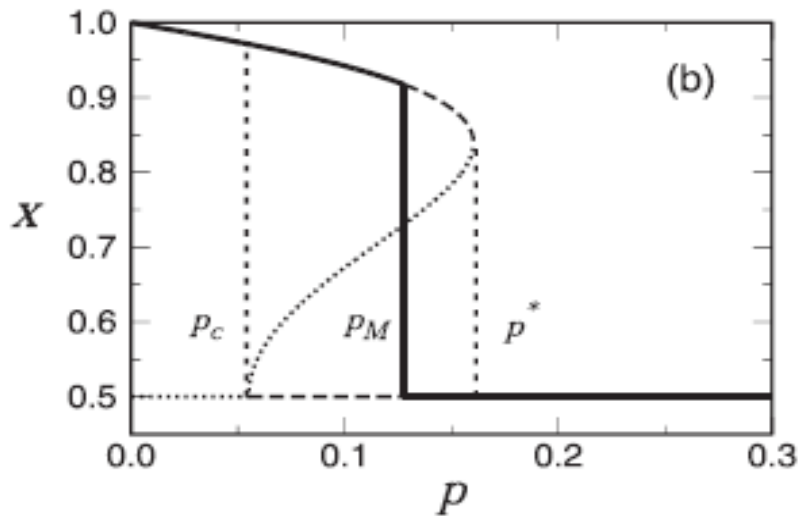
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$$p_c^{-1} = 1 + \frac{2^{q-1} \Gamma(q_0 + 1) \Gamma(q - q_0 + 1)}{\Gamma(q + 1) [q_0 - {}_2F_1(1, q_0 - q, q_0 + 1, -1)]}$$

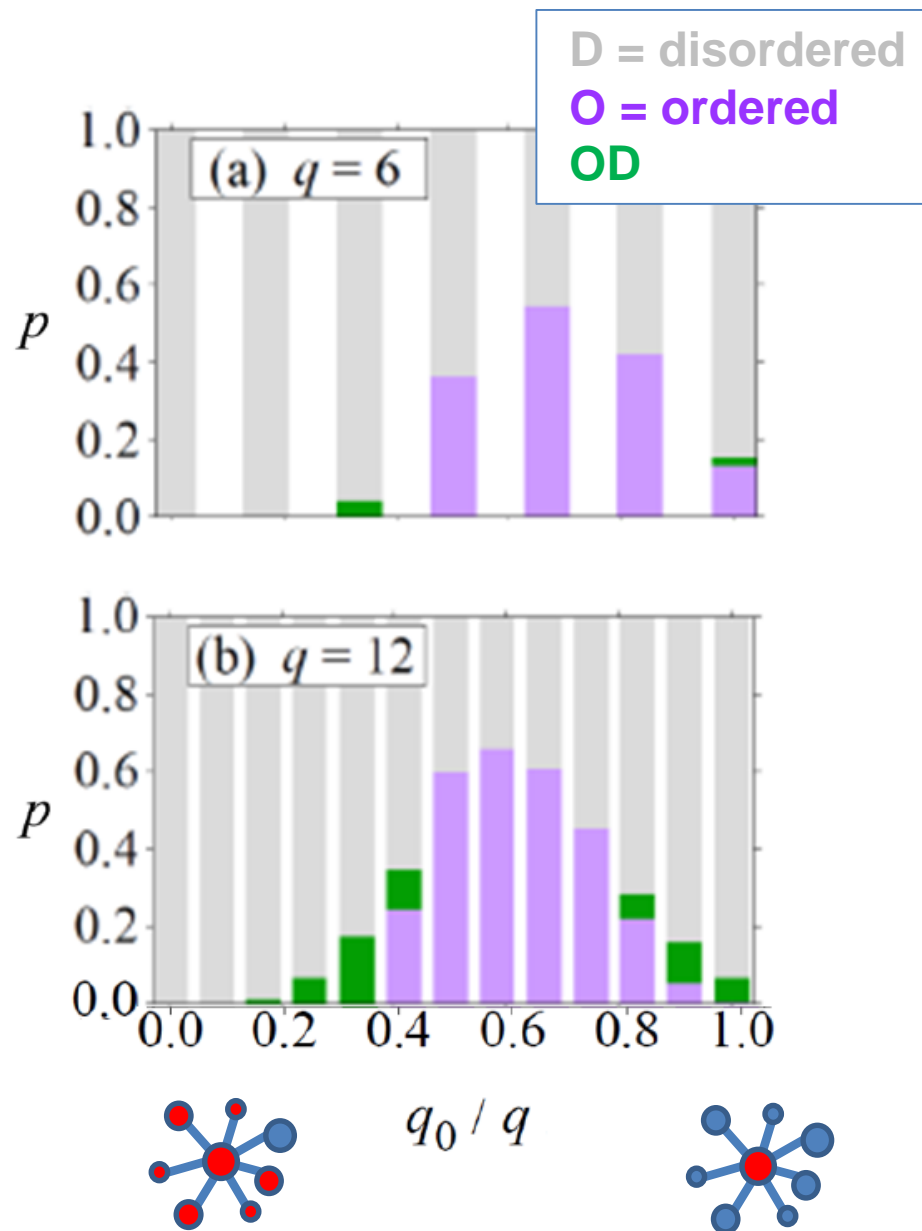
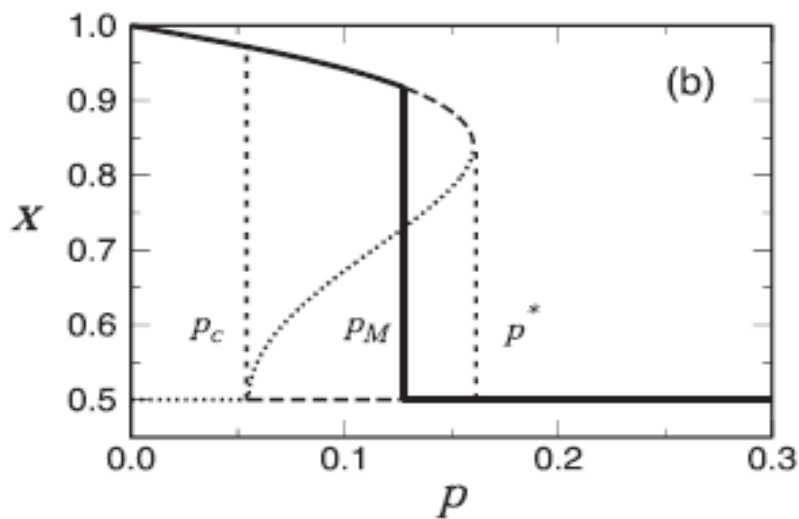
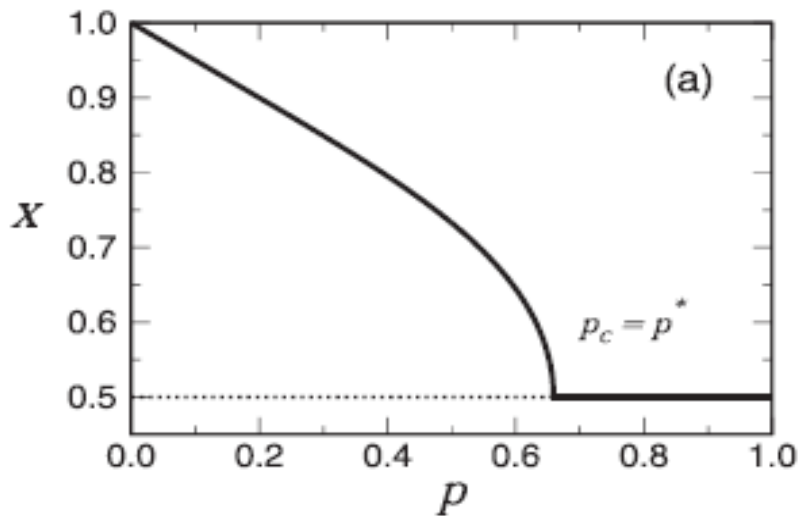


