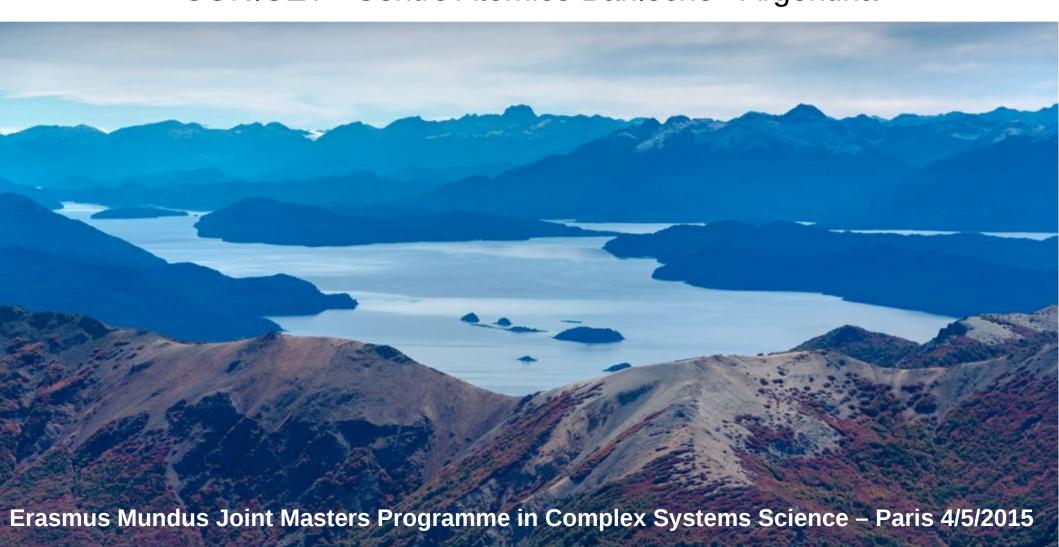
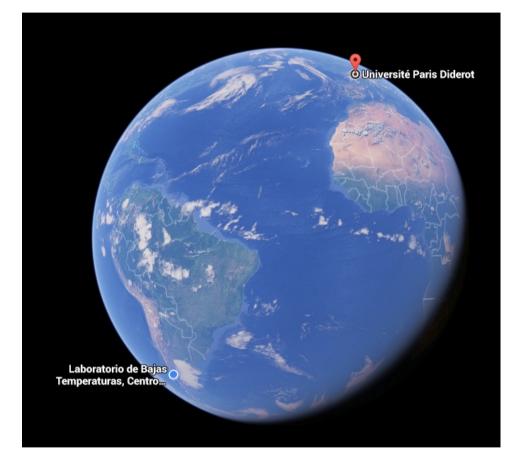
Interfaces in Random Media: "Playing the elastic string game" **Alejandro B. Kolton** CONICET - Centro Atómico Bariloche - Argentina



Bariloche, Argentina





Bariloche, Argentina, by Vivien Lecomte

Agenda

Motivation:

- Nature, experiments
- Phenomenological relevant questions

• Concepts:

- Universality, models, interesting quantities

• Playing the game:

- Numerical Algorithms
- Programming Graphics Cards
- Some answers, a lot of new questions!

Have you ever observed a similar phenomenon?



Magnetic Domain Wall

Domain 1

Wall

Domain 2

Magneto Optics

Thin films $e\approx 0.5nm$

DOWN

 $\sim 10 \mu m$

Disorder vs Elasticity A dynamic competition

UP

Response to H in the UP direction?

Magneto Optics

TRANSPORT PROBLEM

Mean velocity V vs applied field H and Temperature T?

V(H,T) = ?

V(H,T)?... does it matter?

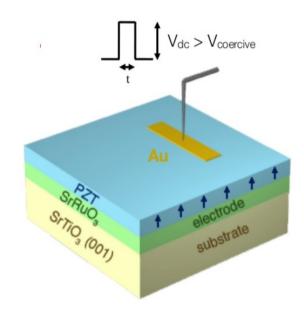
Interfaces Motion control - Applications

Disorder vs Elasticity → Universality [in collective transport]

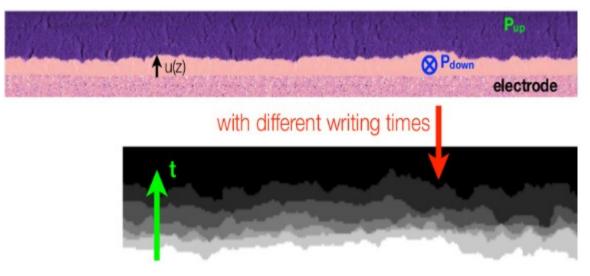
→ Similar Effective Physics shared by rather different systems

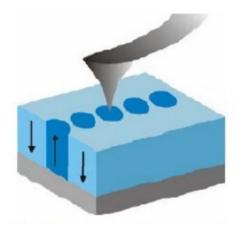
Ferroelectric Domain Walls

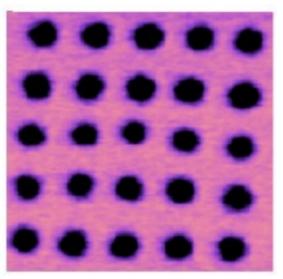
Ferroelectric Domain Walls



Pinning \rightarrow rough walls







Ferroelectric Domain Walls P. Paruch et al., (Geneva).

Contact lines in partial wetting

Wetting: three phases contact

Spreading parameter

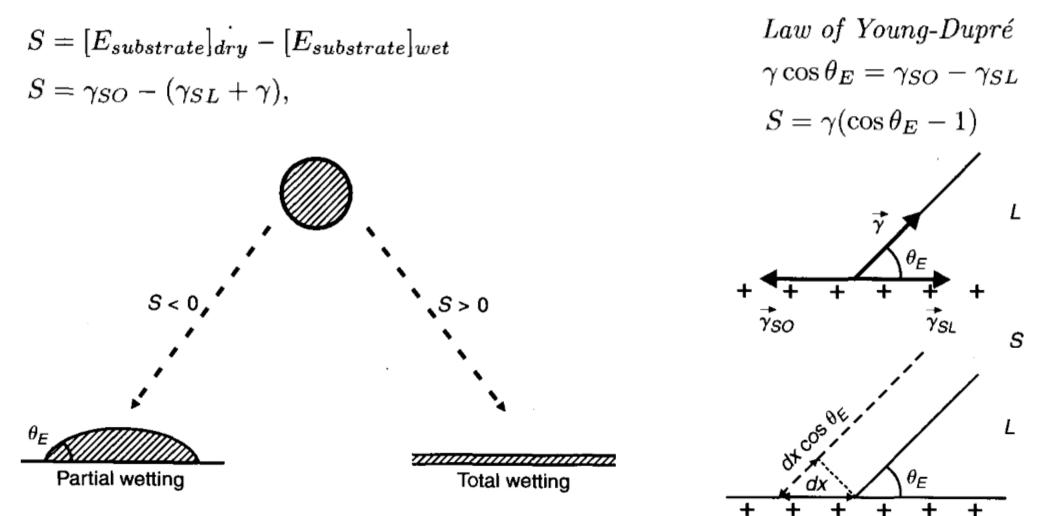


FIGURE 1.13. The two wetting regimes for sessile drops.

Capillarity and Wetting Phenomena de Gennes, Brochard-Wyart, Quere Contact Angle

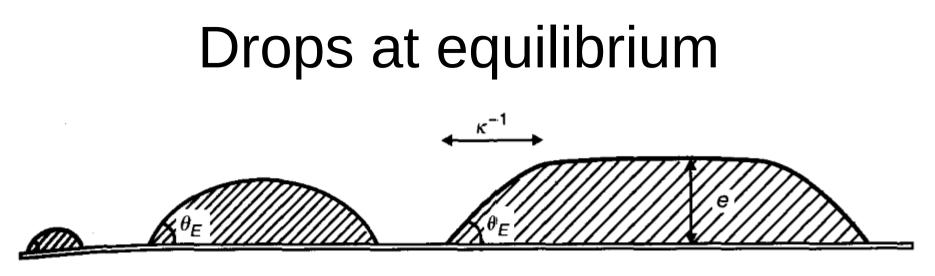


FIGURE 2.3. Water drops of increasing size on a sheet of plastic. Gravity causes the largest drops to flatten.

Capillary length

$$\kappa^{-1} = \sqrt{\gamma/\rho g}$$

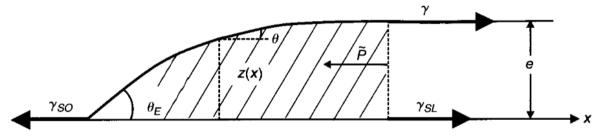


FIGURE 2.4. Equilibrium of the forces (per unit length of the line of contact) acting on the edge of a puddle. $\tilde{P} = \rho g e^2/2$ is the hydrostatic pressure.

$$F_g = -SA + \frac{1}{2}\rho g e^2 A. \longrightarrow e = 2\kappa^{-1} \sin\left(\frac{\theta_E}{2}\right)$$

Capillarity and Wetting Phenomena de Gennes, Brochard-Wyart, Quere

Evaporating Drops (at home)



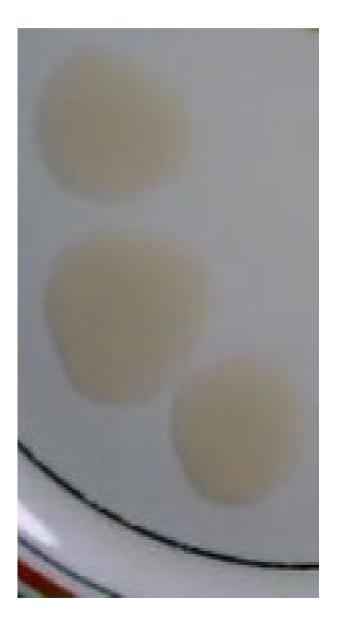
Before

After

Very common phenomenon



Try this at home!



- Why do solutes accumulate in the borders?
- Why they do not just shrink like this?

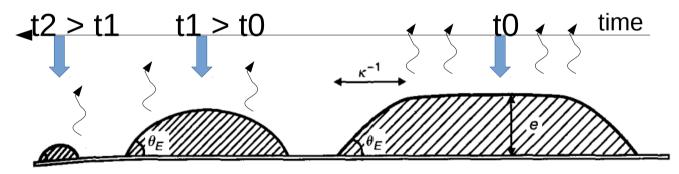


FIGURE 2.3. Water drops of increasing size on a sheet of plastic. Gravity causes the largest drops to flatten.

Capillarity and Wetting Phenomena de Gennes, Brochard-Wyart, Quere

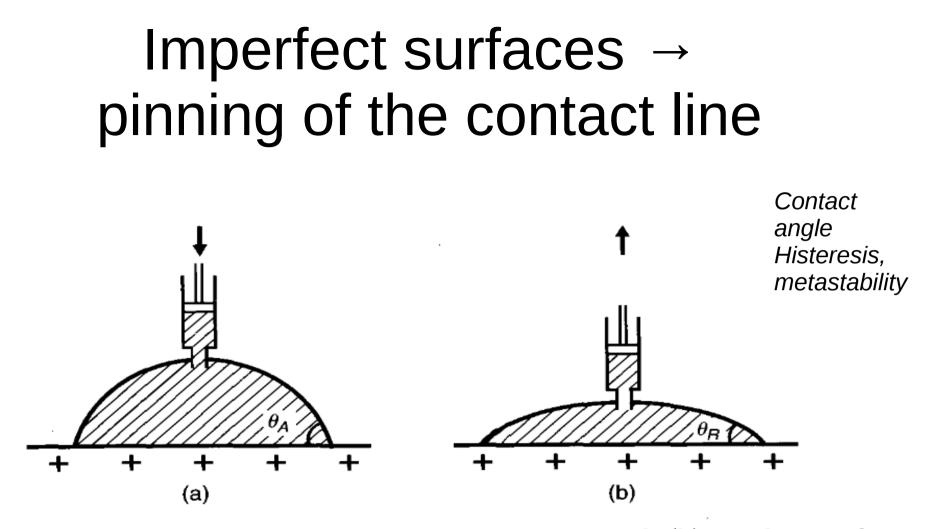


FIGURE 3.1. (a) Advancing angle when the drop is inflated; (b) receding angle when the drop is deflated.

Capillarity and Wetting Phenomena de Gennes, Brochard-Wyart, Quere

Ring Formation

NATURE | VOL 389 | 23 OCTOBER 1997

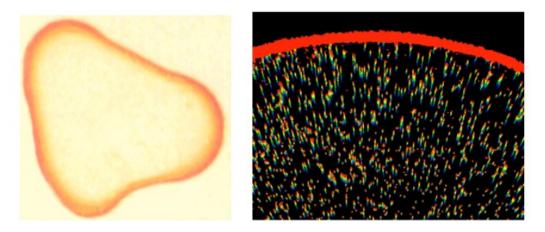


FIG. 1. a) A 2-cm-diameter drop of coffee containing one-weight-percent solids has dried to form a perimeter ring, accentuated in regions of high curvature. b) Video micrographs showing dispersion of fluorescent one-micron polystyrene spheres in water during evaporation, as described in the text. Multiple exposure shows different times in different colors to indicate the motion. Earliest time (blue) is 3 sec before latest time (red).

