

I. DYNAMICAL PHASE TRANSITIONS :
INTRODUCTION & EXAMPLES
LARGE DEVIATIONS FUNCTIONS

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Warwick University
2013

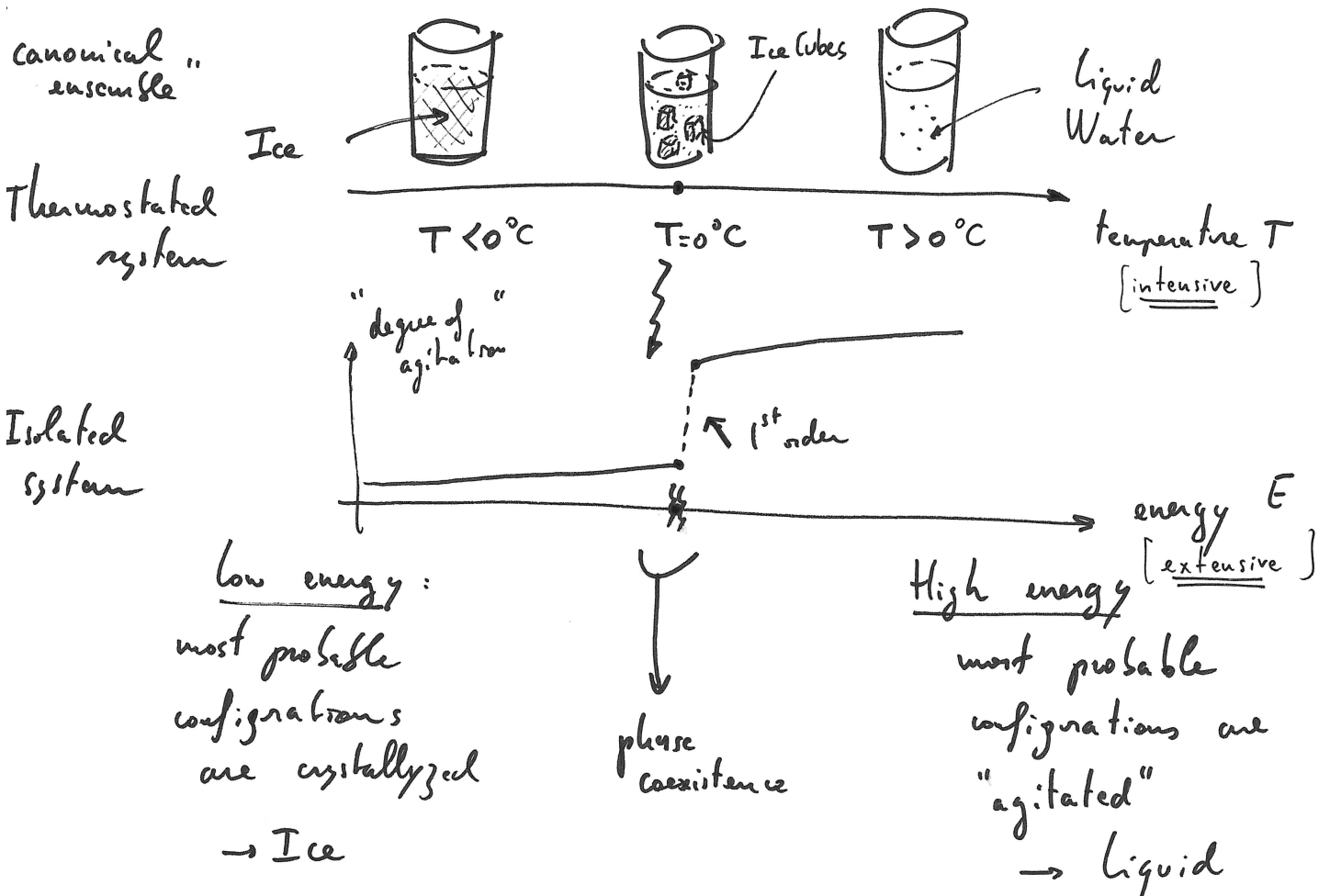
→ 1st lecture, very descriptive -

Outline:

