

Transition Trajectory for Equilibrium Droplet Formation

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*Computation of Transition Trajectories and Rare Events in
Non-Equilibrium Systems*

Centre Blaise Pascal, ENS de Lyon, 13 June 2012



Outline

- 1 Motivation
- 2 Theory
- 3 Monte Carlo (MC) Simulations
 - Square lattice NN Ising model
 - Triangular lattice Ising model
 - Square lattice NNN Ising model



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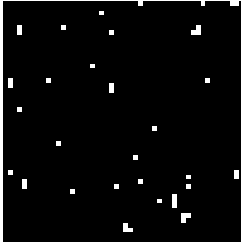


Outline

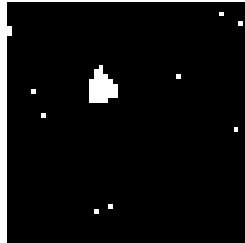
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Evaporated



Condensed

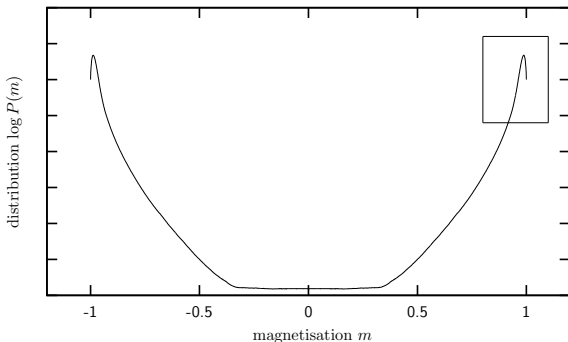


Balancing fluctuations vs interface free energy, i.e.,
entropy vs energy



Droplet formation: nucleation of “wrong” phase

- fluid droplet in gas phase, or
- “-” Ising droplet in “+” phase



Fisher; Binder & Kalos; Furukawa & Binder; Pleimling & Selke; Neuhaus & Hager; ...
 Biskup, L. Chayes & Kotecký, *Europhys. Lett.* **60** (2002) 21; *Comm. Math. Phys.* **242**
 (2003) 137

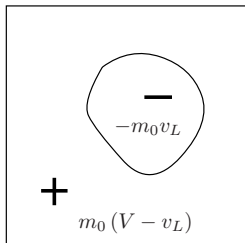
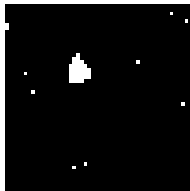


Theory: Equilibrium Droplet Formation

2D Ising model formulation

lattice gas:

- spin up = black = vacancy
- spin down = white = particle



$$M = -m_0 \underbrace{v_L}_{\text{droplet}} + m_0 \underbrace{(V - v_L)}_{\text{background}}$$

$$\Rightarrow \delta M \equiv M - M_0 = -2v_L m_0$$



Gaussian fluctuations around peak:

$$\exp \left[-\frac{(\delta M)^2}{2V\chi} \right] = \exp \left[-\frac{(2m_0 v_L)^2}{2V\chi} \right]$$

$$\chi = \chi(\beta) = \beta V [\langle m^2 \rangle - \langle m \rangle^2] = \text{susceptibility}$$

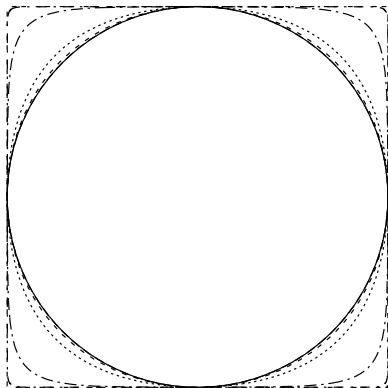
Interface free energy of droplet:

$$\exp \left[-\tau_W \sqrt{v_L} \right]$$

$\tau_W = \tau_W(\beta)$ = interfacial free energy per unit volume of optimal Wulff shaped droplet



2D Ising model Wulff shapes at various temperatures:



$T = T_c$	———
$T = 2.000$	-----
$T = 1.500$	- - - - -
$T = 1.000$
$T = 0.300$	- - - - -
$T = 0.050$	-----
$T = 0.005$	-----

\Rightarrow for $1.0 \lesssim T \leq T_c \approx 2.27$ the Wulff shape is **almost isotropic**