Capillary flow as the cause of ring stains from dried liquid drops

Robert D. Deegan*, Olgica Bakajin*, Todd F. Dupont†, Greb Huber*, Sidney R. Nagel* & Thomas A. Witten*

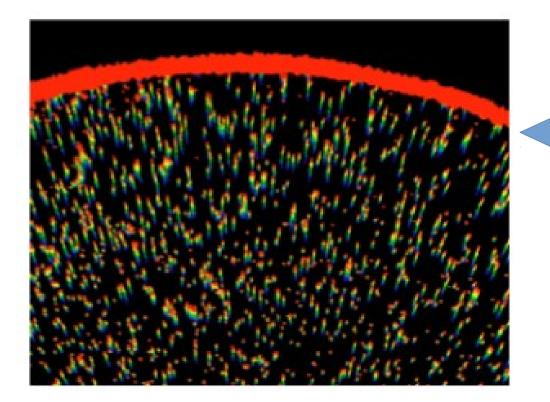
* James Franck Institute, 5640 South Ellis Avenue, Chicago, Illinois 60637, USA † Department of Computer Science, University of Chicago, 1100 East 58th Street, Chicago, Illinois 60637, USA

Our qualitative observations show that rings form for a wide variety of substrates, dispersed materials (solutes), and carrier liquids (solvents), as long as (1) the solvent meets the surface at a nonzero contact angle, (2) the contact line is pinned to its initial position, and (3) the solvent evaporates. In addition, we found that mechanisms typically responsible for solute transport—surface tension gradients, solute diffusion, electrostatic, and gravity effects—are negligible in ring formation. Based on these Ś

Effects of Particle Shape on Growth Dynamics at Edges of Evaporating Drops of Colloidal Suspensions

Peter J. Yunker,¹ Matthew A. Lohr,¹ Tim Still,^{1,2} Alexei Borodin,³ D. J. Durian,¹ and A. G. Yodh¹

¹Department of Physics and Astronomy, University of Pennsylvania, Philadelphia, Pennsylvania 19104, USA ²Complex Assemblies of Soft Matter, CNRS-Rhodia-University of Pennsylvania, UMI 3254, Bristol, Pennsylvania 19007, USA ³Department of Mathematics, Massachusetts Institute of Technology, Cambridge, Massachusetts 02139, USA (Received 23 July 2012; published 18 January 2013)



"surface growth" phenomena from the pinned (rough) contact line

Very common phenomenon

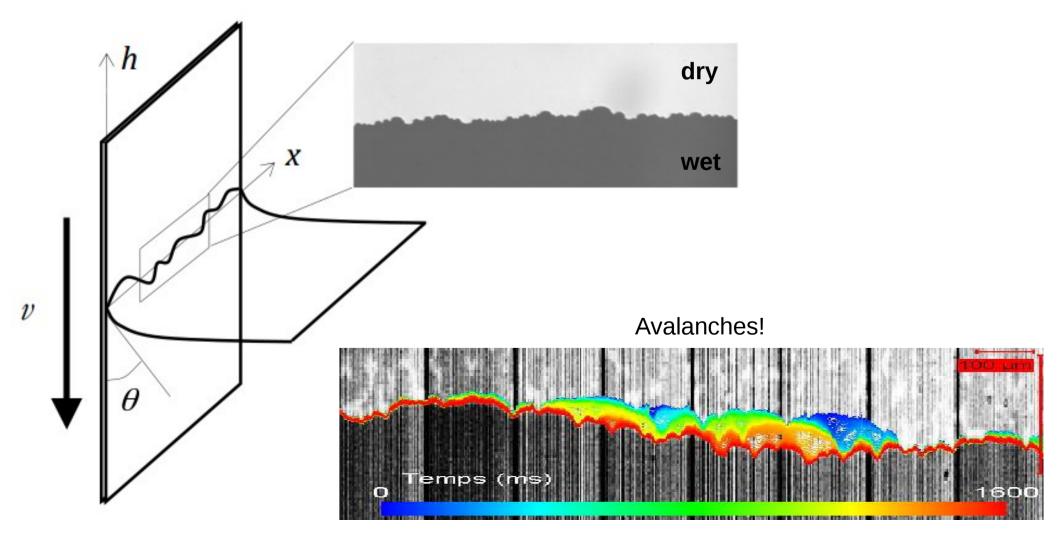


Memory of the pinned contact line (microscopically rough)

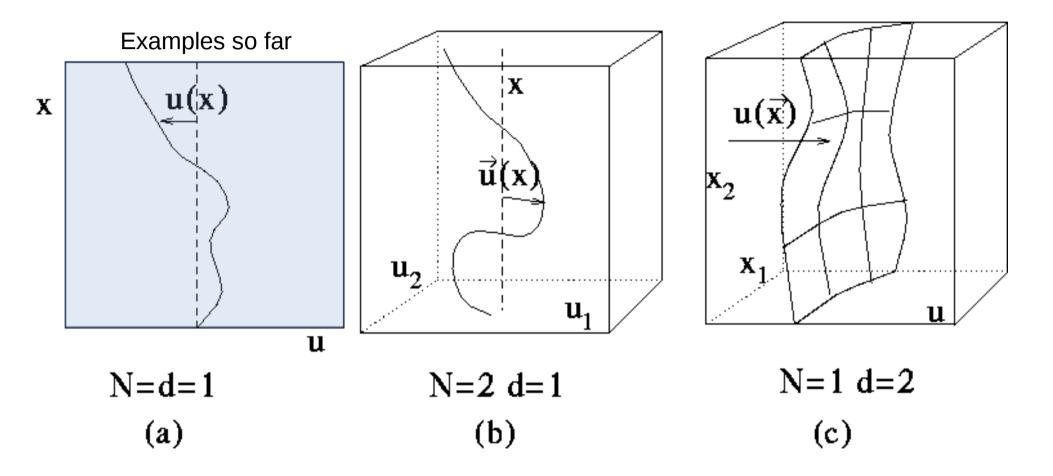
Dynamics of pinned contact lines in the lab

Contact lines in partial wetting Moulinet, Rolley (Paris).

Pinning \rightarrow roughness

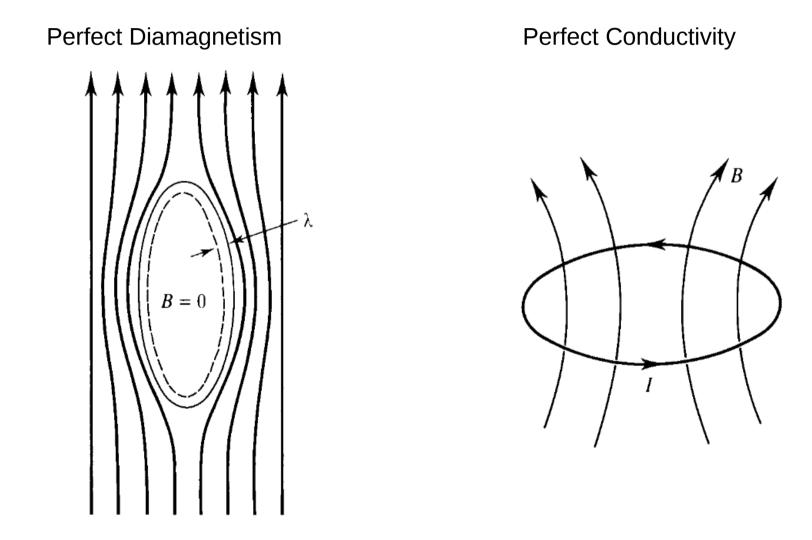


Other Disordered Elastic Manifolds



Or evenmore, a collection of this kind of objects but coupled by some repulsive interaction

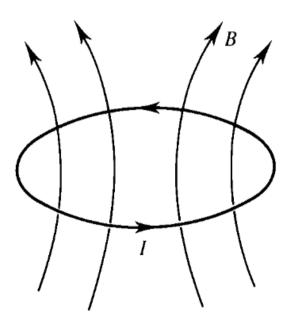
Phenomenological Superconductivity



Perfect Conductivity

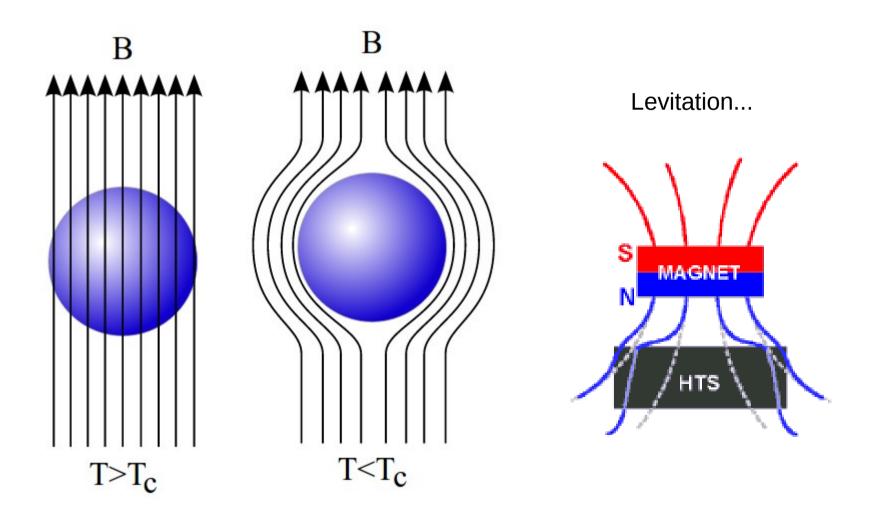
Electric cables for accelerators at CERN. Both the massive and slim cables are rated for 12,500 A. Top: conventional cables for LEP; bottom: superconductorbased cables for the LHC







Perfect diamagnetism



Diamagnetism and Levitation

Magnetic levitation is a method by which an object is suspended with no support other than magnetic fields. Magnetic force is used to counteract the effects of the gravitational and any other accelerations.

Earnshaw's theorem (1842) proves that using only paramagnetic or ferromagnetic materials it is impossible for a static system to stably levitate against gravity. [TRY IT WITH MAGNETS!]

Water is diamagnetic...

- Earnshaw theorem does not forbid stable static equilibrium in a magnetic field...
- Live animals, plants, etc, are full of water...
- Can we levitate them in a magnetic field?

QUIZ

(Dia)magnetic Levitation

Not forbiden by Earnshaw's theorem!



https://www.youtube.com/watch?v=A1vyB-O5i6E

Geim: Ig Nobel To Nobel

How to magnetically levitate yourself [using your diamagnetism]

The minimum criterion for diamagnetic levitation is $B\frac{dB}{dz} = \mu_0 \rho \frac{g}{\chi}$, where:

- χ is the magnetic susceptibility
- ho is the density of the material
- g is the local gravitational acceleration (-9.8 m/s² on Earth)
- μ_0 is the permeability of free space
- B is the magnetic field
- $\frac{dB}{dz}$ is the rate of change of the magnetic field along the vertical axis.

Assuming ideal conditions along the z-direction of solenoid magnet:

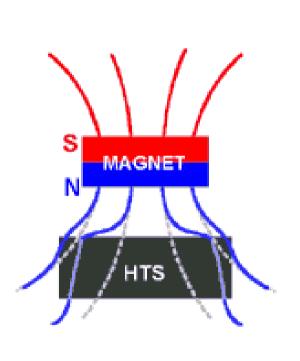
• Water levitates at
$$B \frac{dB}{dz} \approx 1400 \text{ T}^2/\text{m}$$

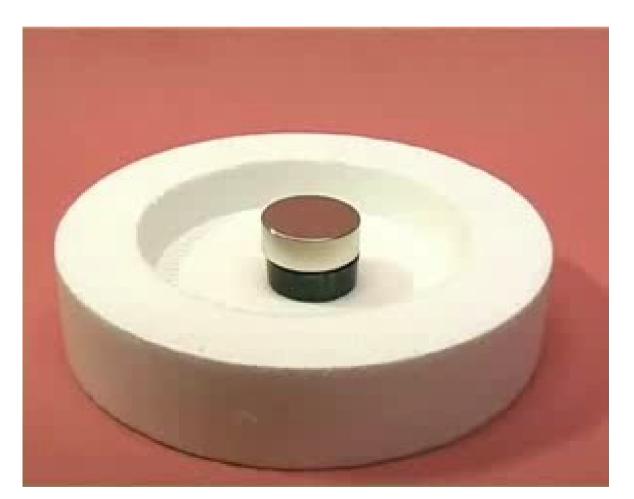
• Graphite levitates at $B \frac{dB}{dz} \approx 375 \text{ T}^2/\text{m}$.