1. Examples of Phase Transitions : static & dynamical coexistence
2. Fluctuations of Dynamical Observables ; Large Deviation Functions

EXAMPLES OF PHASE TRANSITIONS - STATIC & DYNAMICAL COEXISTENCE

1. a (Static) liquid-solid phase coexistence

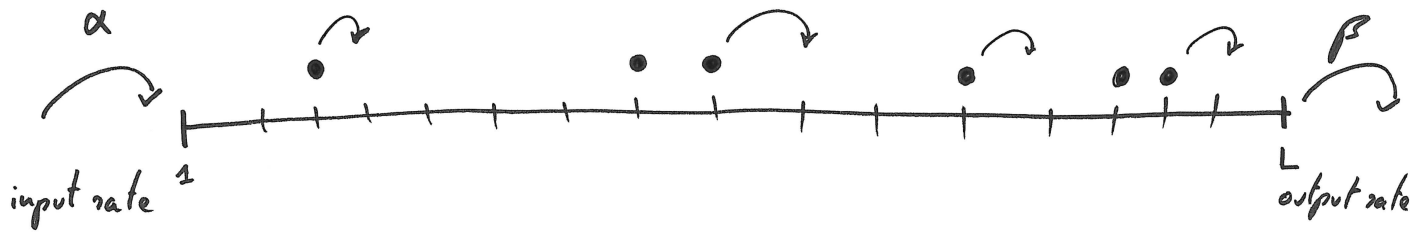


1.6. 1D transport: a traffic model

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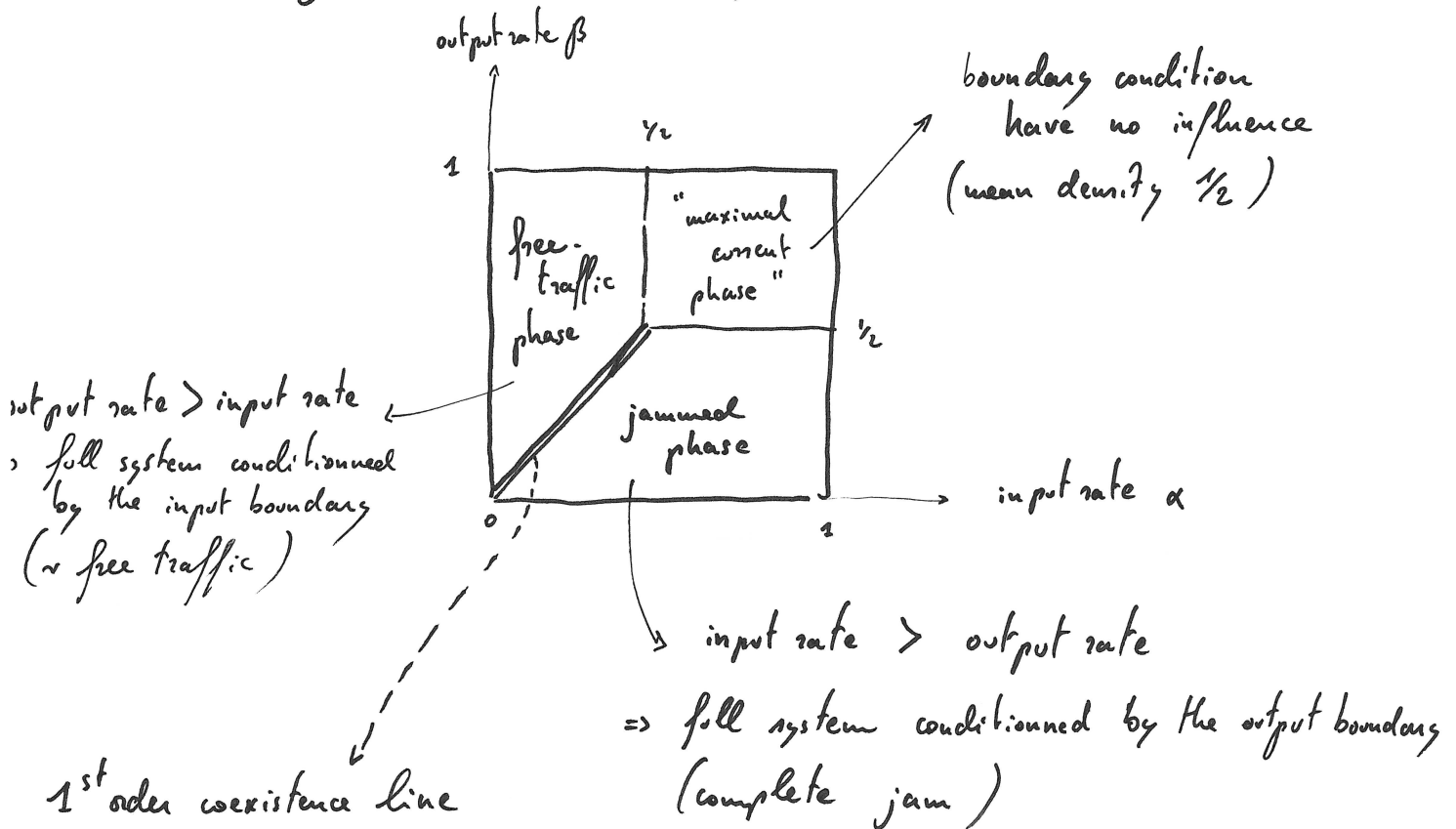
(Driving in dense traffic: one observes large fluctuations of jamming)