Balancing the exponents of the two limiting cases:

$$\Delta = \frac{(2m_0 v_L)^2 / (2V\chi)}{\tau_W \sqrt{v_L}} = 2 \frac{m_0^2 v_L^{3/2}}{\chi \tau_W V}$$

Terms are equally important for $\Delta \stackrel{!}{=} 1$:

$$\Rightarrow v_L \Rightarrow -\delta M = \theta V^{2/3} \quad \text{with} \quad \theta = \left(\frac{2\chi\tau_W}{\sqrt{2m_0}} \right)^{2/3}$$

- $-\delta M \gg \theta V^{2/3}$: **droplet** dominates
- $-\delta M \ll \theta V^{2/3}$: **fluctuations** dominate

“Isoperimetric reasoning” (Biskup *et al.*) shows that either of these two cases dominate – but **no droplets of intermediate size** can exist.



In general, a single large droplet of size v_d coexists with small fluctuations taking $v_L - v_d$ of the total excess.

- large droplet costs $e^{-\tau_W \sqrt{v_d}}$, absorbs fraction $\delta M_d = -2v_d m_0$ of δM
- fluctuations cost $e^{-(\delta M - \delta M_d)^2 / (2V\chi)}$

For large systems, probability for magnetization excess:

$$e^{-\tau_W \sqrt{v_d} - \frac{(\delta M - \delta M_d)^2}{2V\chi}} = e^{-\tau_W \sqrt{\frac{-\delta M}{2m_0}} \Phi_\Delta(\lambda)}, \quad \Phi_\Delta(\lambda) = \left[\sqrt{\lambda} + \Delta(1 - \lambda)^2 \right]$$

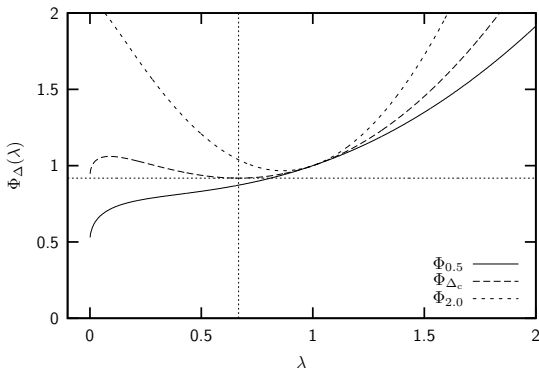
where $\lambda = \delta M_d / \delta M$ is the fraction taken up by the droplet.

⇒ Optimize $\Phi_\Delta(\lambda)$ in λ for given Δ



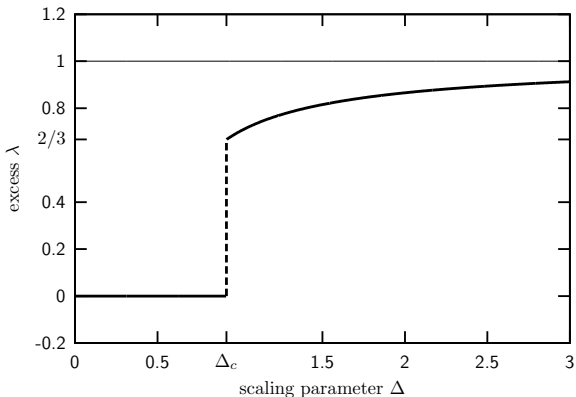
Recall: $\lambda = \delta M_d / \delta M$, $\Delta = 2 \frac{m_0^2}{\chi_{TW}} \frac{v_L^{3/2}}{V}$

- $\Delta < \Delta_c$: $\lambda_{\min} = 0$
- $\Delta = \Delta_c = (1/2)(3/2)^{3/2} \approx 0.92$: $\lambda_{\min} = \lambda_c = 2/3$
- $\Delta > \Delta_c$: $\lambda_{\min} > 2/3$



General d : $\Delta_c = \frac{1}{d} \left(\frac{d+1}{2} \right)^{\frac{d+1}{d}}$, $\lambda_c = \frac{2}{d+1}$