Magnetic fields

10 ⁻⁹	nanotesla	100 pT to 10 nT	1 μG to 100 μG	magnetic field strength in the heliosphere
10 ⁻⁶	microtesla	24 µT	240 mG	strength of magnetic tape near tape head
10 ⁻⁵		31 µT	310 mG	strength of Earth's magnetic field at 0° latitude (on the equator)
		58 µT	580 mG	strength of Earth's magnetic field at 50° latitude
10 ⁻³	millitesla	0.5 mT	5 G	the suggested exposure limit for cardiac pacemakers by American Conference of Governmental Industrial Hygienists (ACGIH)
		5 mT	50 G	the strength of a typical refrigerator magnet [4]
10 ⁻¹		0.15 T	1.5 kG	the magnetic field strength of a sunspot
10 ⁰	tesla	1 T to 2.4 T	10 kG to 24 kG	coil gap of a typical loudspeaker magnet. ^[5]
		1 T to 2 T	10 kG to 20 kG	inside the core of a modern 60 Hz power transformer ^{[6][7]}
		1.25 T	12.5 kG	strength of a modern neodymium–iron–boron (Nd ₂ Fe ₁₄ B) rare earth magnet. A coin-sized neodymium magnet can lift more than 9 kg, pinch skin and erase credit cards. ^[8]
		1.5 T to 3 T	15 kG to 30 kG	strength of medical magnetic resonance imaging systems in practice, experimentally up to 8 T ^{[9][10]}
		9.4 T	94 kG	Modern high resolution research magnetic resonance imaging system; field strength of a 400 MHz NMR spectrometer
		11.7 T	117 kG	field strength of a 500 MHz NMR spectrometer
		16 T	160 kG	strength used to levitate a frog ^[11]

Magnetic Levitation



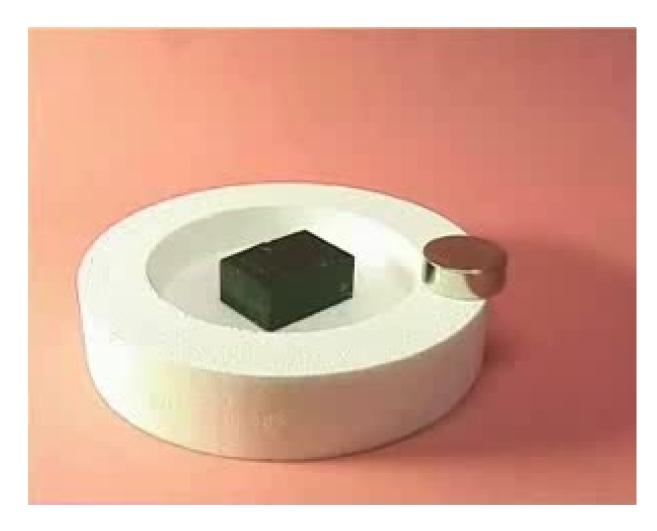


"Field cooling"

http://www.mn.uio.no/fysikk/english/research/groups/amks/superconductivity/levitation/



Magnetic Levitation



"Zero field cooling"

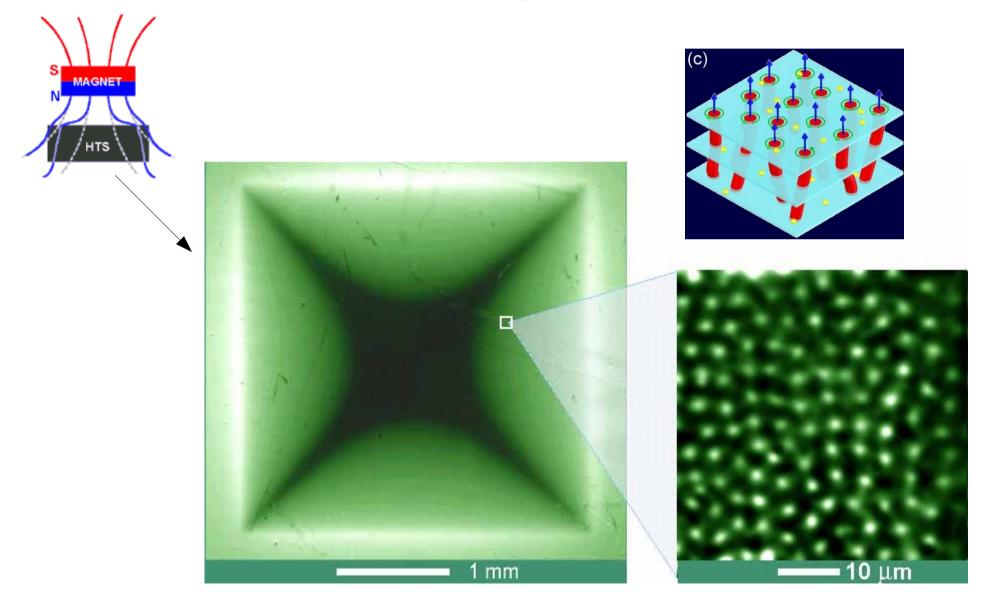
Some differences with respect to normal diamagnetic levitation:

- More effective
- There is also attraction!
- There are many static levitating configurations for the magnet!

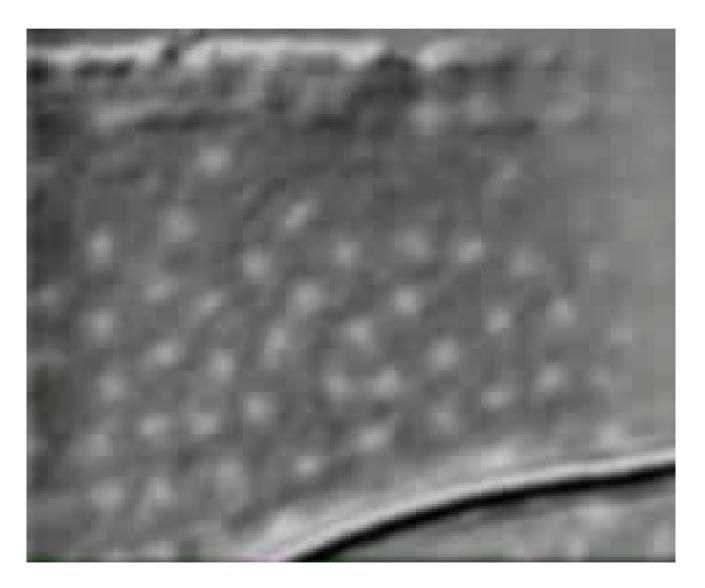
Where does this increased stability and metastable equilibrium comes from?

http://www.mn.uio.no/fysikk/english/research/groups/amks/superconductivity/levitation/

Vortices in superconductors

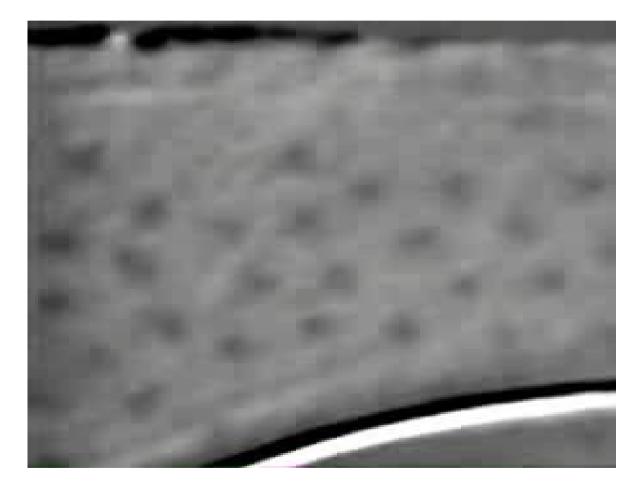


Vortices in superconductores



http://www.mn.uio.no/fysikk/english/research/groups/amks/superconductivity/sv/

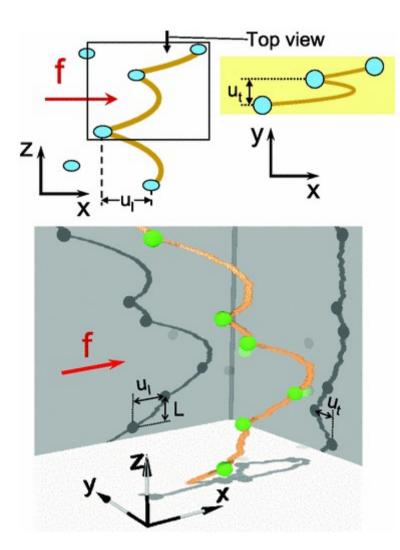
Vortices in superconductores

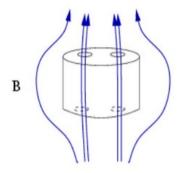


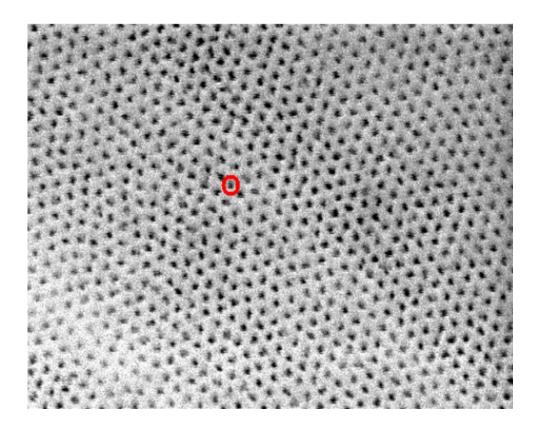
http://www.mn.uio.no/fysikk/english/research/groups/amks/superconductivity/sv/

Vortices in Superconductors

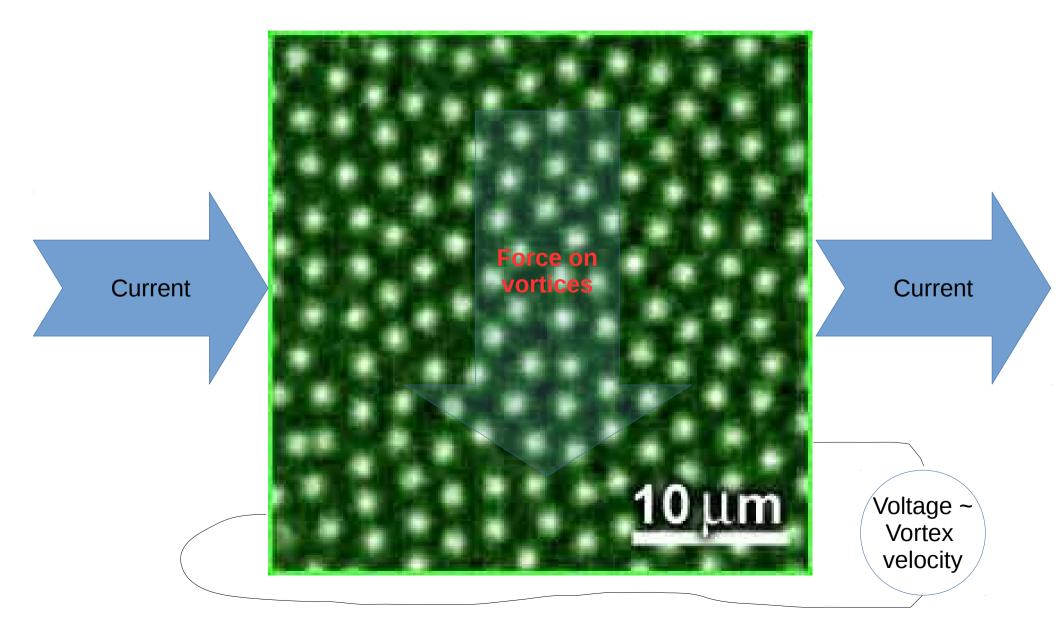
• Pinned elastic strings in 3D





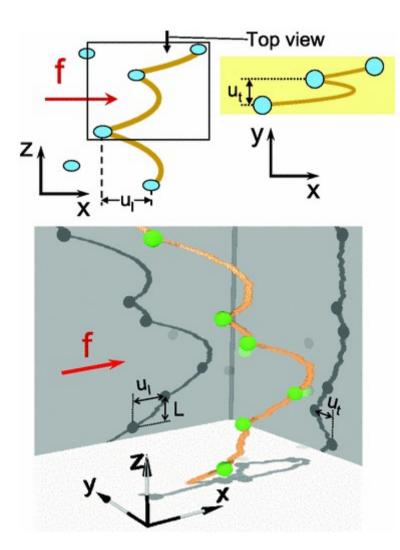


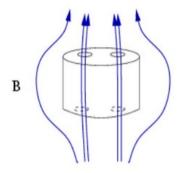
Vortex motion produces dissipation

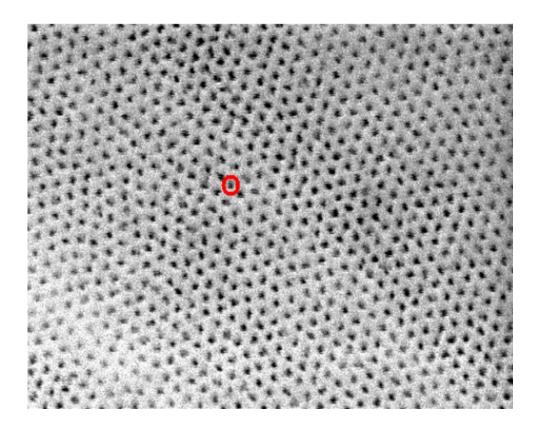


Vortices in Superconductors

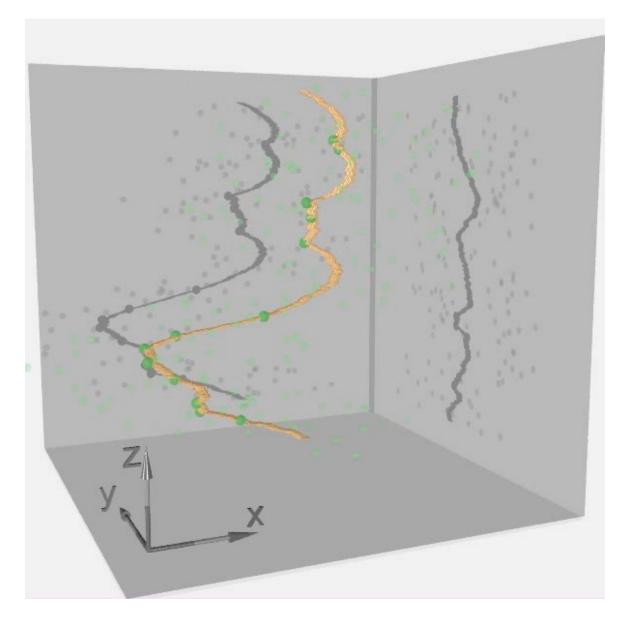
• Pinned elastic strings in 3D







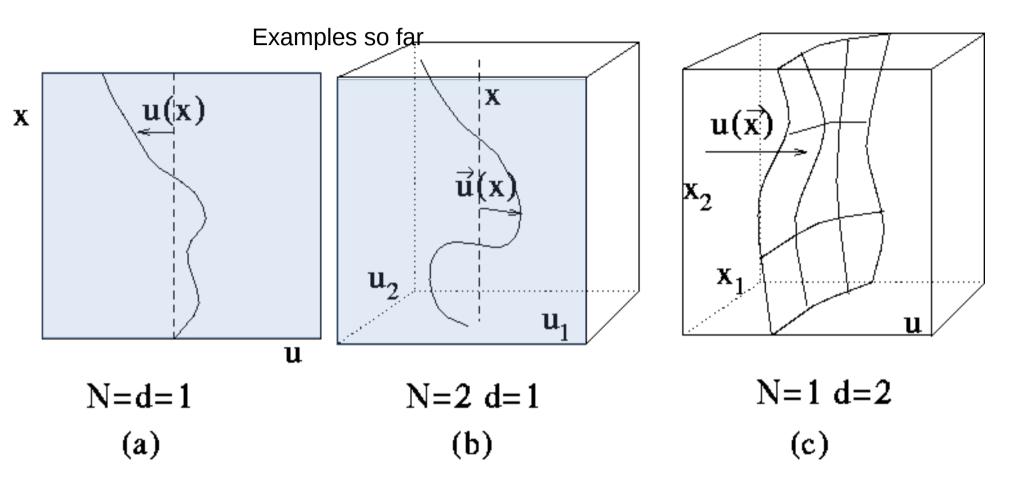
A moving Vortex simulation



A. Koshelev, A. B. Kolton PRB 2011

Vortex pinning reduce dissipation inHigh Tc superconductors!