Very simple model ("Ising model of non-equilibrium"): TASEP
 [Totally Asymmetric Simple Exclusion Process]



- each site either occupied • or not
- jumps to the right occurs with rate 1 provided the target site is empty
 ("Exclusion rule")

Phase diagram (in the large-size limit $L \rightarrow \infty$)

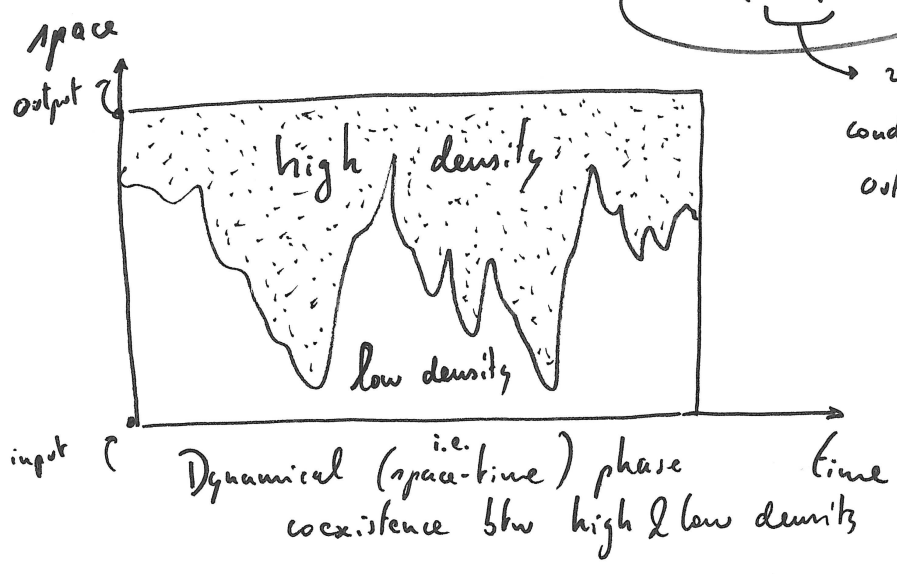


1.6' Coexistence between jammed & free traffic.