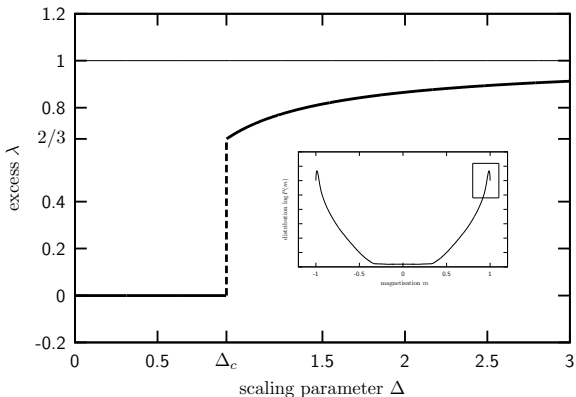


Solution: $\lambda = \delta M_d / \delta M \simeq$ droplet size $\Delta = 2 \frac{m_0^2}{\chi \tau_w} \frac{v_L^{3/2}}{V} \simeq$ scaled magnetization ($\Delta = 0$: peak location)

Solution:

$\lambda = \delta M_d / \delta M \simeq$ droplet size

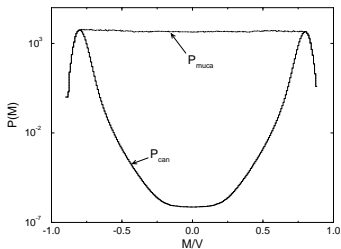
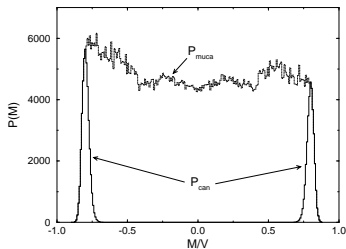
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Biskup *et al.*, *Europhys. Lett.* **60** (2002) 21; *Comm. Math. Phys.* **242** (2003) 137

Numerical Studies

Suppressed two-phase region: use multicanonical type of simulation in magnetisation



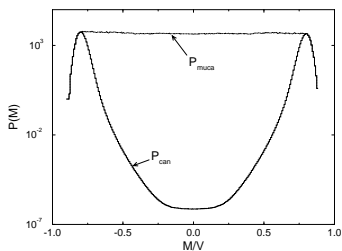
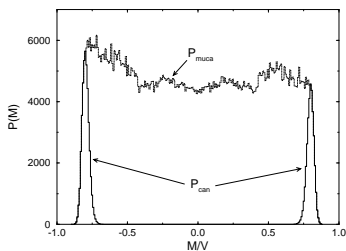
Clear NON-random-walk behaviour observed !

Neuhaus & Hager, J. Stat. Phys. 116 (2003) 47



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Suppressed two-phase region: use multicanonical type of simulation in magnetisation

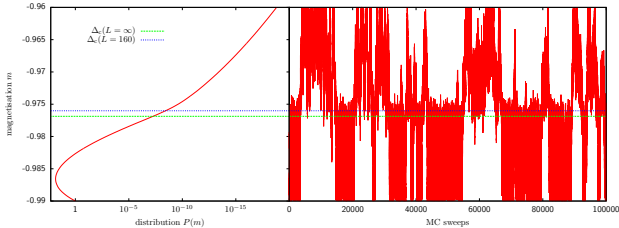


Clear NON-random-walk behaviour observed !

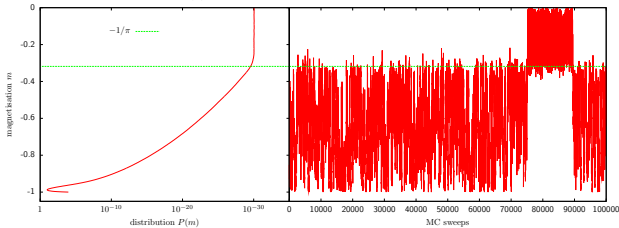
Neuhaus & Hager, J. Stat. Phys. **116** (2003) 47