First-order  $q_0 < q_0^-$  and  $q_0 > q_0^+$

$$p_c = p^* \Rightarrow q_0^\pm(q) = \frac{1}{4}(5 + 2q \pm \sqrt{5 + 4q}).$$

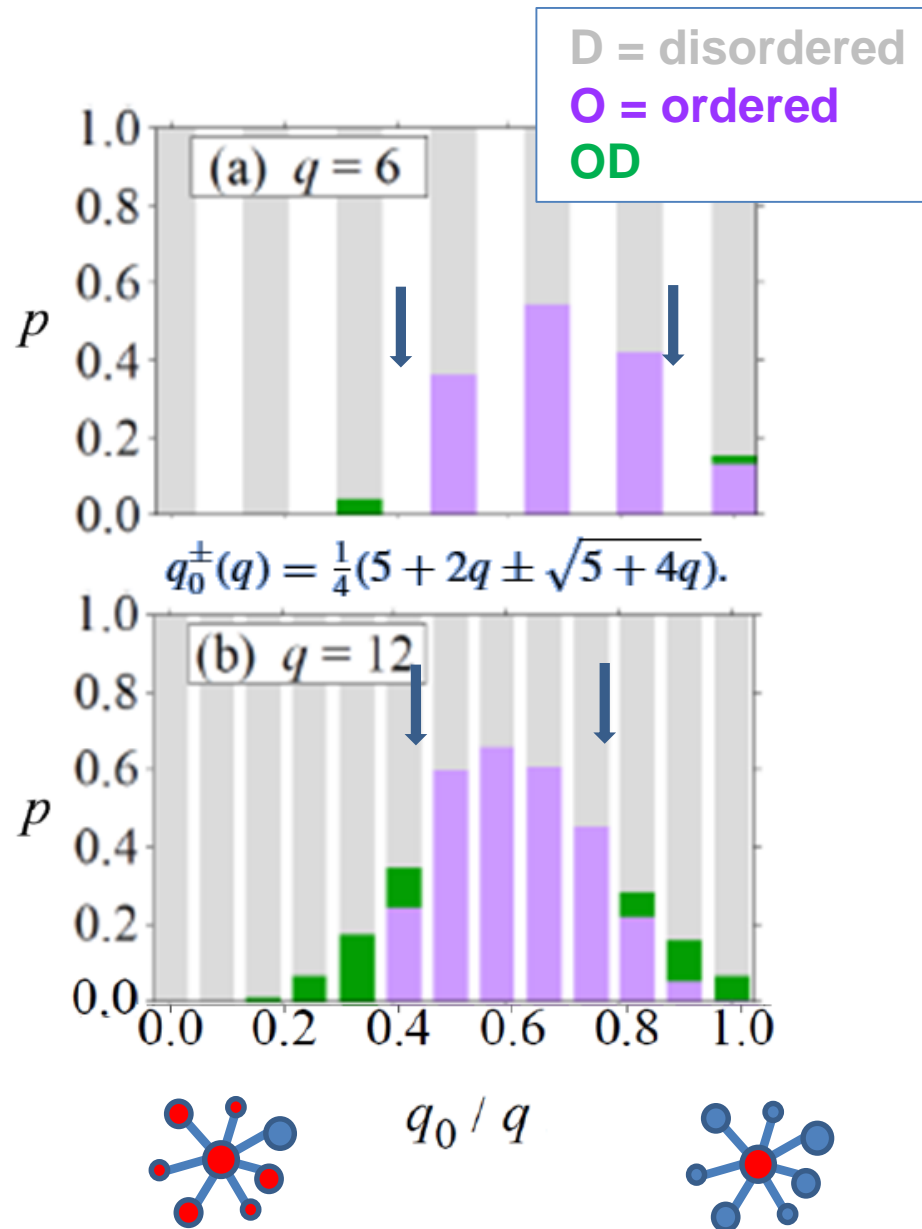
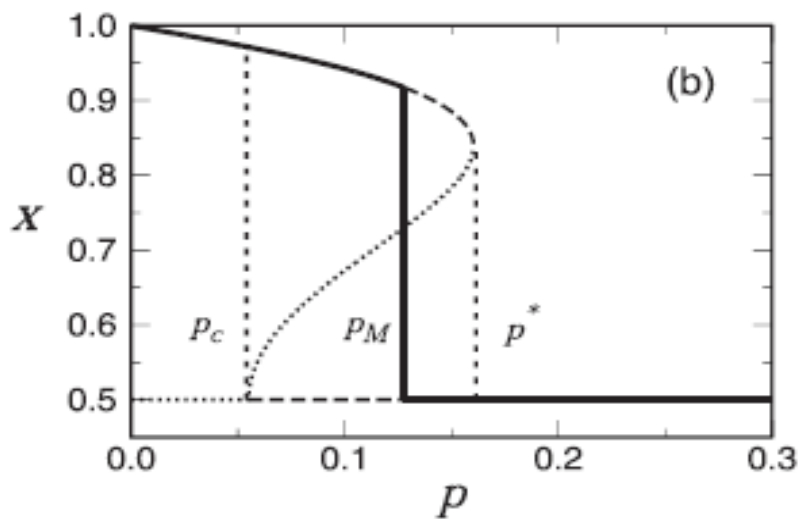
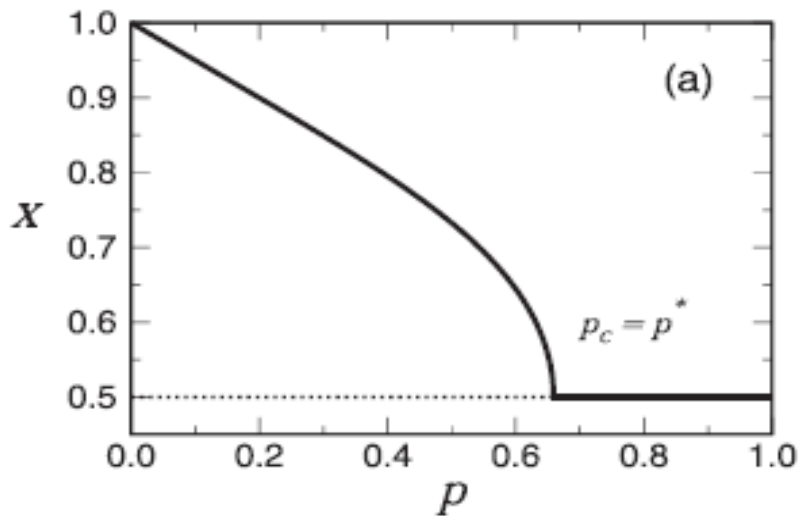
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