Other Disordered Elastic Manifolds



- Ferromagnetic Domain walls
- Ferroelectric Domain walls
- Contact lines of liquids
- Etc...

- Polymers in random media. Earthquakes?
- Etc.
- Vortices in superconductors. Two dimensional domain walls

Earthquakes!

Earthquakes!

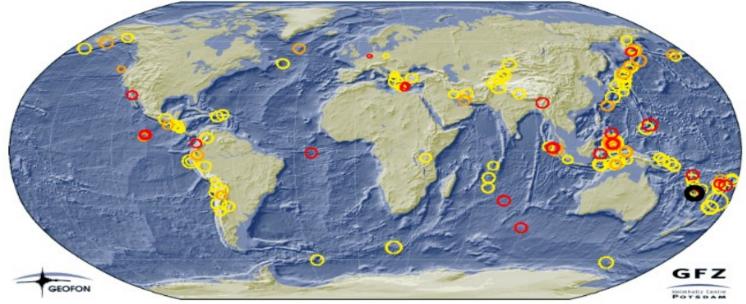


Valdivia (Chile), 22/5/1960; 9,5° Richter scale; largest registered earthquake in the world.

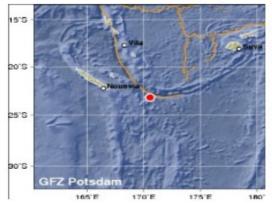


Where?

Automatic GEOFON Global Seismic Monitor



The displayed events are within the last 24 hours / 1-4 days / 4-14 days .



Most recent large event: Southeast of Loyalty Islands Magnitude: 5.8 Origin time: 2009-02-24 12:39:47.8 UTC Epicenter: 170.57°E 23.20°S Depth: 71 km Location: automatic See also: • The specific page for this event • The complete list of automatic GEOFON alerts

http://geofon.gfz-potsdam.de/geofon//seismon/globmon.html

Where?

Preliminary Determination of Epicenters 358,214 Events, 1963 - 1998

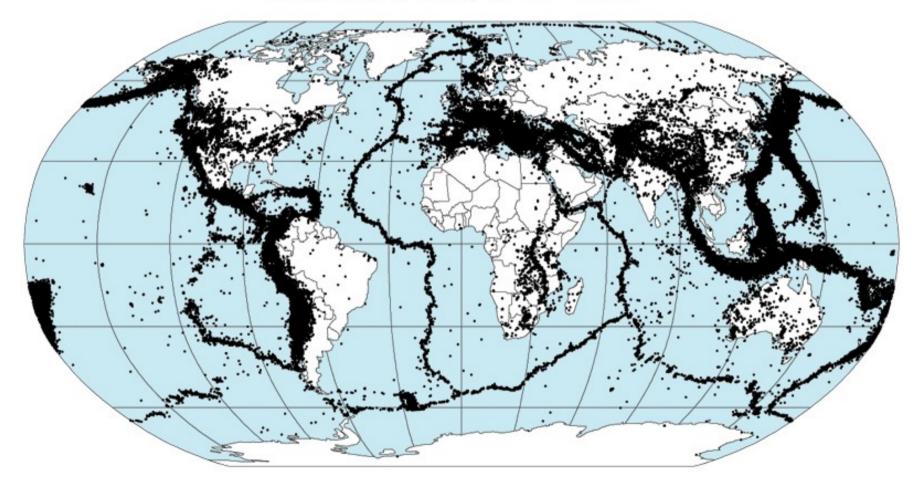


Plate Tectonics (1960!)

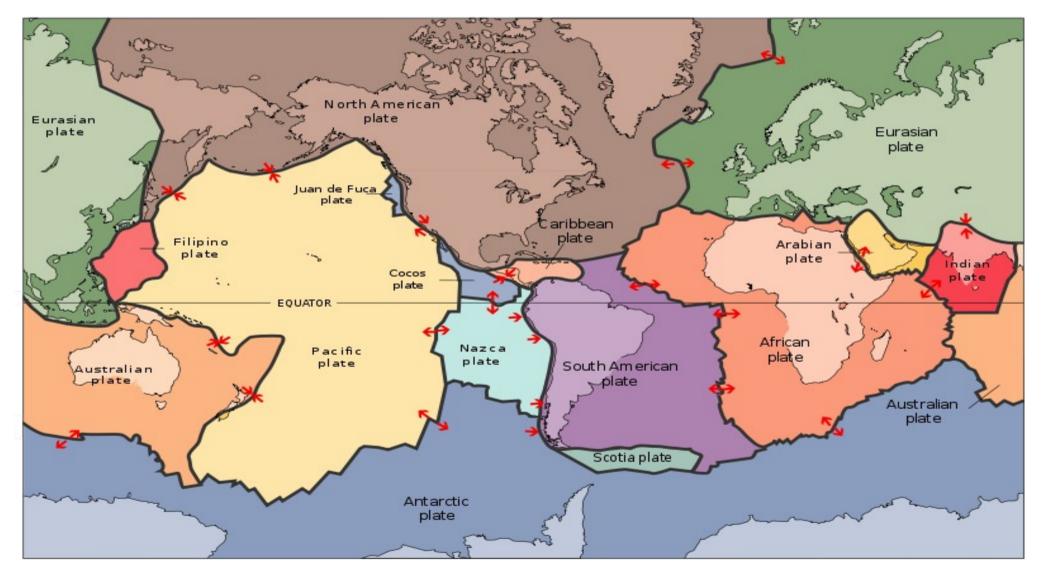
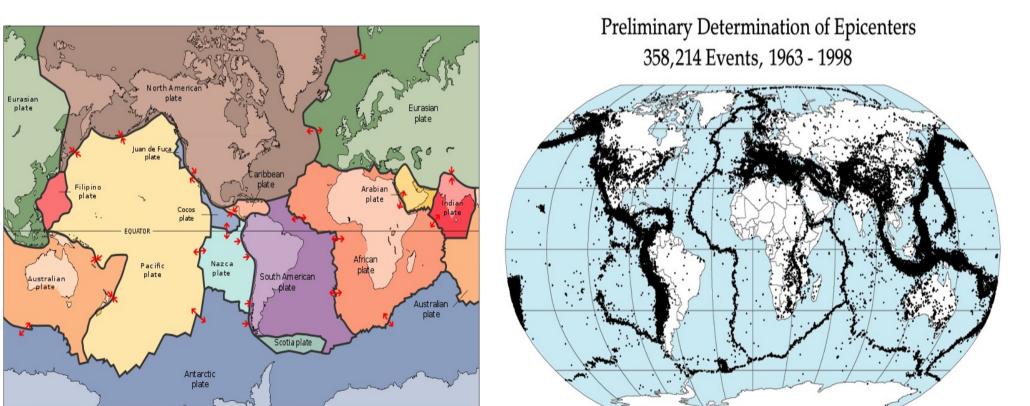