Reason: Two “hidden” barriers along transition trajectory



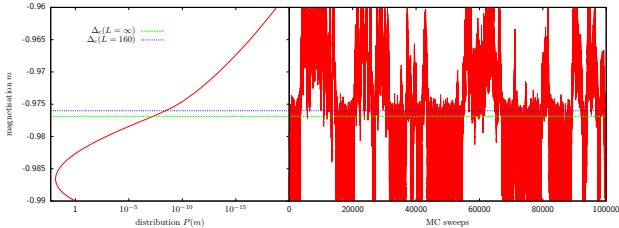
evaporation/
condensation



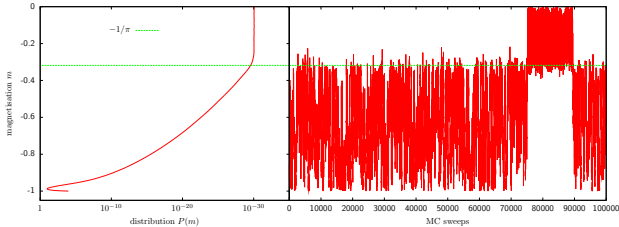
droplet/strip



Reason: Two “hidden” barriers along transition trajectory



evaporation/
condensation



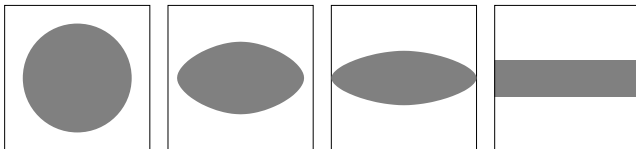
droplet/strip



Droplet/strip barrier rather well understood in 2D:

- strip: interface length = $2L$
- circular droplet: Same length for radius $R = L/\pi$, area = L^2/π
- # overturned spins = L^2/π , hence (assuming isotropic interface tension)
- barrier located at about $m = m_0/\pi$

But $2R = \frac{2}{\pi}L < L$:



Leung & Zia, J. Phys. A **23** (1990) 4593

MC Simulations: Equilibrium Droplet Formation

Three goals:

- Test analytical prediction for the thermodynamic limit
- Investigate finite-size corrections
- Study lattice universality

Simulation strategy:

- fix the total excess v_L
- v_L together with the “known constants” m_0 , χ , τ_W yields $\Delta(m_0, \chi, \tau_W, v_L)$.
- “micro-magnetical” simulation at:

$$M = -m_0 v_L + m_0 (V - v_L) \Rightarrow M = m_0 V \left(1 - 2 \frac{v_L}{V}\right)$$

- measure λ (\simeq relative size of largest droplet)



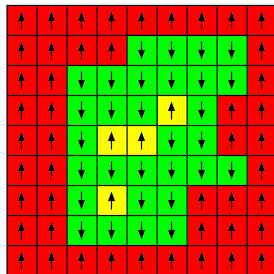
Algorithm:

- Kawasaki dynamics ($M = \text{const.}$)
- measure

$$\lambda = v_d/v_L,$$

the largest droplet size v_d (i.e., second largest cluster), by using the Hoshen-Kopelman algorithm

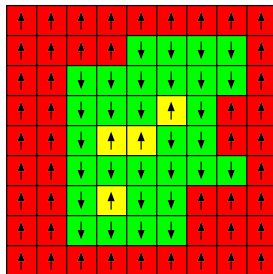
Difficulty: v_d is the **area** of the second largest cluster
 \Rightarrow What is inside and what is outside?



Idea: The Hoshen-Kopelman algorithm assigns to every cluster a unique number \Rightarrow **interface** between spins of the **largest** and **second largest** cluster

Droplet algorithm:

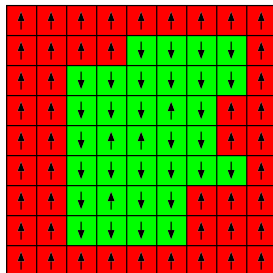
- do Hoshen-Kopelman
- for every new cluster, mark positions of spins belonging to that cluster
- starting from a spin inside the second largest cluster, apply “flood-fill”
- flood-fill stops only at spins belonging to the largest cluster



Idea: The Hoshen-Kopelman algorithm assigns to every cluster a unique number \Rightarrow **interface** between spins of the **largest** and **second largest** cluster