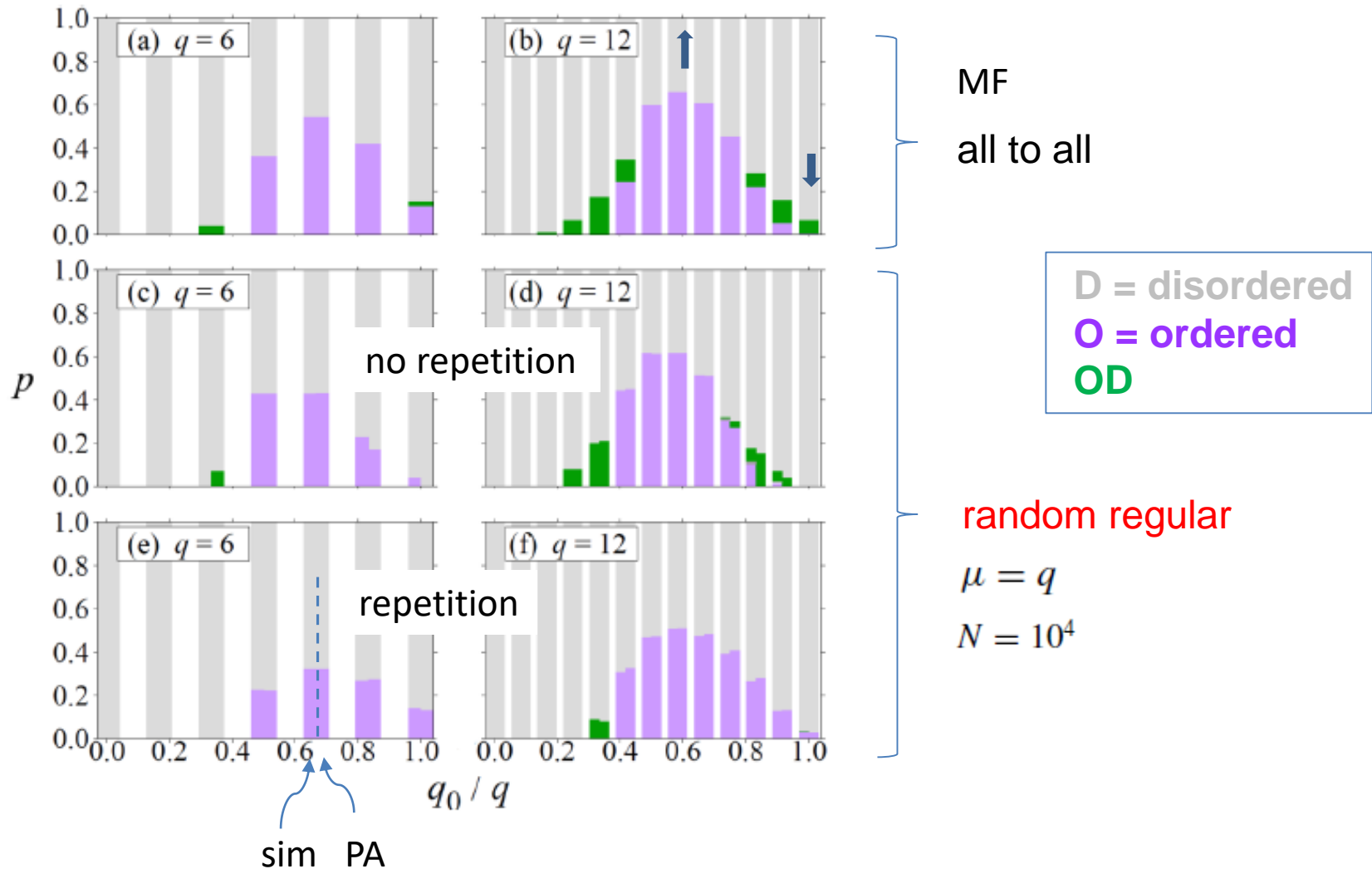
# THRESHOLD q-VOTER

## ALL-TO-ALL (MEAN-FIELD)



# THRESHOLD $q$ -VOTER

## In networks (random regular) simulations vs theory



## Pair approximation (PA)

We look at the fraction of **active links** (joining two nodes in different states)