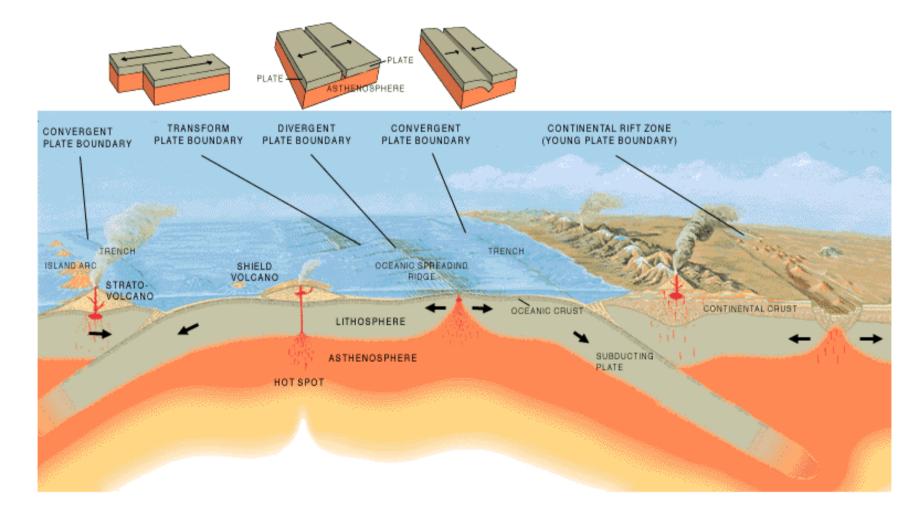
Plate tectonics & Earthquakes



Terremotos

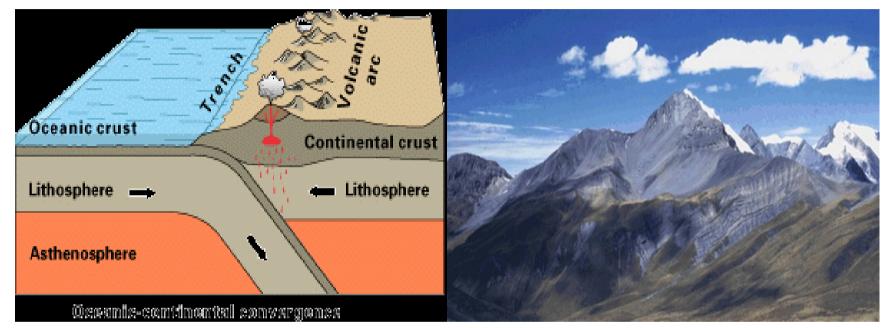


Plates borders

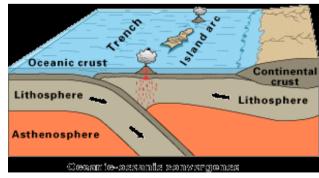


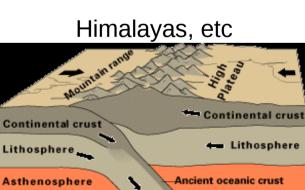
Convergent borders

Los Andes



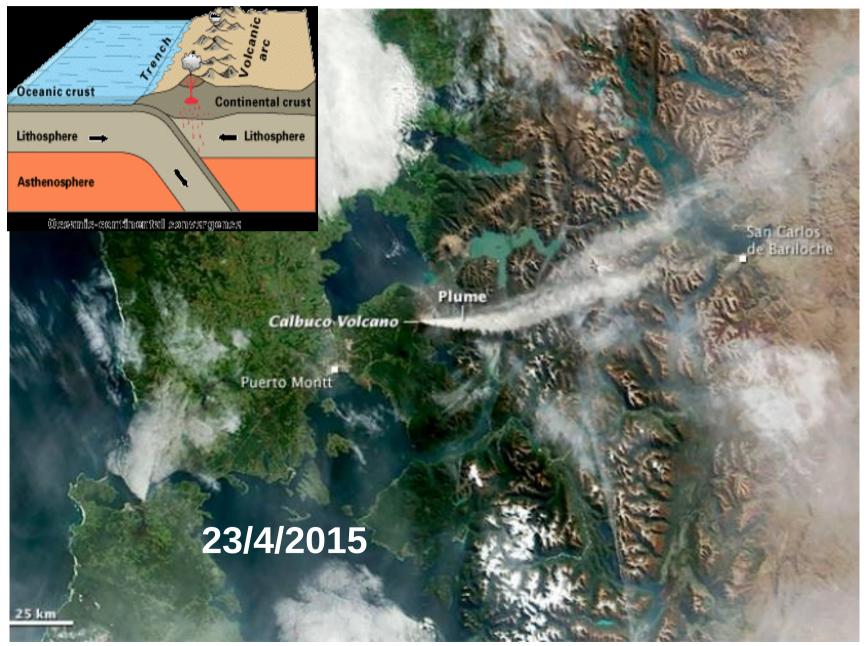
Marianas Islands, etc



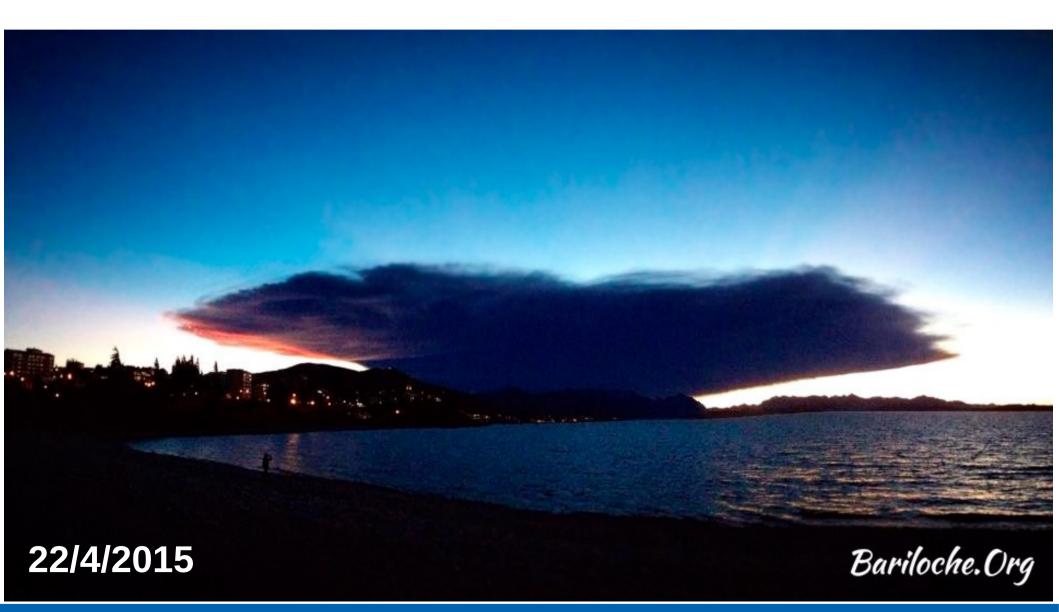


്രസ്തിനമസ്ക് ംരംസ്ത്രന്നായി മത്താശ്യാവന്നത

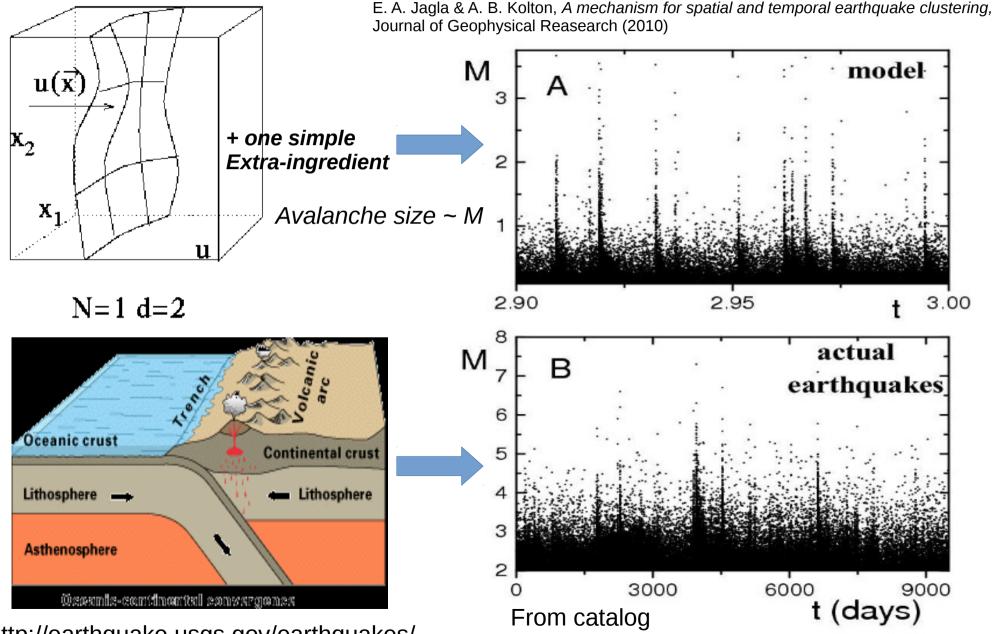
Convergent borders



Calbuco's volcanic plume

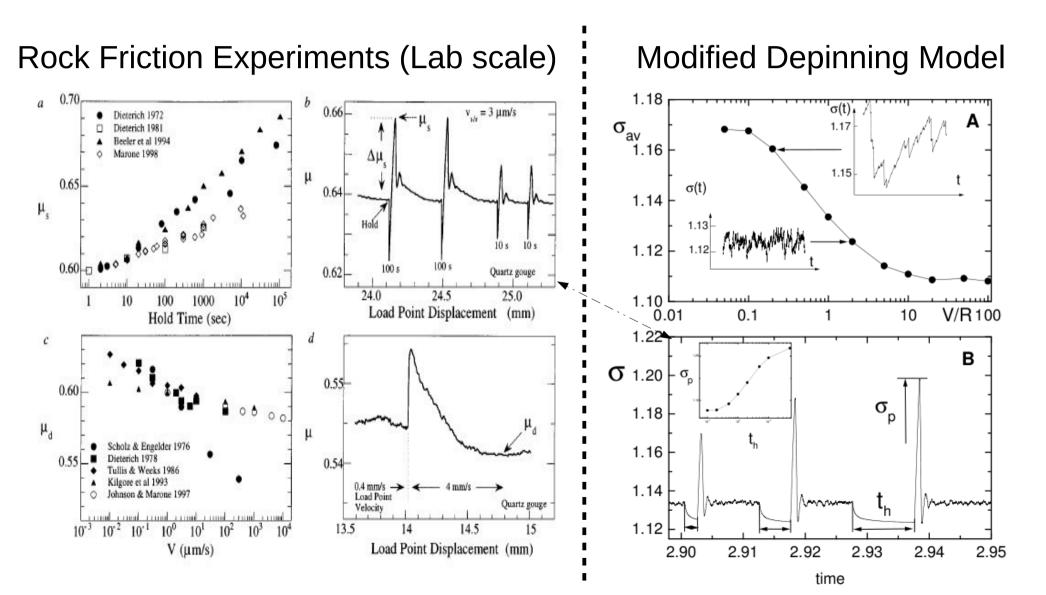


Simple model for earthquake statistics?



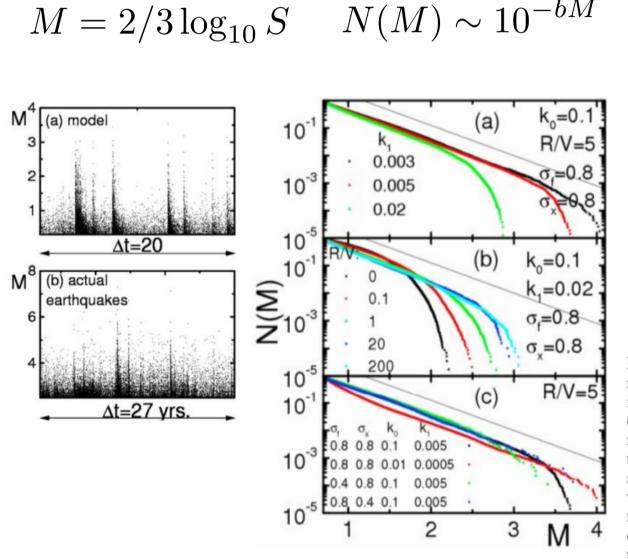
A mechanism for spatial and temporal earthquake clustering E. A. Jagla, A. B. Kolton, J. Geophys. Res. 115, B05312 (2010)

Original Motivation: Frictional properties of the model



A mechanism for spatial and temporal earthquake clustering E. A. Jagla, A. B. Kolton, J. Geophys. Res. 115, B05312 (2010)

Gutenberg-Richter



EVENTS

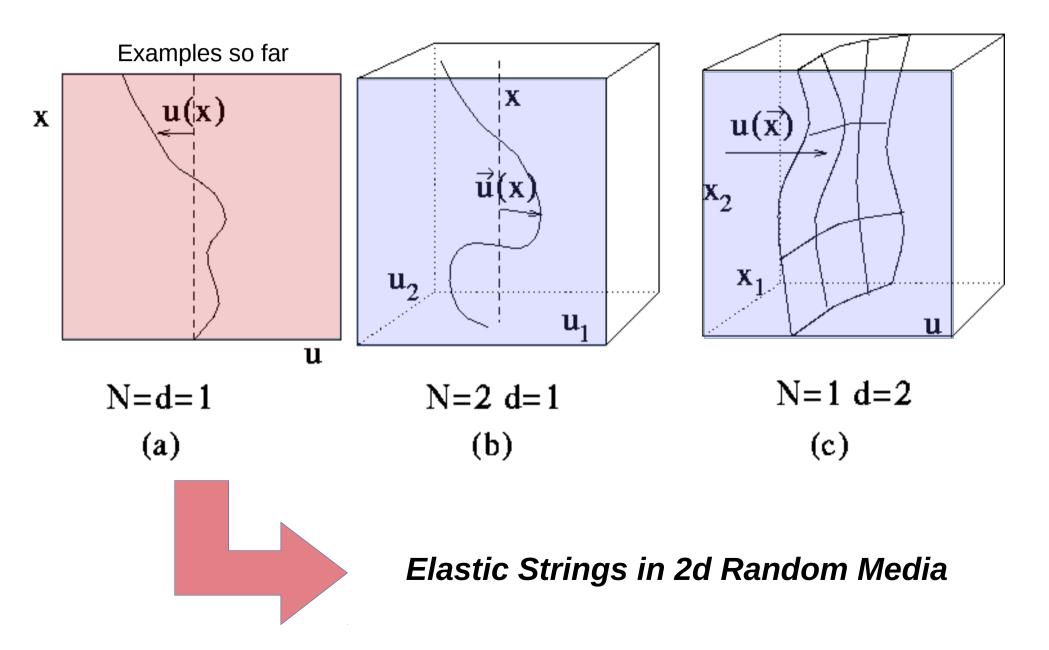
$N(t) = A/(t+c)^{p} + N_{0}$ t(sec) $10^{3} \cdot 10^{4} \cdot 10^{5} \cdot 10^{6} \cdot 10^{7} \cdot 10^{8}$ 10^{2} N(t) 10^{1} p=1.5

 $10^{-4} \ 10^{-3} \ 10^{-2} \ 10^{-1} \ 10^{0} \ 10^{1}$

Omori

Figure 7. Number of events after main shocks in our model (solid symbols) for parameters and system size as in Figure 3a, normalized to the background seismicity (i.e., $N(t \rightarrow \infty) = 1$). We stacked the events after about 450 main shocks with M > 3, and the lower magnitude cutoff used to count aftershocks is $M_0 = 1.5$. For comparison, the same analysis for data in the California region is presented with open symbols. In this case we sum over 7 events of magnitude larger than 6.0. For this case $M_0 = 2.0$. For reference, the continuous lines correspond to decays following the Omori law, with exponents p = 1.0 and p = 1.5.

Disordered Elastic Manifolds

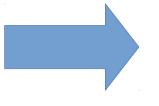


Models

Main ingredients

- Elasticity
- Disorder
- Drive

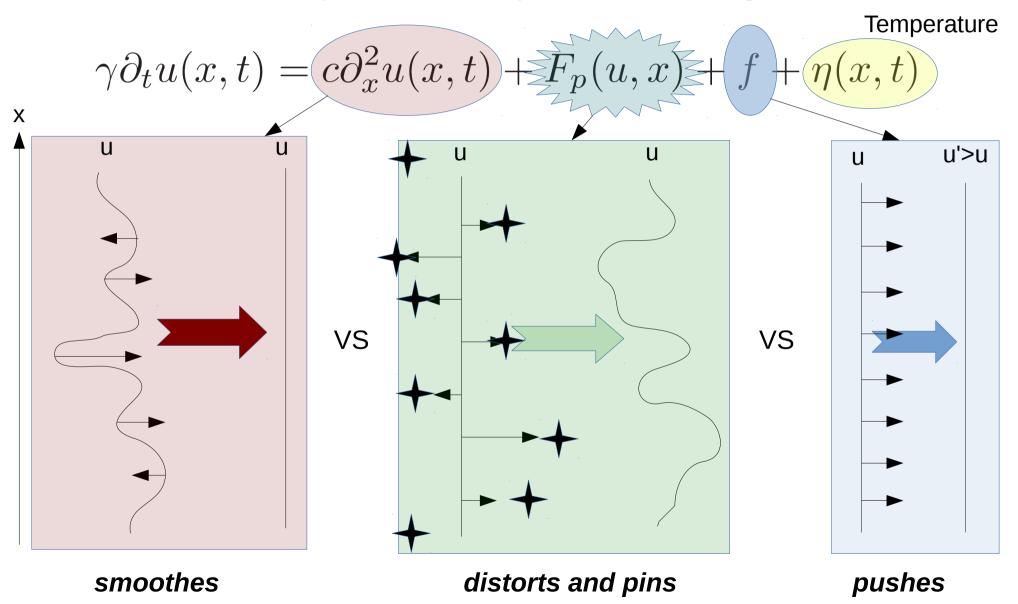
If dynamics has universal features, we should choose **minimal models** to predict them accurately.



Model to capture their universal interplay?

A minimal model?

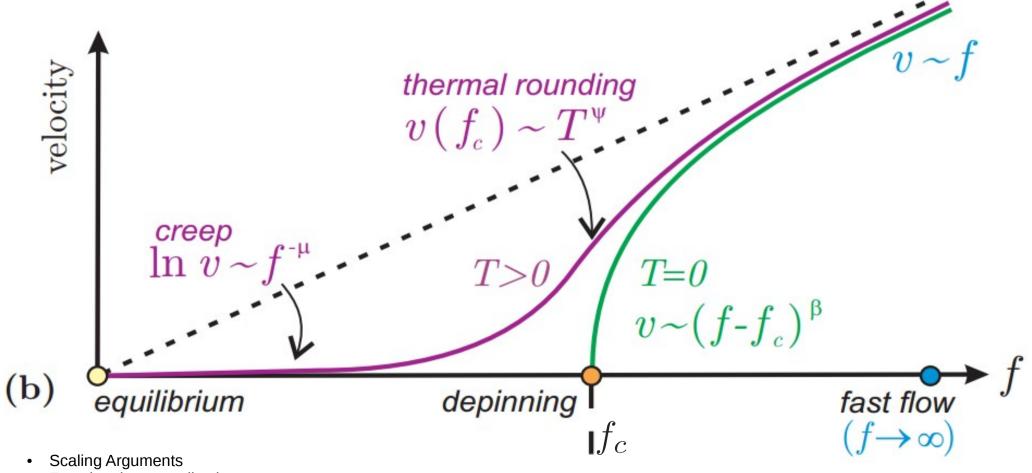
... to capture the competition between disorder and elasticity and to predict the response to a driving field



[Note: far more simple than a micro-magnetic model... too simple?]