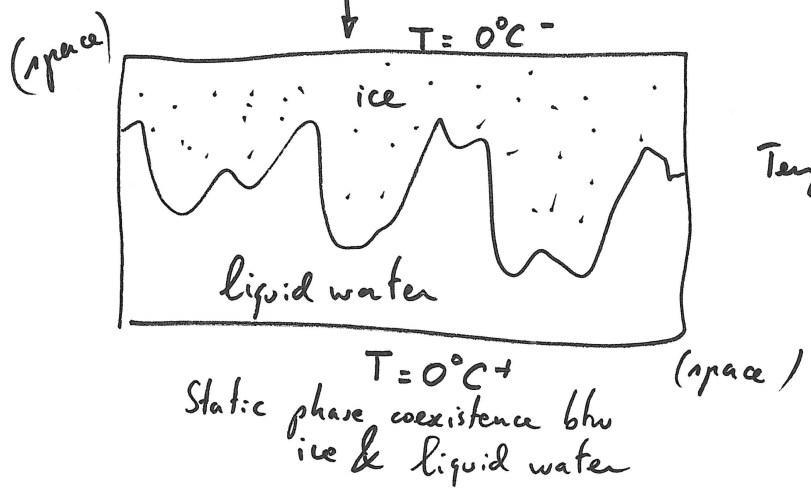
$0 < \alpha = \beta < 1/2$

represents realistic conditions where output \approx input

TASEP



analogy with static (1st order) phase coexistence



Temperature = 0°C

⚠ However, this is only an analogy: there is no exact mapping (time $\rightarrow \infty$; long-range correlations in TASEP)

Question: for the TASEP

- How can one quantify the dynamical phase coexistence?
- Can it explain the large fluctuations of the density?
- Can one "isolate" the coexisting phases & characterize their properties?

↳ A tool: Large Deviation Functions -

c. Glass Formers; Kinetically Constrained Models

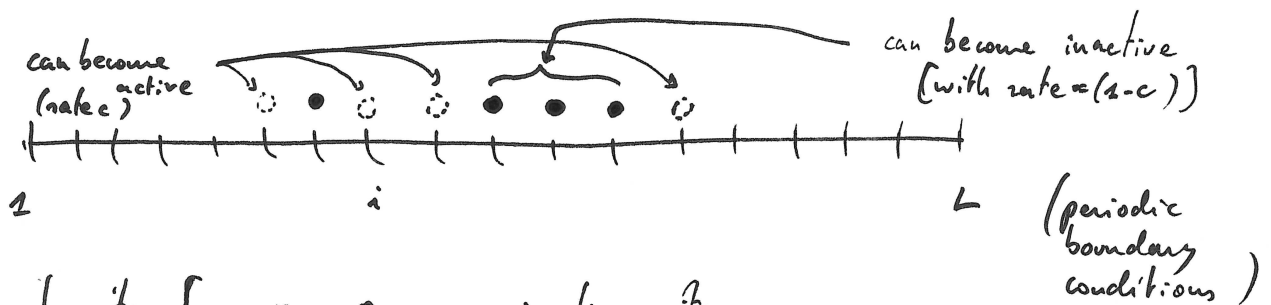
• Motivations: from realistic glassy phenomena / models

[See Keys et al, PRX 1 021013 (2011)] for binary mixtures

→ physical picture: active regions enhance activity in their vicinity

• An example of 'Kinetically Constrained Model' (KCM)

a 1D Fredrickson-Anderson (FA) model:



On each site: $\begin{cases} n_i = 0 & : \text{inactive site} \\ n_i = 1 & : \text{active site} \end{cases}$

KINETIC CONSTRAINT

Dynamics:

transition rates are:

Each site follows a death & birth process, conditioned by the number of neighbors & $n_i \in \{0,1\}$

$$\begin{cases} \text{activation } W(n_i=0 \rightarrow n_i=1) = c \cdot (n_{i-1} + n_{i+1}) \\ \text{inactivation } W(n_i=1 \rightarrow n_i=0) = (1-c) \cdot (n_{i-1} + n_{i+1}) \end{cases}$$

Link with discrete time:

during a small time interval Δt :

$$\text{prob}(n \rightarrow n') = \begin{cases} \Delta t c (n_{i-1} + n_{i+1}) & \text{activation} \\ \Delta t (1-c) (n_{i-1} + n_{i+1}) & \text{inactivation} \\ 1 - \Delta t [\dots] & \text{nothing happens} \end{cases}$$