In all-to-all connected network the fraction of active links  $\rho$  is related to the density  $x$  of nodes in the +1 state by  $\rho = 2x(1 - x)$

In a general network, we must look at  $\rho(t)$  and  $x(t)$  independently



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In a general network, we must look at  $\rho(t)$  and  $x(t)$  independently

For a chosen node that has  $k$  links amongst which  $0 \leq \ell \leq k$  are active

$$F(\ell; k, q, q_0, p) = (1 - p)f(\ell; k, q, q_0) + p/2$$

$$f(\ell; k, q, q_0) \equiv \begin{cases} \sum_{j=q_0}^q \binom{q}{j} \binom{k-q}{\ell-j} / \binom{k}{\ell}, & \text{forbidden} \\ \sum_{j=q_0}^q \binom{q}{j} \left(\frac{\ell}{k}\right)^j \left(1 - \frac{\ell}{k}\right)^{q-j} & \text{allowed} \end{cases}$$

repetition

# THRESHOLD $q$ -VOTER

## Pair approximation (PA)

Version of the PA based on the single degree distribution  $P(k)$ , expected to work well for random networks not highly clustered or correlated (otherwise a heterogeneous version of the PA)

$$\frac{d\rho}{dt} = \frac{2}{\mu} \sum_k \sum_{i=\oplus, \ominus} P(k) P_{i,k} \langle (k - 2\ell) F(\ell; k, q, q_0, p) \rangle_{\rho_i},$$

$$\frac{dx_k}{dt} = - \sum_{i=\oplus, \ominus} S_i P(k) P_{i,k} \langle F(\ell; k, q, q_0, p) \rangle_{\rho_i}$$

$$S_{\oplus} = 1, S_{\ominus} = -1 \quad \mu \equiv \sum_k P(k)k \quad \langle \dots \rangle_{\rho_i} \text{ average over } \binom{k}{\ell} \rho_i^{\ell} (1 - \rho_i)^{k-\ell}$$

$$P_{\oplus, k} = x_k, P_{\ominus, k} = 1 - x_k,$$

$$\rho_{\oplus} = P(\ominus|\oplus) = \rho / (2x_L)$$

$$\rho_{\ominus} = P(\oplus|\ominus) = \rho / [2(1 - x_L)]$$

$$\left. \begin{array}{l} \rho_{\oplus} = P(\ominus|\oplus) = \rho / (2x_L) \\ \rho_{\ominus} = P(\oplus|\ominus) = \rho / [2(1 - x_L)] \end{array} \right\} x_L = \sum_k P(k) k x_k / \mu$$

$P(i'|i)$  conditional probability of selecting a neighbor with opinion  $i'$

# THRESHOLD $q$ -VOTER

## Pair approximation (PA)

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$$P_{\oplus} = x, \quad P_{\ominus} = 1 - x,$$

$$\rho_{\oplus} = \rho/(2x), \quad \text{and} \quad \rho_{\ominus} = \rho/[2(1 - x)]$$

# THRESHOLD $q$ -VOTER

## No repetition

$$\frac{d\rho}{dt} = p(1 - 2\rho) + \frac{2(1 - p)}{\mu} [(1 - x)G_2(\rho_{\ominus}; q, q_0, \mu) + xG_2(\rho_{\oplus}; q, q_0, \mu)]$$

$$\frac{dx}{dt} = (1 - x)G(\rho_{\ominus}; q, q_0, p) - xG(\rho_{\oplus}; q, q_0, p)$$



$$p_c^{-1} = 1 + \frac{2^{q-1} \left(\frac{\mu}{\mu-2}\right)^{q_0} \left(\frac{\mu-1}{\mu}\right)^q \Gamma(q_0 + 1) \Gamma(q - q_0 + 1)}{\Gamma(q + 1) [q_0 - {}_2F_1(1, q_0 - q, q_0 + 1, 2/\mu - 1)]},$$

$$q_0^{\pm}(q, \mu) \approx \frac{1}{2} \left\{ 6 + (2q - 7)\rho_c \pm \sqrt{16 + \rho_c[-44 + 4q(1 - \rho_c) + 29\rho_c]} \right\}$$

$\mu/q \rightarrow \infty, p_c \rightarrow$  mean-field values

# THRESHOLD q-VOTER

## With repetitions

For this modality, we did not manage to find general closed expressions for the averages  $\langle \dots \rangle_{\rho_i}$  over the binomial.

Besides the mean degree  $\mu$ , also  $\langle k^{-m} \rangle$ , with  $1 \leq m \leq q - 1$ , contribute.

## For more details

Vieira, Peralta, Toral, Anteneodo, *Phys Rev E* 101, 052131 (2020)

Recent work in q-voter:

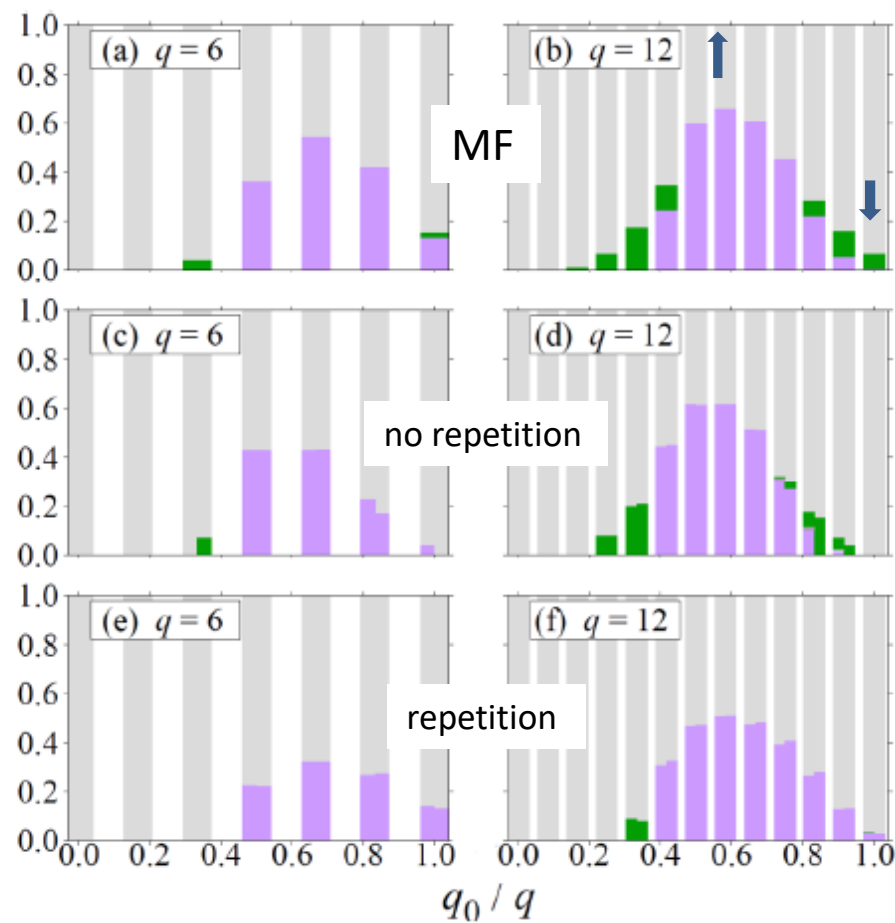
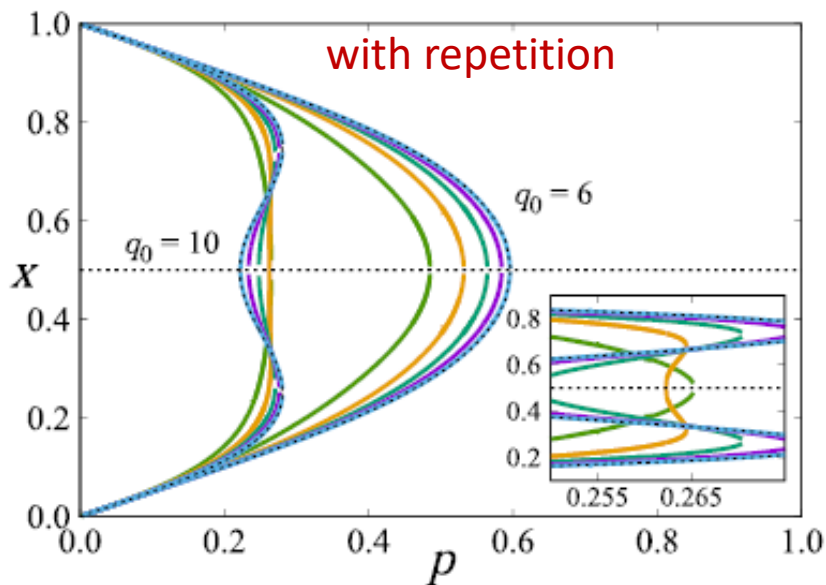
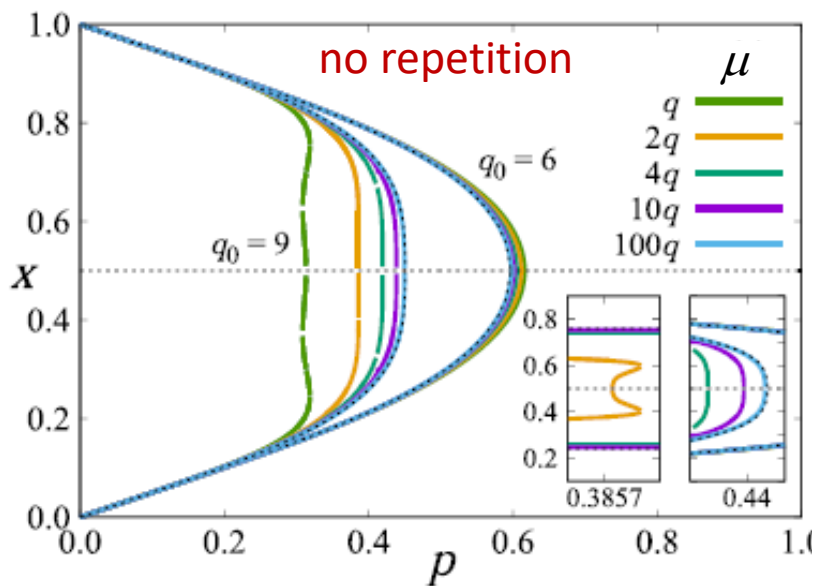
Weron, Nyczka, Szwabinski, *Entropy* 26, 132 (2024)

- PA the most accurate vs (network-aware and heuristic)
- hMFA: without repetition