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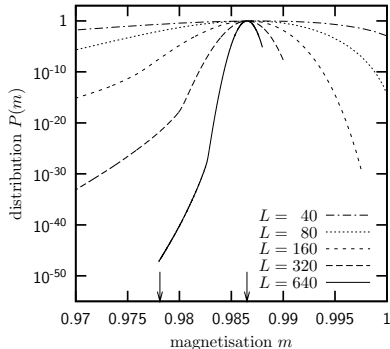
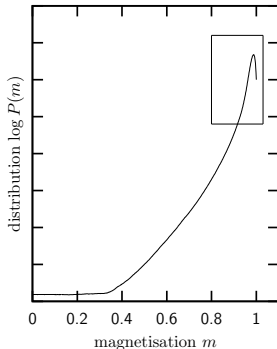


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Square lattice NN Ising model ($T = 1.5 \approx 0.66T_c$)



m range between arrows scanned with Kawasaki dynamics



Square lattice NN Ising model

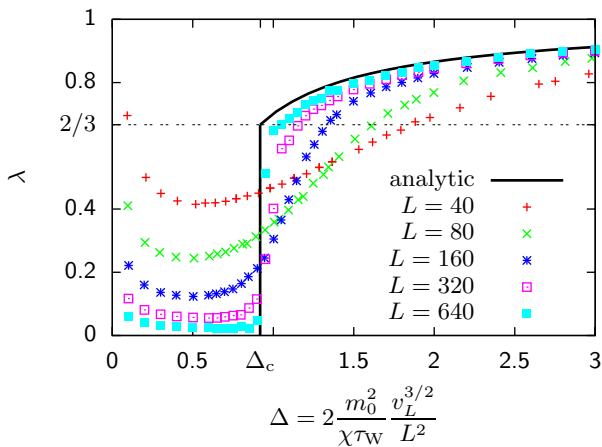
- $m_0(\beta) = \left(1 - \sinh^{-4}(2\beta)\right)^{1/8}$
- $\chi(\beta) = \beta \sum_{i=0}^n c_i u^{2i}$ with $u = 1/(2 \sinh(2\beta))$
with $c = \{0, 0, 4, 16, 104, 416, 2224, 8896, 43840, 175296, 825648, 3300480, 15101920, \dots\}$ up to order 323
(last term at $T = 1.5$: 10^{-158}). 2008–10: order 2000 ...
- $\tau_W(\beta) = 2\sqrt{W}$ with
$$W = \frac{4}{\beta^2} \int_0^{\beta\sigma_0} dx \cosh^{-1} \left[\frac{\cosh^2(2\beta)}{\sinh(2\beta)} - \cosh(x) \right]$$
and $\sigma_0 = 2 + \ln[\tanh(\beta)]/\beta$ the tension of an (1,0) interface.

Note: At $T = 1.5 \approx 0.66 T_c$, assuming isotropy:

$\tau_W \approx 2\sqrt{\pi}\sigma_0 = 4.219$; exact $\tau_W = 4.245$ (0.6% difference).



Square lattice NN Ising model ($T/T_c \approx 0.66$)