Predictions from the minimal model: Dynamical Regimes

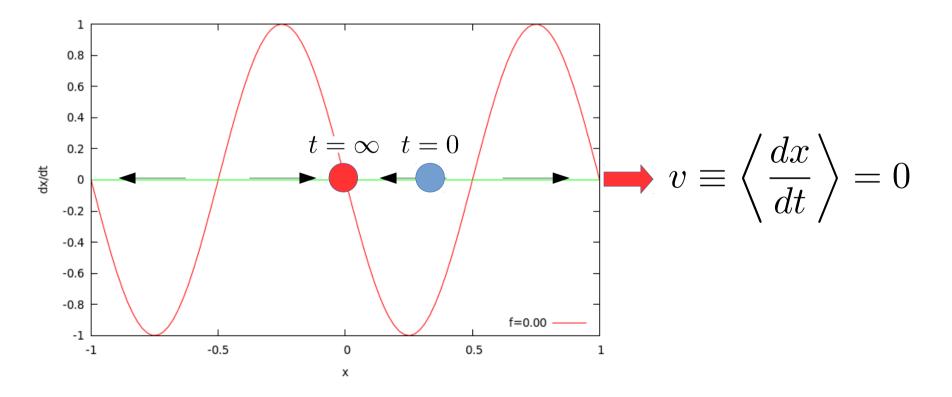
 $\gamma \partial_t u(x,t) = c \partial_x^2 u(x,t) + F_p(u,x) + f + \eta(x,t)$

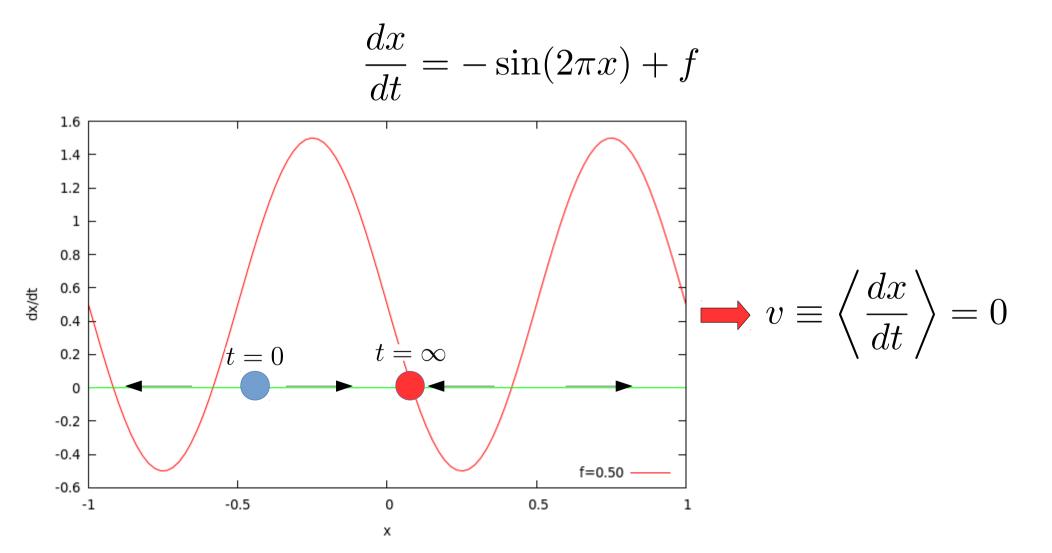


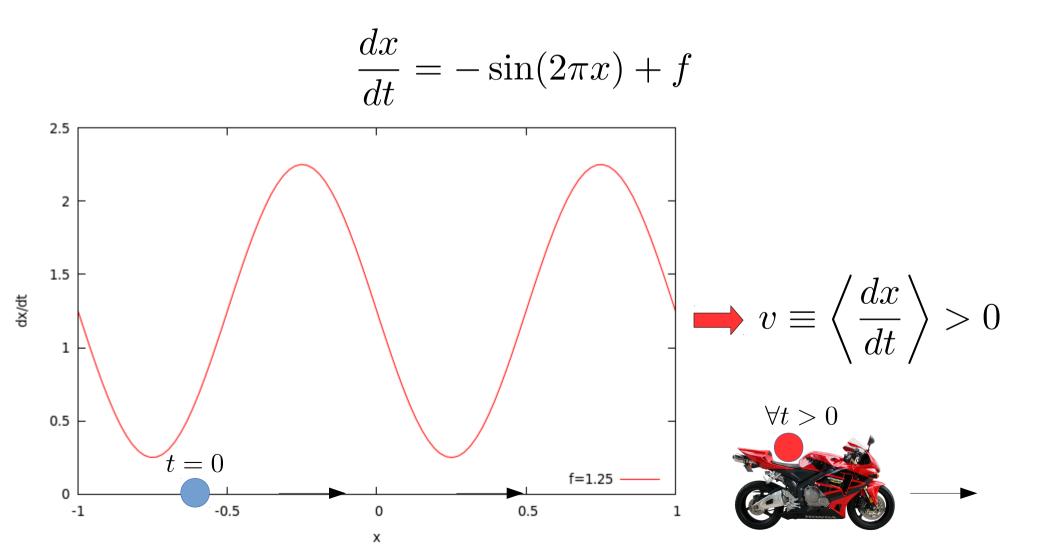
- Functional renormalization group
- Numerical Simulations

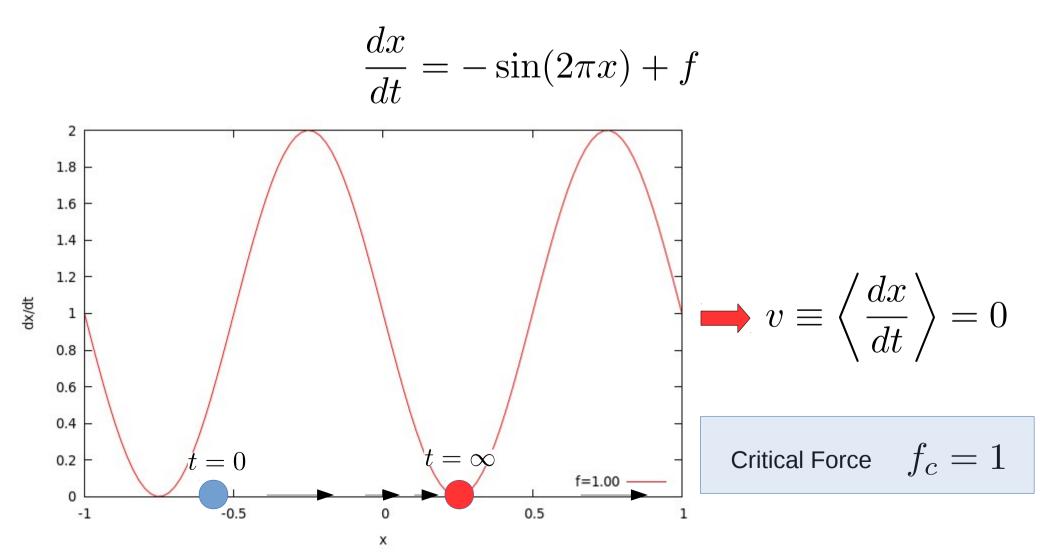
Three-reference steady states

$$\frac{dx}{dt} = -\sin(2\pi x) + f$$





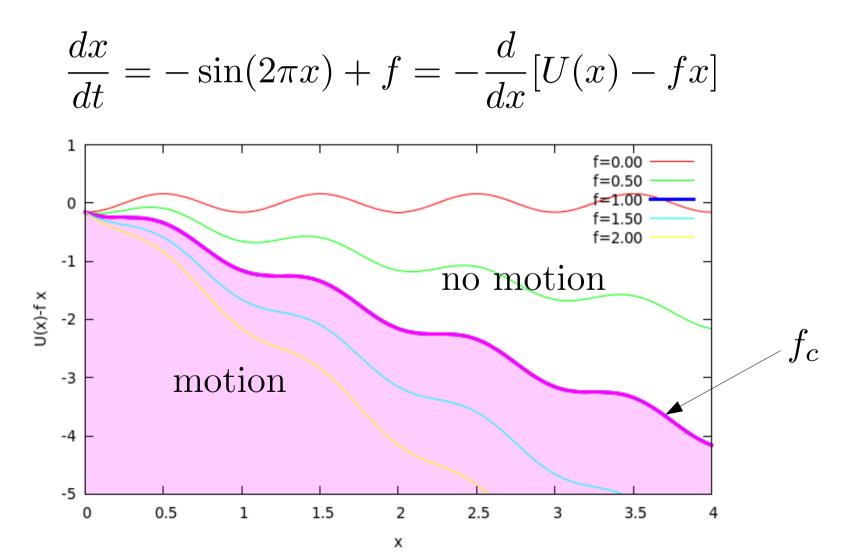


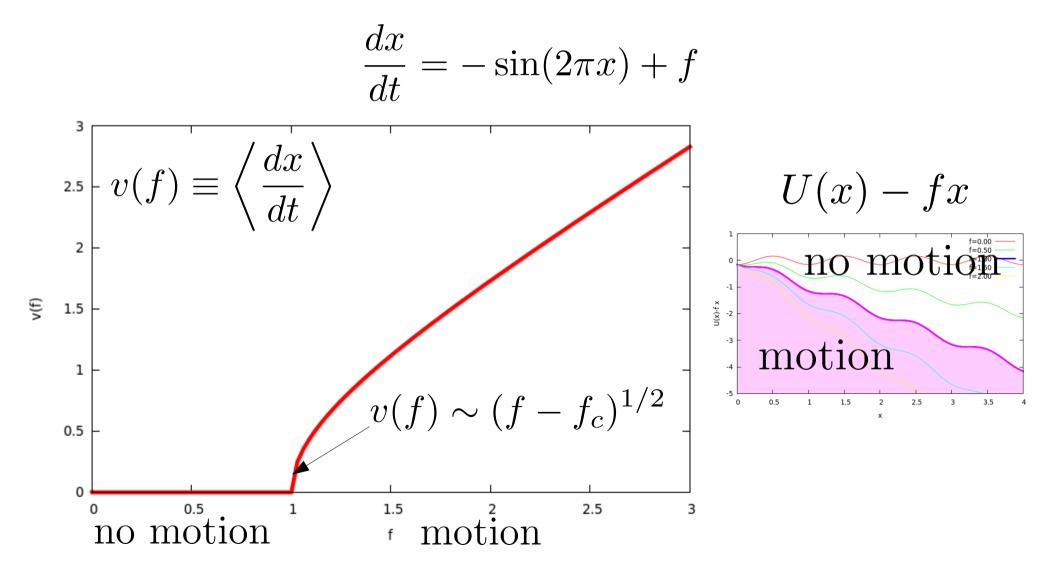


• One particle in a simple periodic potential

$$\frac{dx}{dt} = -\sin(2\pi x) + f = -\frac{d}{dx}U(x) - fx$$

х

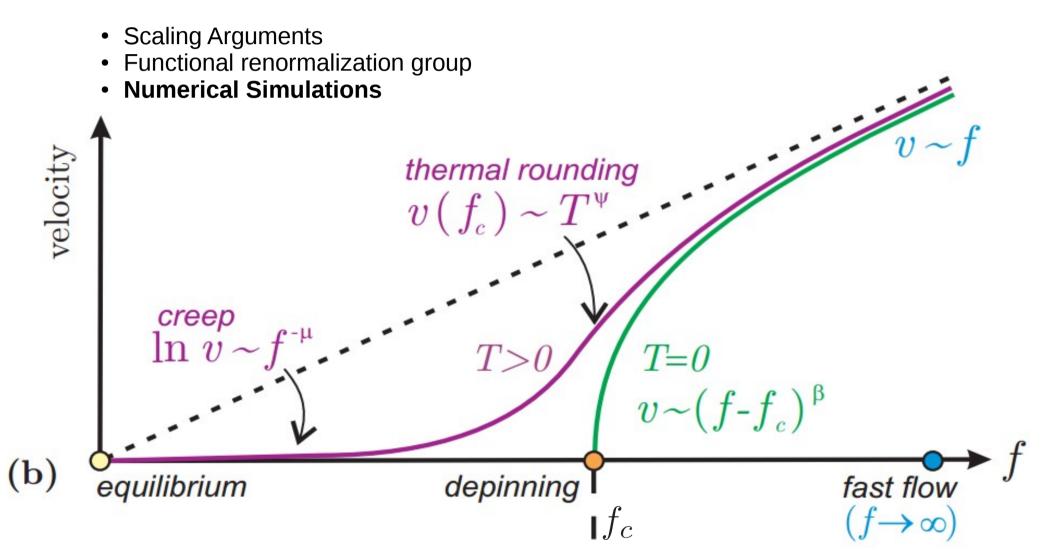




Equation of motion
$$\frac{dx}{dt} = -\sin(2\pi x) + f$$

Time Period $\tau = \int_0^1 \frac{dx}{f - \sin(2\pi x)} = \frac{1}{\sqrt{f^2 - 1}}$
Velocity-force $v = \langle \frac{dx}{dt} \rangle = \tau^{-1} = \sqrt{f^2 - 1}$
"critical" behaviour Near the threshold $f \to f_c^+$ $\tau \approx (f - f_c)^{-1/2}$