• Remarkable Feature:

Trivial Equilibrium Distribution

$$P_{eq}(\{n_i\}) = \prod_i c^{n_i} (1-c)^{1-n_i}$$

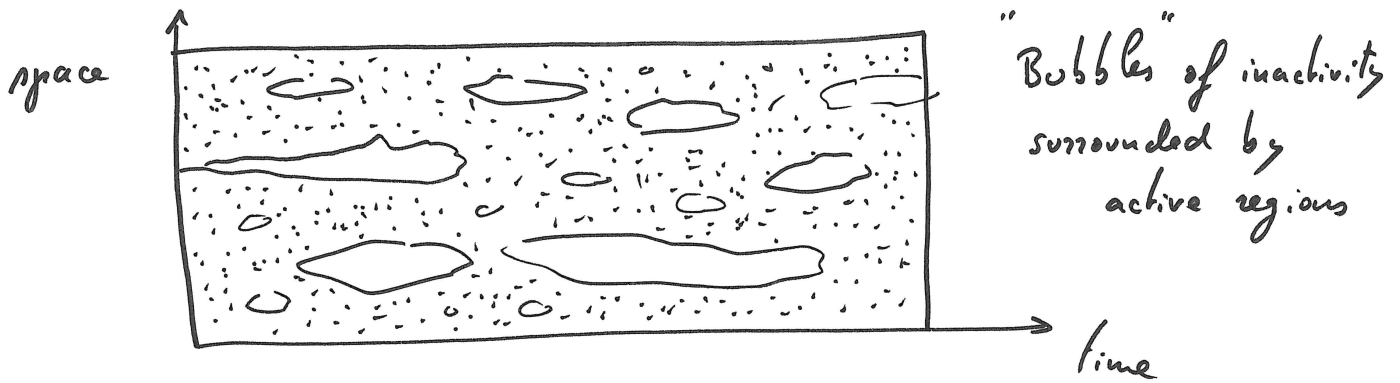
→ uncorrelated; same as WITHOUT CONSTRAINT

↕
Non-trivial Dynamics

Slow decay of correlation functions
Hysteresis & Ageing
→ Glassy-like properties

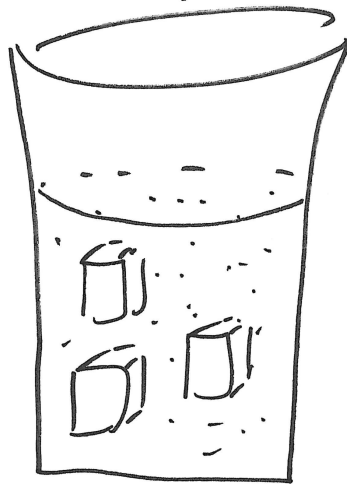
• Space-time diagram :

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"Bubbles" of inactivity
surrounded by
active regions

analogy



ice cubes in
liquid water @ $T = 0^\circ\text{C}$

Again : there is no exact mapping

However, this suggest a picture in terms of a

dynamical phase coexistence btw active & inactive regions

→ Questions: how can one formalize / quantify this observation?

• can one study the properties of the active / the inactive phase?

• can this be used to explain the slow dynamics?

↳ use large deviation functions -

2. FLUCTUATION OF DYNAMICAL OBSERVABLES LARGE DEVIATION FUNCTIONS

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2.a - Dynamical order parameters

	Static phase transition	Dynamical Phase Transition
extensive order parameter	energy E (volume-extensive)	<ul style="list-style-type: none"> activity $K = \#\{\text{events}\}$ current $Q = \#\{\text{jump } \uparrow\} - \#\{\text{jump } \downarrow\}$ integrated density $\int_0^t n(\tau) d\tau$ → "time-extensive", i.e. $\propto t =$ duration of the history
intensive conjugated variable	$\beta = 1/\text{Temperature}$ fixes the mean energy	s

One focuses on histories followed by the system btw 0 & time t .