# THRESHOLD $q$ -VOTER

## Comparisons

random regular  $P(k) = \delta(k - \mu)$   $\mu = q = 12$



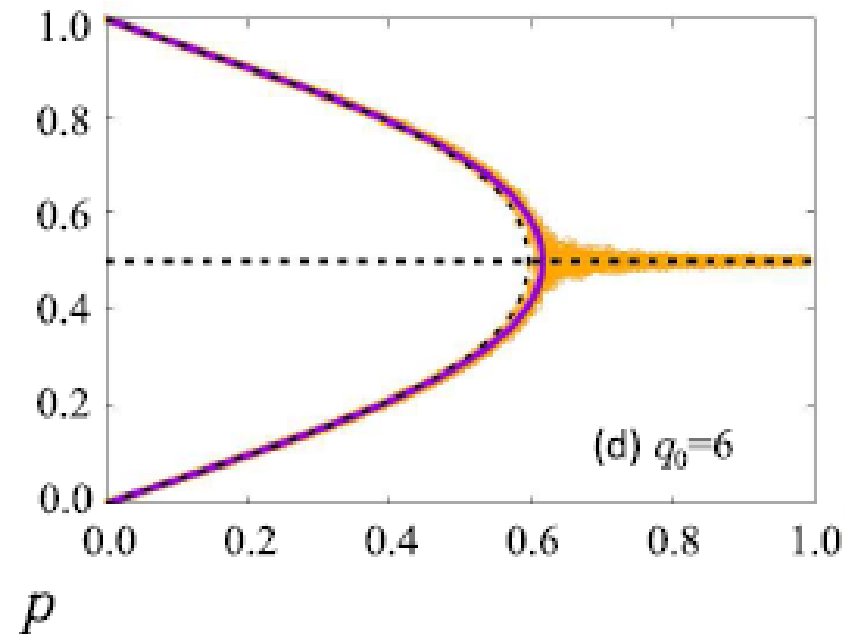
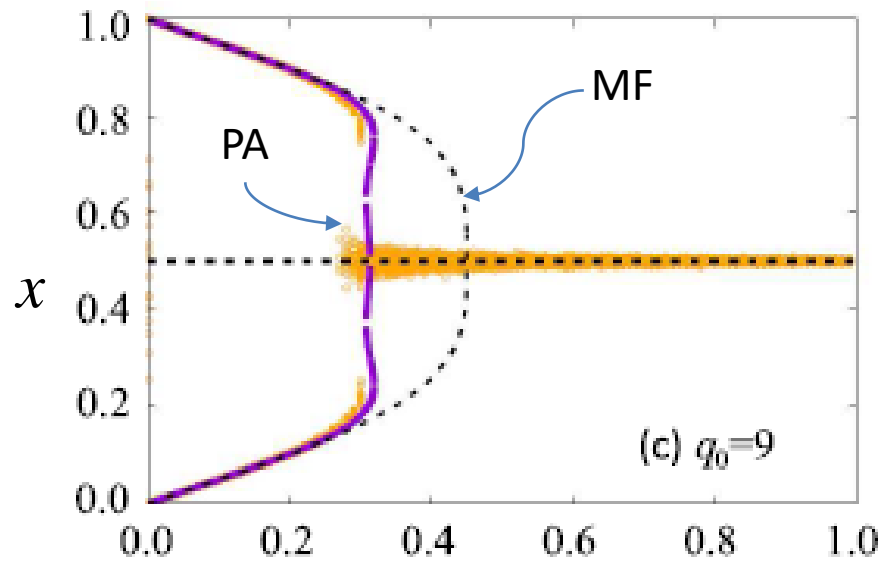
# THRESHOLD $q$ -VOTER

## NETWORKS (simulations vs theory)

RR, no-repetitions

$$P(k) = \delta(k - \mu)$$

$$\mu = q = 12$$



In the numerical simulations, we did not detect any significant discrepancy between random regular, Erdős-Rényi, and power-law networks, with the same  $\mu$ , in agreement with PA predictions.

# THRESHOLD $q$ -VOTER

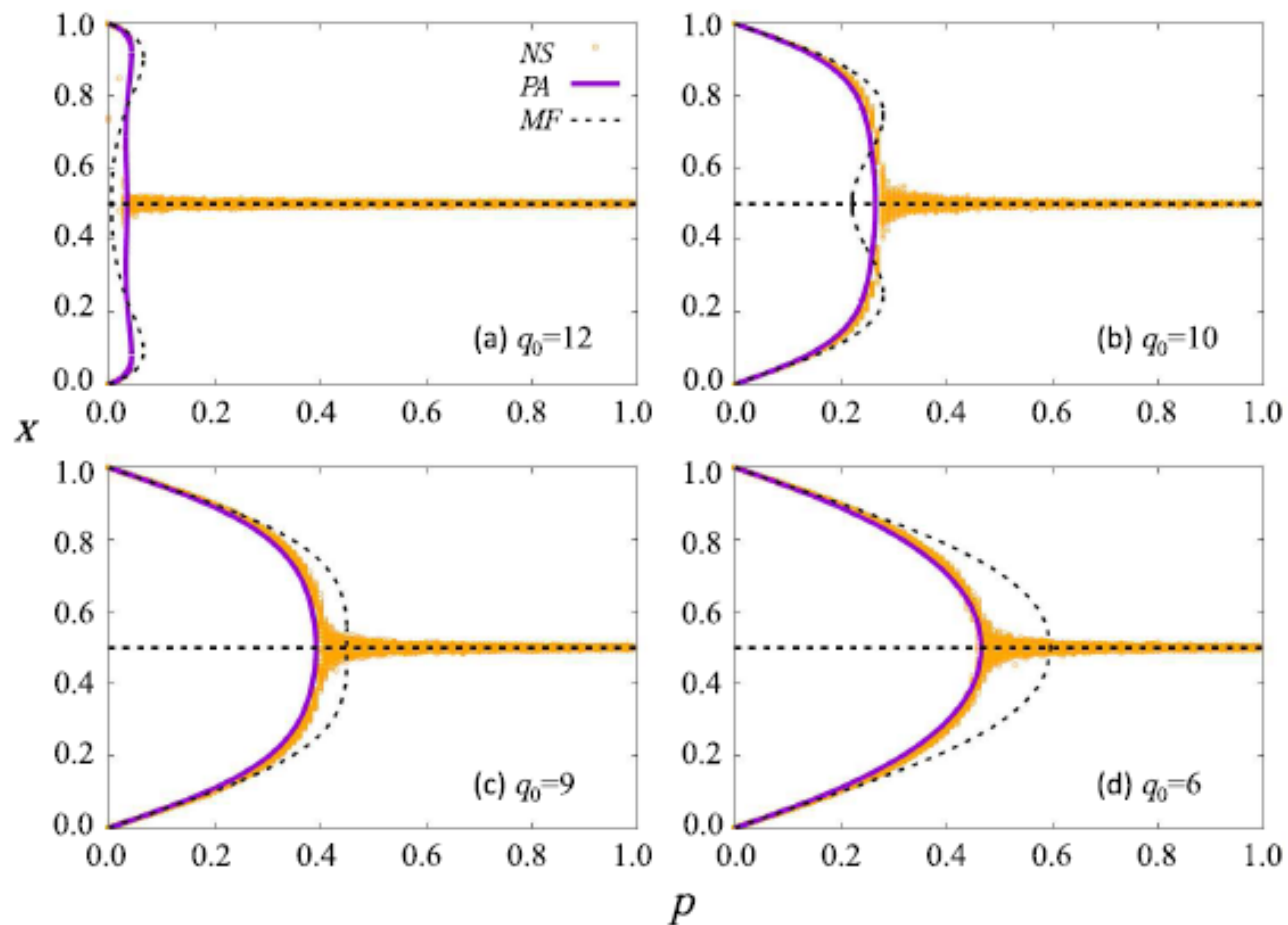
## NETWORKS (simulations vs theory)