Predictions from the Elastic String model: Dynamical Regimes



Three-reference steady states

Thermal Effects

• Particle in an arbitrary one dimensional potential

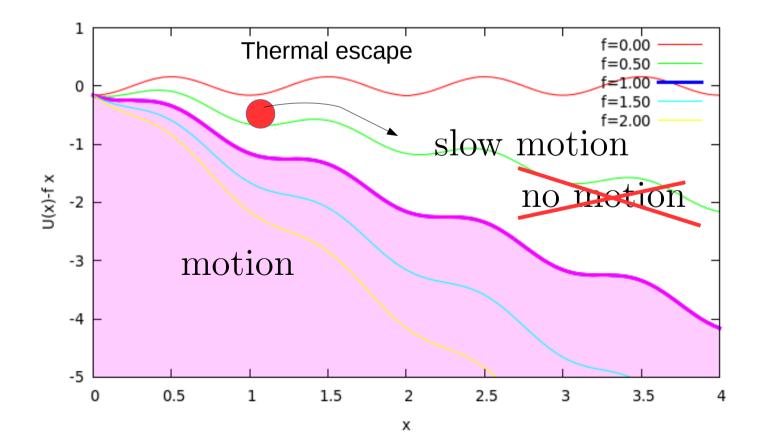
$$\frac{dx}{dt} = -\frac{dU}{dx} + f + \eta(t)$$
 Thermal noise
$$\langle \eta(t)\eta(t')\rangle = 2T\delta(t-t')$$

General solution for the mean velocity

$$v = \frac{T}{\int_0^\infty dz \; e^{-fz/T} \langle e^{\frac{U(x+z)-U(x)}{T}} \rangle_x}$$

Thermal Effects

$$\frac{dx}{dt} = -\sin(2\pi x) + f + \eta(t)$$
 Thermal noise



Thermal Effects

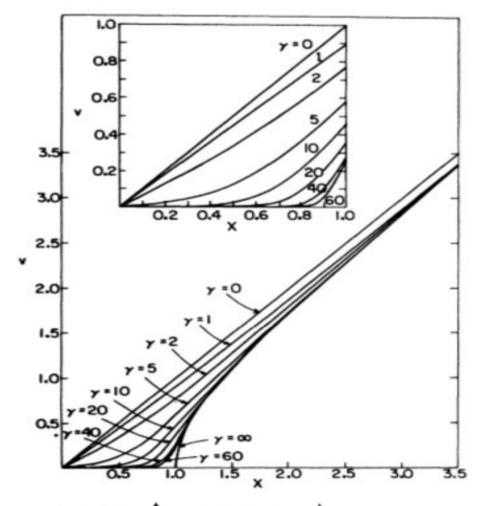
$$U(x) = -\frac{\cos(2\pi x)}{2\pi}$$

$$\frac{dx}{dt} = -\sin(2\pi x) + f + \eta(t)$$

$$v(f,T) = \frac{2T\sinh(f/2T)}{|I_{if/2T}(1/2\pi T)|^2}$$

Thermal Effects

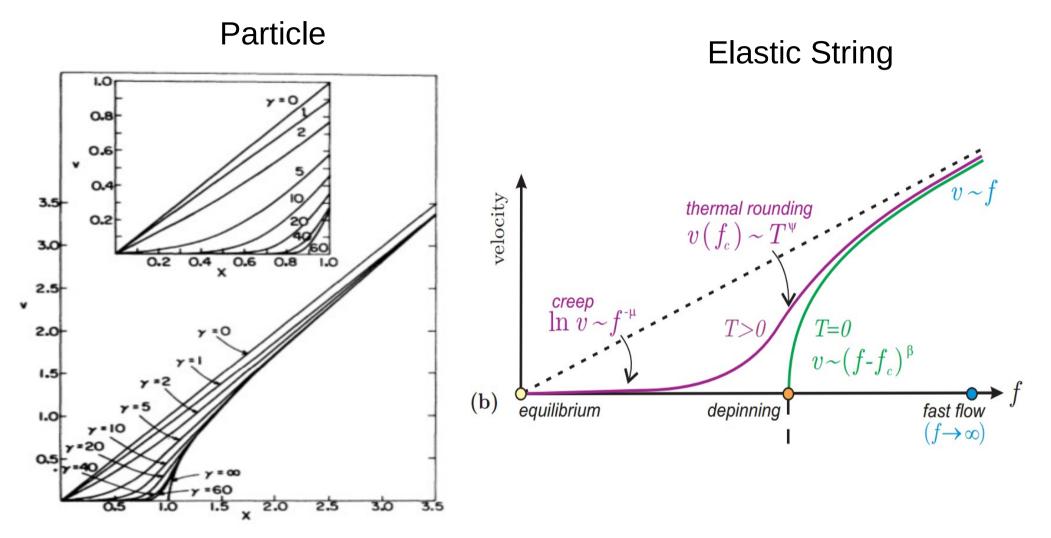
One particle in a simple periodic potential



 $\frac{dx}{dt} = -\sin(2\pi x) + f + \eta(t)$ $f_c = 1$ $v(f \gtrsim f_c, T = 0) \sim (f - f_c)^{1/2}$ $v(f = f_c) \sim T^{1/3}$ $v(f \ll f_c, T) \sim f e^{-U_c/T}$

A. R. Bishop[†] and S. E. Trullinger[‡]

Similar dynamical Regimes



A. R. Bishop[†] and S. E. Trullinger[‡]

Does this means that the Elastic Interface Model phenomenology can be understood with a particle model?

Particle vs String in random media

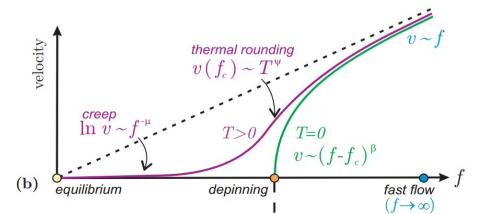
 One particle in a simple periodic or bounded random potential

$$v(f = f_c) \sim T^{1/3}$$

$$v(f \gtrsim f_c, T = 0) \sim (f - f_c)^{1/2}$$

$$v(f \ll f_c, T) \sim f e^{\underbrace{U_c/T}}$$

$$\psi = 0.33, \ \beta = 0.5, \ \mu = 0$$



 Elastic String in a Random potential

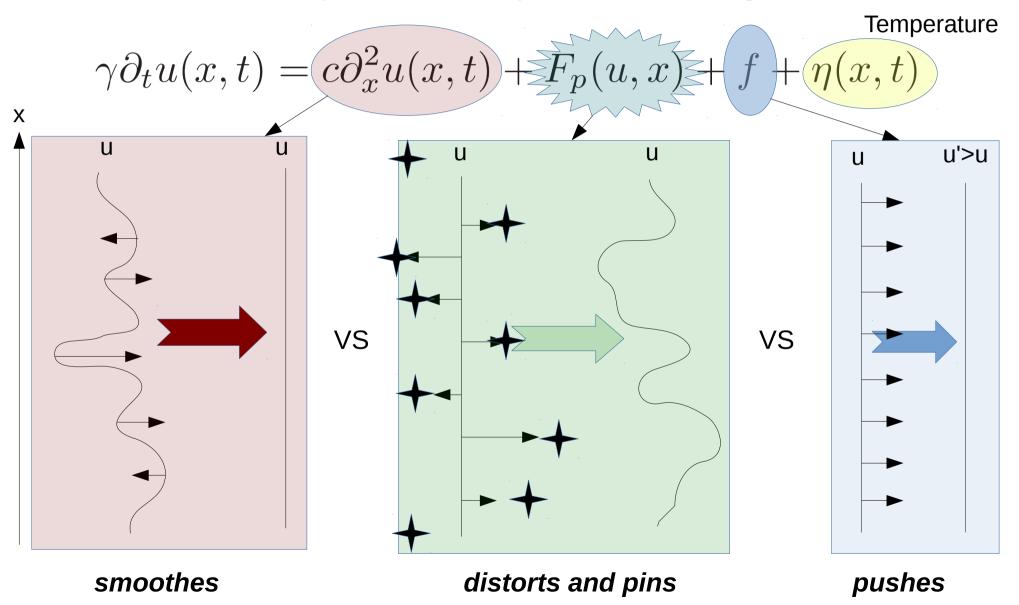
$$\begin{aligned} v(f = f_c) \sim \mathbf{T} \\ v(f \gtrsim f_c, T = 0) \sim (f - f_c) \\ v(f \ll f_c, T) \sim e^{-U_c(f/f_c)} \mathbf{T} \\ \psi = 0.15, \ \beta = 0.245, \ \mu = 1/4 \end{aligned}$$

Qualitatively similar, but not quantitatively. What is producing the difference?.

COLLECTIVE DYNAMICS

A minimal non trivial model

... to capture the competition between disorder and elasticity and to predict the response to a driving field



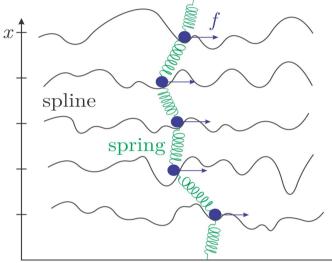
[Note: far more simple than a micro-magnetic model... too simple?]

Collective dynamics



Numerical Implementation

$$\gamma \partial_t u(x,t) = c \partial_x^2 u(x,t) + F_p(u,x) + f + \eta(x,t)$$



 $x = X\Delta x, \ t = n\Delta t, \ u$ continua $F_p(u, x) \to F_p(u, X) = \text{random splines}$ $\eta(x, t) \to \eta(X, n) = \text{random numbers}$

interface~directed polymer^u

simplest: finite differences, explicit Euler method $F_{tot}(X,n) = C[u(X-1,n) + u(X+1,n) - 2u(X,n)]$ $+F_p[u(X,n),X] + f + \sqrt{\frac{2T}{\Delta t}}R(X,n)$ $u(X,n+1) = u(X,n) + F_{tot}(X,n)\Delta t$

Numerical Implementation

$$n = 0, 1, ..., \infty$$

$$X = 0, 1, ..., L - 1$$

$$F_{tot}(X, n) = C[u(X - 1, n) + u(X + 1, n) - 2u(X, n)]$$

$$+F_p[u(X, n), X] + f + \sqrt{\frac{2T}{\Delta t}}R(X, n)$$

$$X = 0, 1, ..., L - 1$$

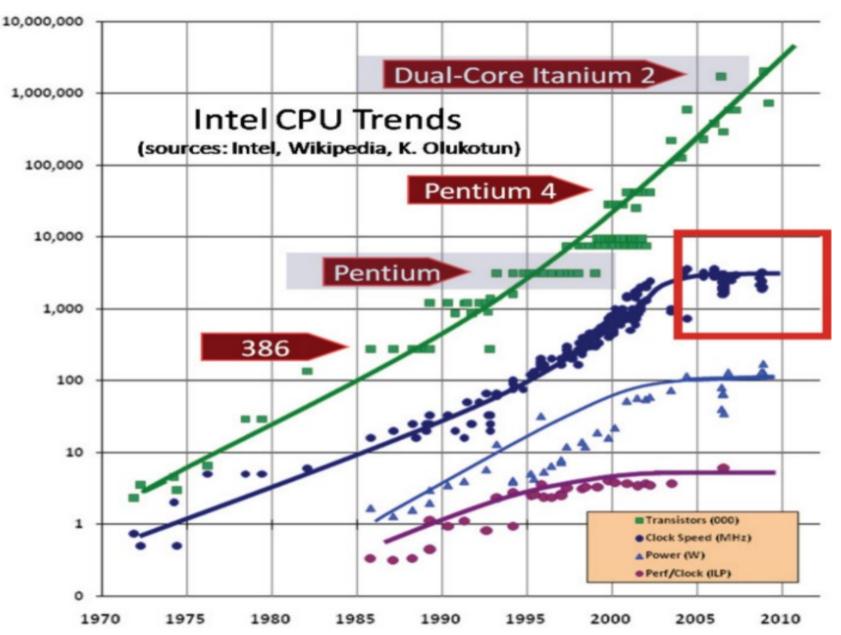
$$u(X, n + 1) = u(X, n) + F_{tot}(X, n)\Delta t$$

Numerical challenge:

to get accurate results we need to simulate large system sizes

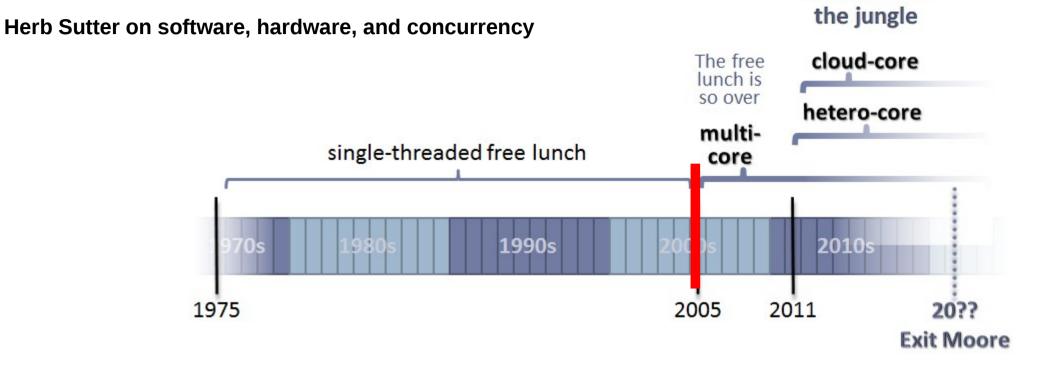
 \rightarrow Buy a faster CPU?