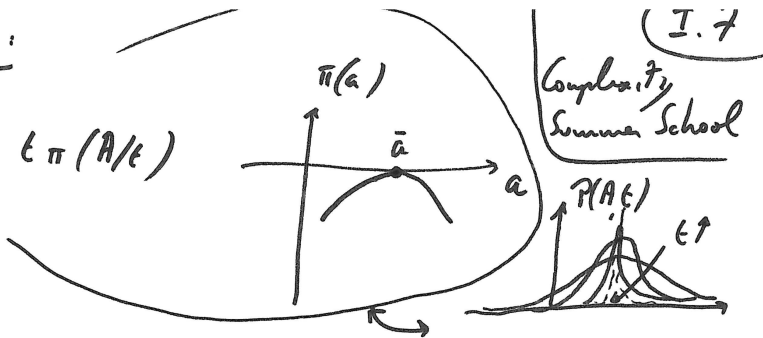
- * Dynamical order parameters: $A = K$ or Q or $\int_0^t n(\tau) d\tau$ or ...
- depend on (the realization of) the history followed by the system
 - "time-extensive", i.e. on average $\propto \left\{ \begin{array}{l} \text{duration } t \text{ of the} \\ \text{considered histories} \end{array} \right\}$
 - allow to classify histories and detect phase coexistence (or phase transition in general), in a dynamical sense.

* Equivalent of the microcanonical \rightarrow canonical: $\rightarrow [cf e^{-\beta E}]$
weight histories by $A e^{-sA}$

2.6 Large Deviation Functions:

P_A $P(A)$:

$$P(A \in \epsilon) \underset{\epsilon \rightarrow \infty}{\sim} e^{-\epsilon \pi(A/\epsilon)}$$

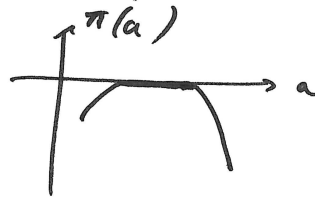


"normal" case $\pi(a)$ analytic

$$\pi(a) = \frac{(\bar{a}-a)^2}{2\sigma^2} + \dots \quad \text{Gaussian fluctuations of the observable } A.$$

"atypical case" $\pi(a)$ non analytic:

↳ converges to different cases of phase-transitions



'Ensemble inequivalence' which arises in the $L \rightarrow \infty$ limit in our case

→ Good tool to probe dynamical phase transitions.

⚠ However: difficult to understand -

Conjugated variable s :
["s-ensemble"]

s will fix the mean value of A for (slightly atypical) histories

$$\langle e^{-sA} \rangle \underset{\epsilon \rightarrow \infty}{\sim} e^{-\epsilon \psi(s)}$$

$\psi(s)$ cumulant generating function

$$\psi(s) = \max_a \{ \pi(a) - sa \}$$

$$\langle e^{-sA} \rangle = \int da P(a) e^{-sa} \underset{\epsilon \rightarrow \infty}{\sim} \int da e^{-\epsilon(\pi(a)-sa)}$$

dominated by the max

$$\psi^{(k)}(s) \Big|_{s=0} = \langle A^k \rangle$$

k-th cumulant

Physical interpretation

[lim $\epsilon \rightarrow \infty$]

$\langle O \rangle$ {histories with $A=at$ }

= $\langle O \rangle_{s^*(a)}$

with $\langle O \rangle_s = \frac{\langle O e^{-sA} \rangle}{\langle e^{-sA} \rangle}$

$\langle \dots \rangle$ = average for histories of duration t

$$\pi(a) = \min_s \{ \psi(s) + sa \}$$

$$s^*(a) = \operatorname{argmax}_a \{ \pi(a) - sa \}$$

biased statistics over histories

↳ In practice: biasing histories with s allows to track down dynamical phase transitions