RR, repetitions

$$P(k) = \delta(k - \mu)$$

$$\mu = q = 12$$

$$C \sim (\mu - 1)^2 / (N\mu)$$



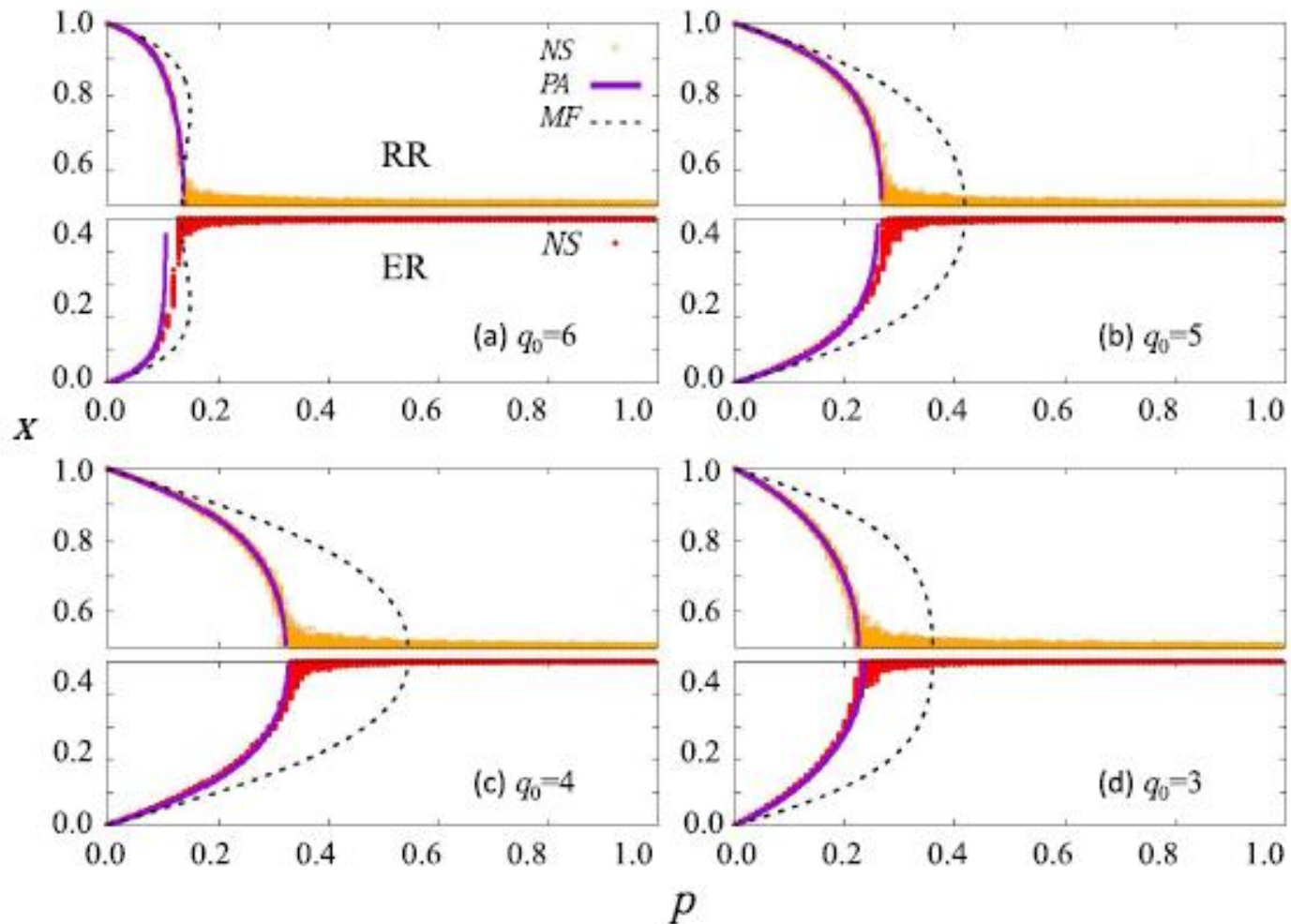


# THRESHOLD $q$ -VOTER

## NETWORKS (simulations vs theory)

RR and ER, repetitions

$$\mu = q = 6$$



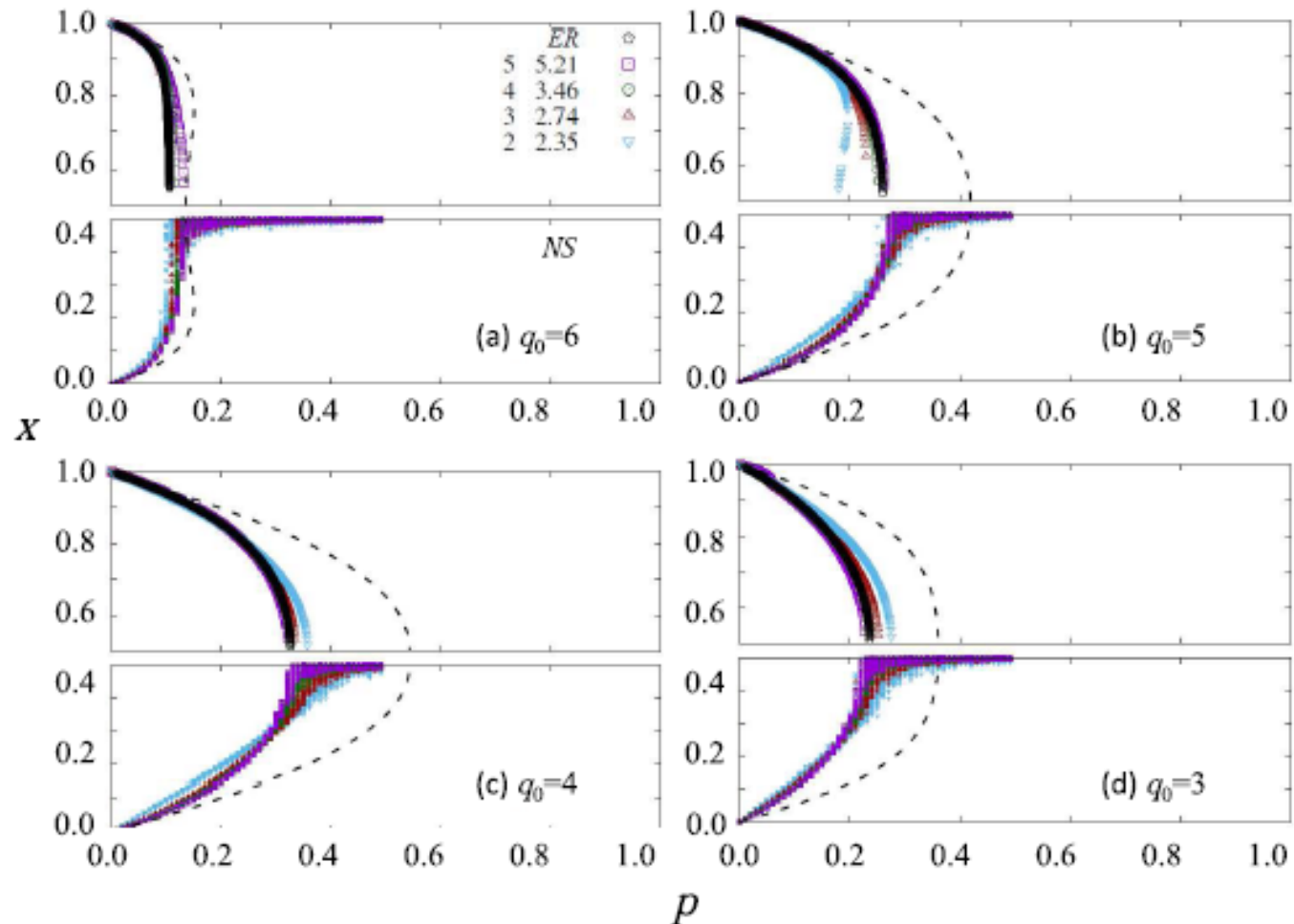
# THRESHOLD $q$ -VOTER

## NETWORKS (simulations vs theory)

ER and power-law, repetitions

$$\mu = q = 6$$

$$P(k) = \mathcal{N}/k^a, \text{ for } k_{\min} \leq k \leq k_{\max} = N/5$$



# CONCLUSIONS

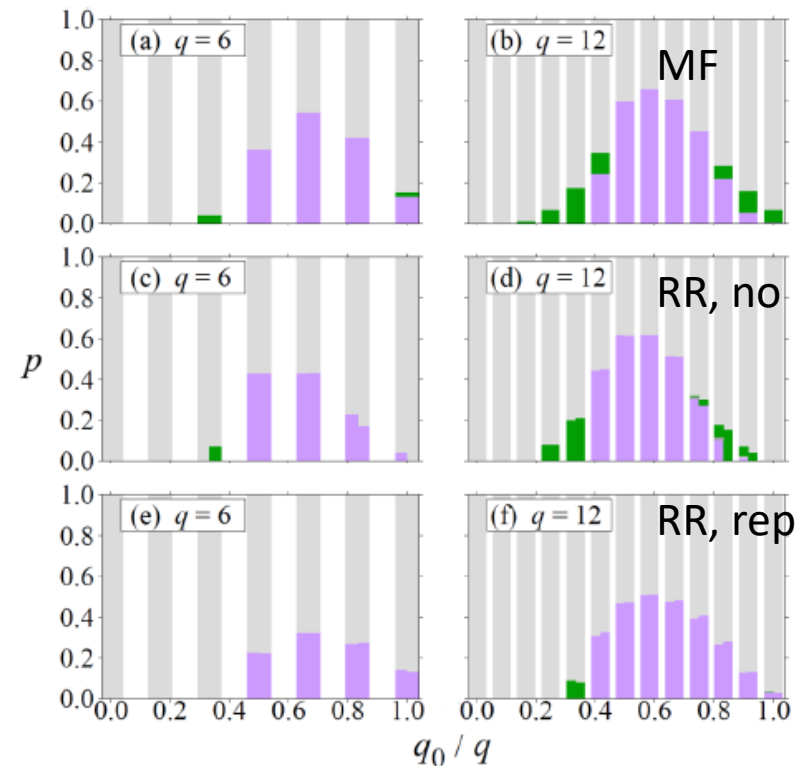
For **noisy threshold  $q$ -voter** model, possibility of **repetitions** or not in the selection of  $q$  amongst  $k$  neighbors.

The threshold  $q_0$  influences the nature of transitions. Optimal value for consensus.

The structure has a **stronger influence** in the case **with repetition**, where the **discontinuous transitions are less common** than in fully connected networks.

Differently to the case without repetitions, with repetitions results depend not only on  $m$  ( $\rightarrow$  stronger effects for long-tails)

Analytical results were derived using the **pair approximation**, for random networks with arbitrary degree distribution  $P(k)$ .



# CONCLUSIONS

With regard to the performance of the **pair approximation**:

The critical points estimated through the PA are in good agreement with simulations in **random regular** networks.

This is especially true when  $\mu$  increases approaching the exact MF result where repetition and other issues related to structure become irrelevant.