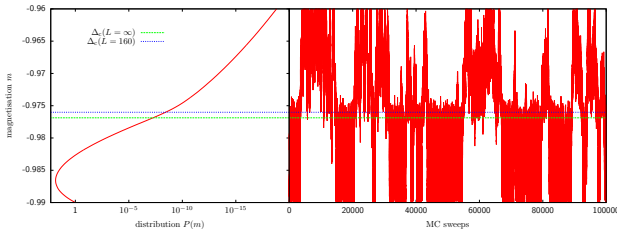


Nußbaumer, Bittner, Neuhaus & WJ, *Europhys. Lett.* **75** (2006) 716



Square lattice NN Ising model

Numerical problem: “Hidden” barrier is reflected in simulations

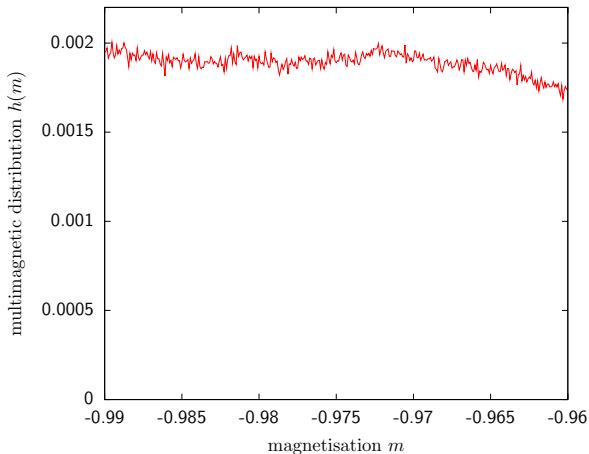


evaporation/
condensation

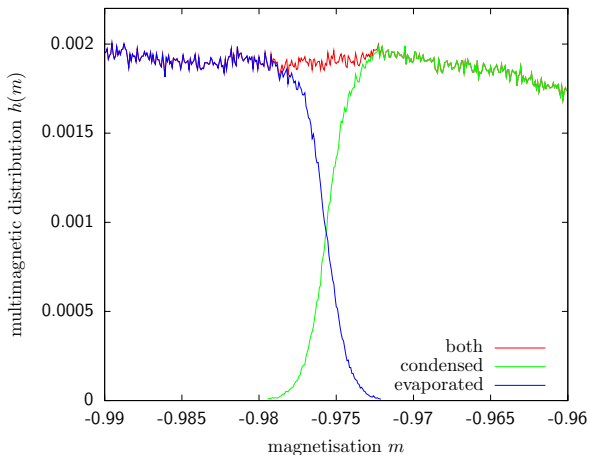
Similar “hidden droplet barriers” in spin glasses?



Evaporation/condensation barrier:

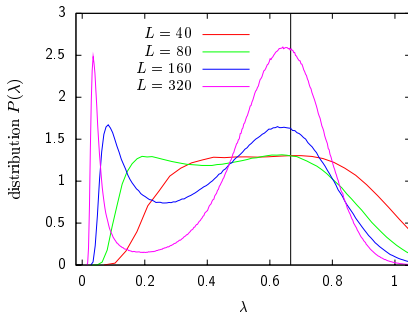


Evaporation/condensation barrier:



Square lattice NN Ising model

Distribution of fraction $\lambda = v_d/v_L$ for Δ close to Δ_c

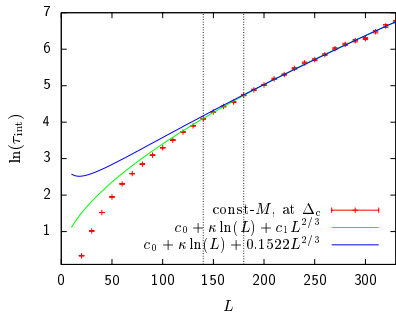
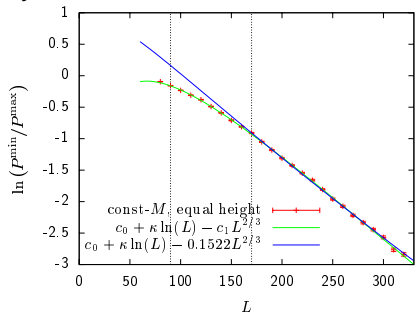


⇒ Clear coexistence signal !



Square lattice NN Ising model

Free-energy barrier resp. autocorrelation time scaling with system size



Theory: $\beta\Delta F \approx 0.1522 L^{2/3}$ (at $T = 1.5$)

Nußbaumer, Bittner & WJ, Prog. Theor. Phys. Suppl. **184** (2010) 400

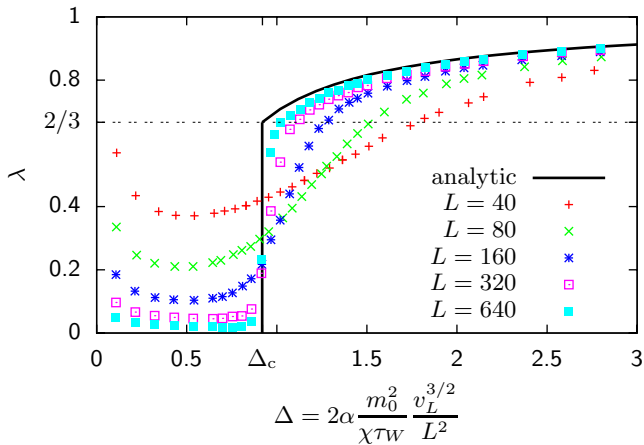


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Triangular lattice Ising model ($T/T_c \approx 0.66$)