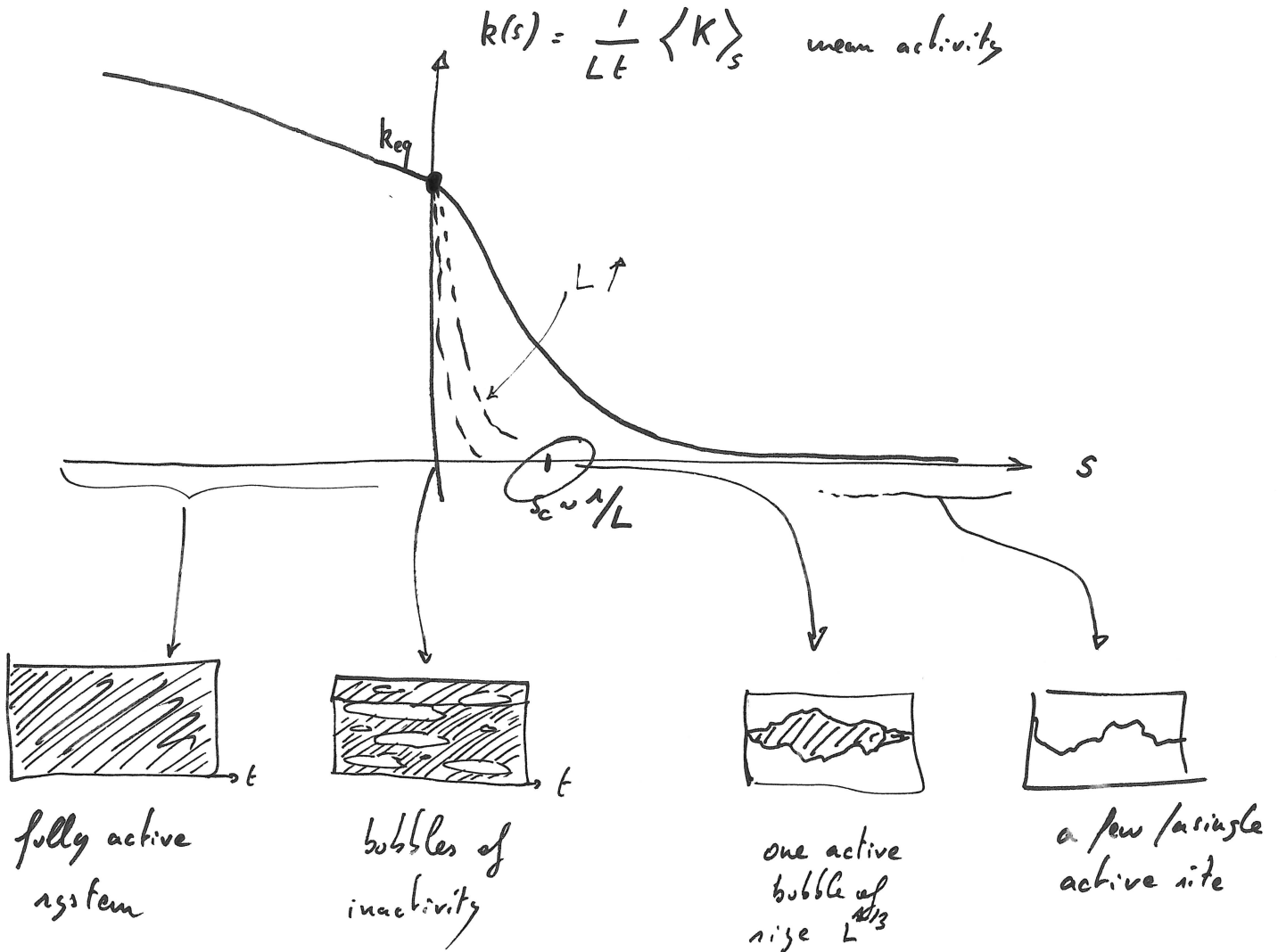
* Preview of the results for KCMs

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[An aim in the next lectures: provide details & algorithms]

L : system size

$$k(s) = \frac{1}{L} \langle K \rangle_s \quad \text{mean activity}$$



$\rightarrow s=0$ lies close to a ^{coexistence} critical dynamical point $s_c \sim 1/L$
 i.e. very close to