*Without repetitions*, spurious results, such as multistability beyond three states, are observed in cases with  $q_0 \approx q \approx \mu$ .

For **ER and power-law** networks, good predictions, specially *without repetitions*, but limited structures can be visited.

*With repetitions*, the PA predicts dependency on network structure beyond the average degree  $\mu$ , in accord with simulations. These effects are weak in ER networks, but stronger in networks with power-law degree distribution where long tails are concomitant with high probability of poorly connected nodes to realize small values of  $\mu$ , far away from the MF. Also in this case  $q_0/q$  plays a crucial role.

Deviations are expected in networks with higher correlations (large clustering coefficient.)

# PERSPECTIVES

- Connection with experiments (compare, suggest)
- Times to consensus
- Theory for no-repetition (heuristic-MFA)
- Aging effects

# REFERENCES

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Vieira, Anteneodo, *Phys Rev E* **97**, 052106 (2018)

## **Pair approximation for the noisy threshold $q$ -voter model**

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## **Nonlinear $q$ -voter model**

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## **Anticonformity or independence? Insights from statistical physics**

Nyczka, Sznajd-Weron, *J Stat Phys* **151**, 174 (2013)

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## **Composition of the Influence Group in the $q$ -Voter Model and Its Impact on the Dynamics of Opinions**

Weron, Nyczka, Szwabinski, *Entropy* **26**, 132 (2024)