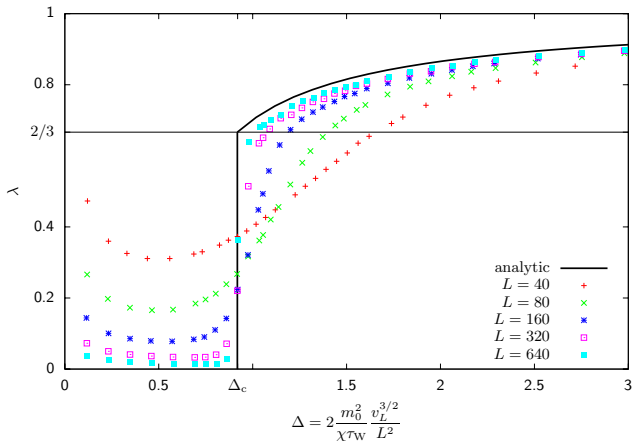


Outline

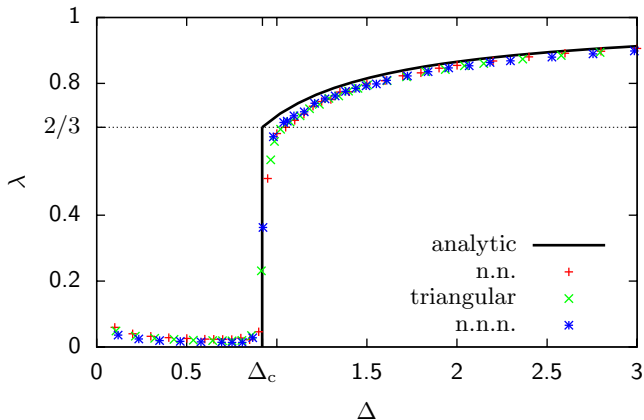
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Square lattice NNN Ising model ($T/T_c \approx 0.66$)



Check lattice universality ($L = 640, T/T_c \approx 0.66$)



Summary

- Analytical predictions for the asymptotic behaviour of the evaporation/condensation transition of the 2D square lattice NN Ising model confirmed numerically
- Finite-size corrections investigated
- Universality tested by studying triangular lattice and square lattice NNN Ising models

Nußbaumer, Bittner, Neuhaus & WJ, *Europhys. Lett.* **75** (2006) 716; Nußbaumer, Bittner & WJ, *Phys. Rev. E* **77** (2008) 041109; *Prog. Theor. Phys. Suppl.* **184** (2010) 400



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- EU–RTN Network “ENRAGE”
- Graduate School “BuildMoNa”
- Computer time NIC, Forschungszentrum Jülich

THANK YOU !



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See you there ?



Oops – wait another minute,
it's football time



Footballphysics

How probable is the next goal in soccer?

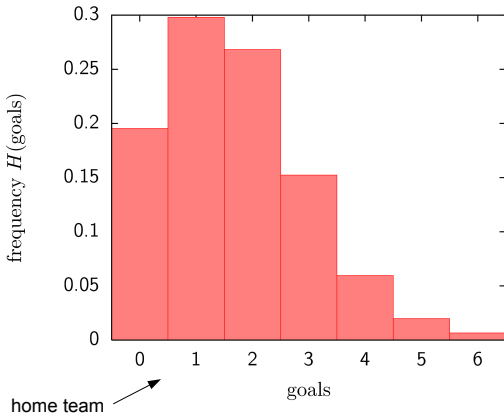
or

**Football fever: goal distributions and
non-Gaussian statistics**



A "typical" result:

1. Fußball-Bundesliga
2004/05 (Germany)



Goals	Matches
0	153
1	189
2	148
3	77
4	26
5	6
6	5



Question: what kind of distribution is this?



Simplest possible idea:

Goals are scored with a constant probability (in time)!

The scores of the home and the away team don't influence each other!



Binomial distribution

N number of time steps (?)

p probability to score a goal during one time step

goals

$$P_{N,p}(k) = \binom{N}{k} p^k (1-p)^{N-k}$$

$$N p = \lambda \rightarrow p = \frac{\lambda}{N}$$

Poisson distribution

$$\lim_{N \rightarrow \infty} P_{N,p}(k) = P_{\lambda}(k) = \frac{\lambda^k}{k!} e^{-\lambda}$$

average number of goals: $\langle P_{\lambda}(k) \rangle_k = \lambda$



Maybe a model can help?



Ingredients:

1. Starting probability
2. Probability to score a goal depends on the actual score



3 different updates if a goal is scored:

home scores away scores

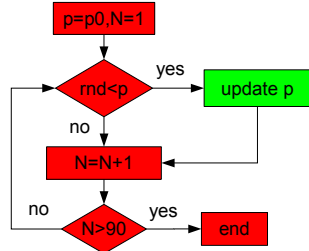
model "A": $p_h \leftarrow p_h + \kappa_h$ $p_a \leftarrow p_a + \kappa_a$

model "B": $p_h \leftarrow p_h \cdot \kappa_h$ $p_a \leftarrow p_h \cdot \kappa_a$





model "C": $p_h \leftarrow p_h \cdot \kappa_h$ $p_h \leftarrow p_h / \kappa_h$
 $p_a \leftarrow p_a / \kappa_a$ $p_a \leftarrow p_a \cdot \kappa_a$

κ : additional motivation
"success succeeds"

Algorithm:



What distribution is described by model "A", "model B", ...

take recursion relation  transform in difference equation
 perform continuum limit $N \rightarrow t$
 transform in differential equation
 solve differential equation using $p(n)$

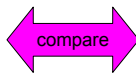
} independent of a specific $p(n)$

Finally,
for model "A"

$$P_{r,p}(n) = e^{-p_0 t} \frac{\Gamma(\frac{p_0}{\kappa} + n)}{n! \Gamma(\frac{p_0}{\kappa})} (1 - e^{-\kappa t})^n$$

negative binomial distribution

$$P_{r,p}(n) = \frac{\Gamma(r+n)}{n! \Gamma(r)} p^n (1-p)^r$$



$$p = 1 - e^{-\kappa t} \quad r = \frac{p_0}{\kappa}$$

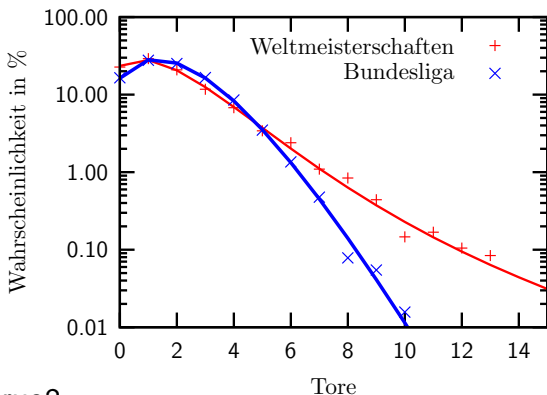
No closed expressions for model "B" and model "C" so far!



The **negative binomial distribution** describes a **process** with:

- $n+r$ Bernoulli trials (0,1)
- success probability p
- n times success
- r times failure
- last attempt is a failure

Bio-, Econo-, Sociophysics, . . . , **Footballphysics**

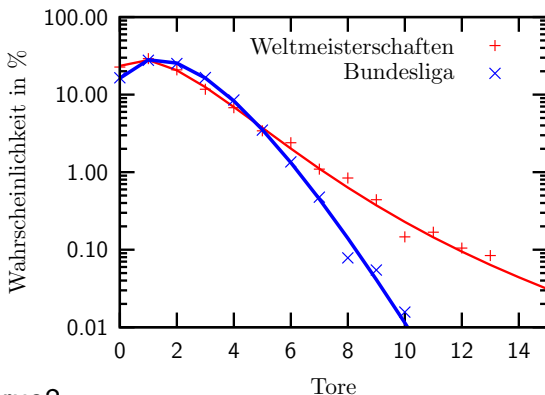


Is all that true?

Disclaimer: No NIC supercomputer (JUMP, JUBL, . . .) time was used
 E. Bittner, A. Nußbaumer, M. Weigel & WJ, Eur. Phys. J. B **67** (2009)
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Bio-, Econo-, Sociophysics, . . . , **Footballphysics**



Is all that true?

Watch the EURO2012 games and check yourself !!!

Disclaimer: No NIC supercomputer (JUMP, JUBL, . . .) time was used

E. Bittner, A. Nußbaumer, M. Weigel & WJ, Eur. Phys. J. B **67** (2009)

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