"Computers no longer get faster, just wider"



http://www.drdobbs.com/web-development/a-fundamental-turn-toward-concurrency-in/184405990

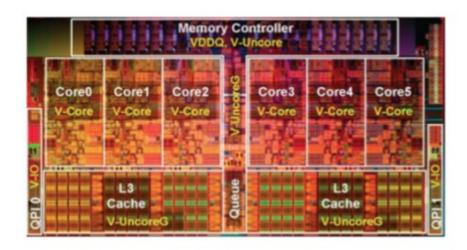
"For the first time in the history of computing, mainstream hardware is no longer a single-processor von Neumann machine, and never will be again."



- From 1975 to 2005 our industry put a **personal computer** on every desk, in every home, and in every pocket.
- In 2005, however, mainstream computing hit a wall.... From 2005 to 2011 our industry undertook a new mission: to put a personal parallel supercomputer on every desk, in every home, and in every pocket.
- From 2005 to 2011 we put a **personal** *parallel* **supercomputer** on every desk, in every home, and in every pocket (multicore tablets, smartphones, etc)

http://www.drdobbs.com/parallel/welcome-to-the-parallel-jungle/232400273?queryText=free%2Blunch

Many cores in just one machine



Intel Core i7-980X Extreme 6 cores

1.17B transistors

@3.33GHz, ~100 GFlops

Bandwidth 25.6Gb/s

TDP 130W

~USD 1000

NVIDIA GTX 580 SC 512 cores **3B** transistors

建設 建装置器 经财利利益 经财富 经公司公司

@1.54GHz, 1581 GFlops Bandwidth 192.4Gb/s **TDP 244W** ~USD 500

"green HPC"

Hybrid programming

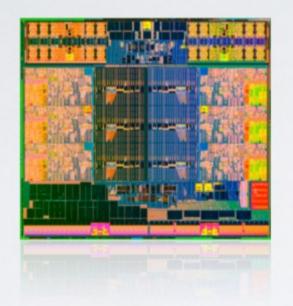
http://en.wikipedia.org/wiki/Transistor_count

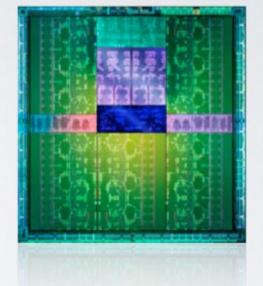
"Task Paralellism" (each thread a different task) "Data Parallelism"

(each thread same task, different data)

How to use this computing power?

Intel i7-4960x 1 860 M 6 Cores





NVIDIA Kepler 7 100 M 2688 Cores

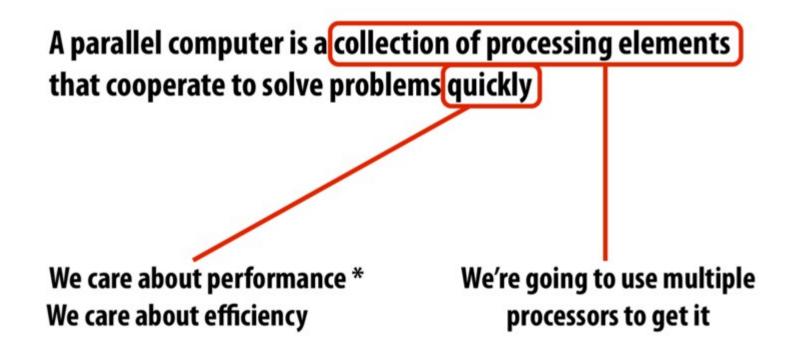
@3.9 GHz, ~100 GFlops Bandwidth 25.6Gb/s Bandwidth 192.4Gb/s TDP 130W ~USD 1000

@1.54GHz, 1.3 TFlops TDP 244W ~USD 900

We must rethink our algorithms to be parallel ...

Parallel Computing

One common definition

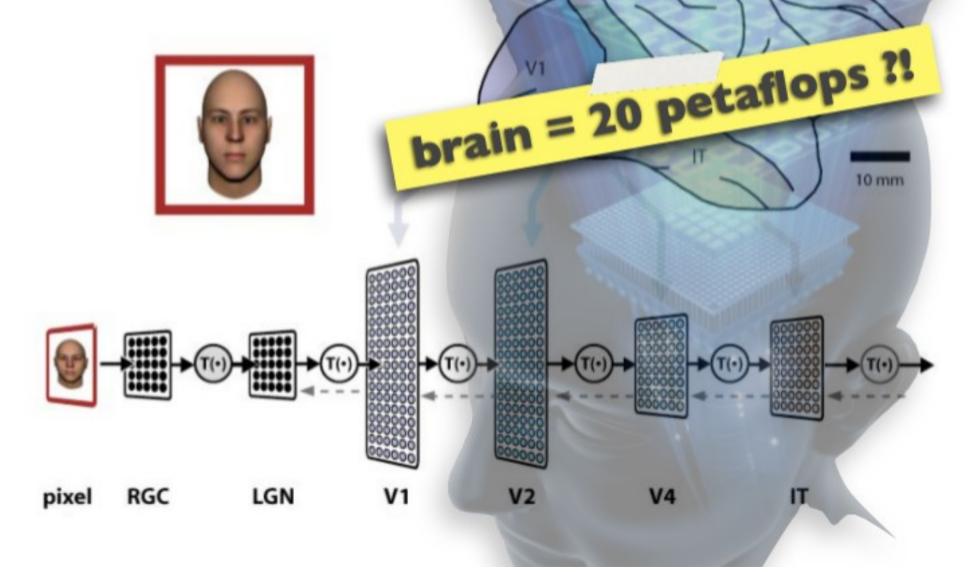


Parallel Computing, socially...



Parallel Computing, biologically... The brau ~50% of that is for vision! Nicolas Pinto (MIT), Cordoba 2013

Reverse Engineering The Ventral Visual Stream



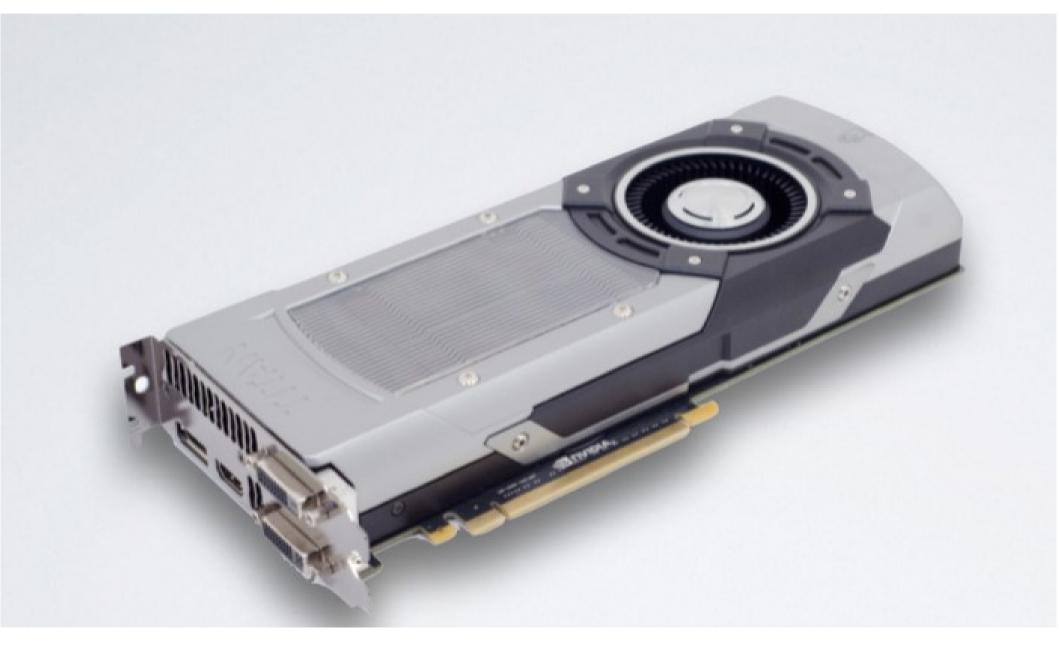
Nicolas Pinto (MIT), WHPC 2013, Cordoba

Parallel processing performance vs power

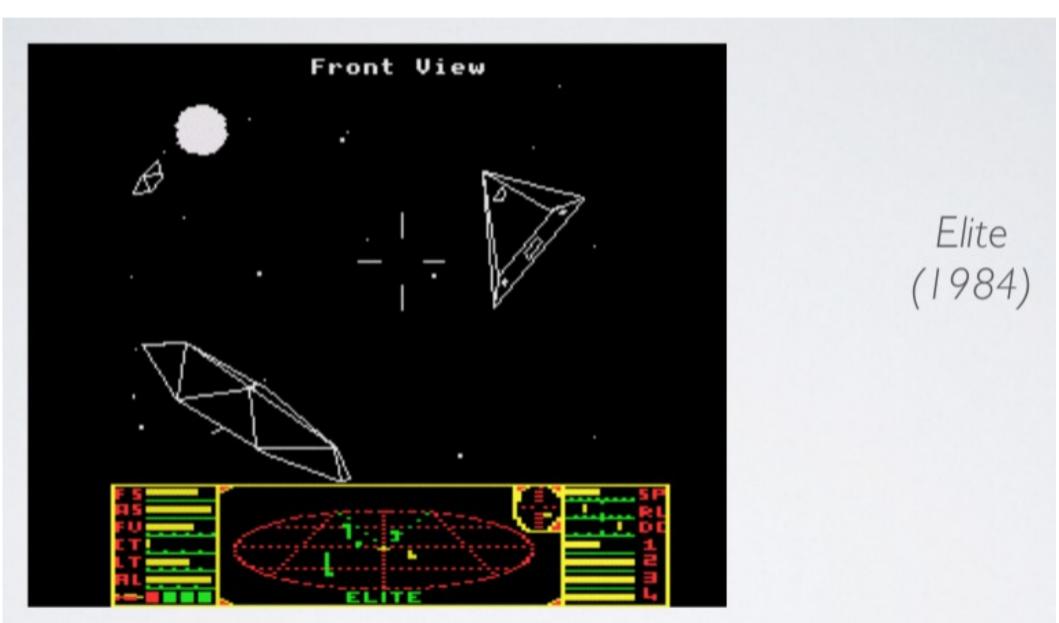
	JUNE 2		presented by <u>UNIVERSITY O</u> MANN	F NHEIM		BERKELEY LAB	Lawrence Berkele National Laborato	ey ery	FIND OUT I www.toj	
	NAME	SPECS				SITE	COUNTRY	CORES	R _{MAX} PFLOP/S	POWER MW
1	Tianhe-2 (Milkyway-2)	2) NUDT, Intel Ivy Bridge (12C, 2.2 GHz) & Xeon Phi (57C, 1.1 GHz), Custom interconnect			NUDT	China	3,120,000	33.9	17.8	
2	Titan	Cray XK7, Opteron 6274 (16C, 2.2 GHz) + Nvidia Kepler (14C, .732 GHz), Custom interconnect			DOE/SC/ORNL	USA	560,640	17.6	8.3	
3	Sequoia	IBM BlueGene/Q, Power BQC (16C, 1.60 GHz), Custom interconnect			DOE/NNSA/LLNL	USA	1,572,864	17.2	7.9	
4	K computer	Fujitsu SPARC64 VIIIfx (8C, 2.0GHz), Custom interconnect			RIKEN AICS	Japan	705,024	10.5	12.7	
5	Mira	IBM BlueGene/Q, Power BQC (16C, 1.60 GHz), Custom interconnect			DOE/SC/ANL	USA	786,432	8.16	3.95	
				perfo Name	nputer ormance FLOPS OPS 10 ²⁴	brain	n = 2	20V	Rop	10 mm
		Petaflops	Watts	exaFLO		2000		200		
Tia	anhe-2	33.9	17800000	teraFLO			→ (1(·)-	► (T(·)		•((•)-
Br	ain (Visual)	20	20	gigaFLC megaFL kiloFLO	OPS 10 ⁶	V1	V2	V4	IT	

Nicolas Pinto (MIT), Cordoba 2013

GPUs: cheap and massively parallel!



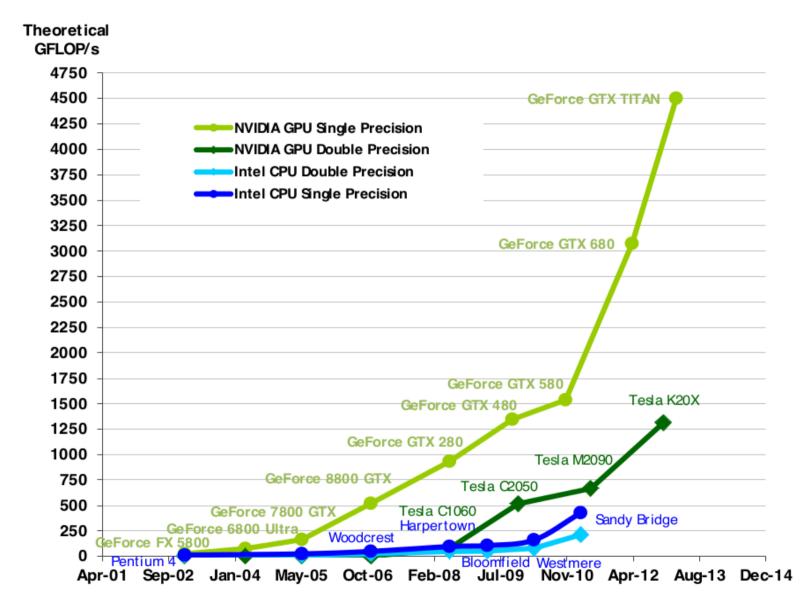
Computer games



Parallel computing for accelerating computer games



GPUs



If you write a scalable parallel program, next year will run faster...

How do we use this for science?



GPGPU for scientific applications

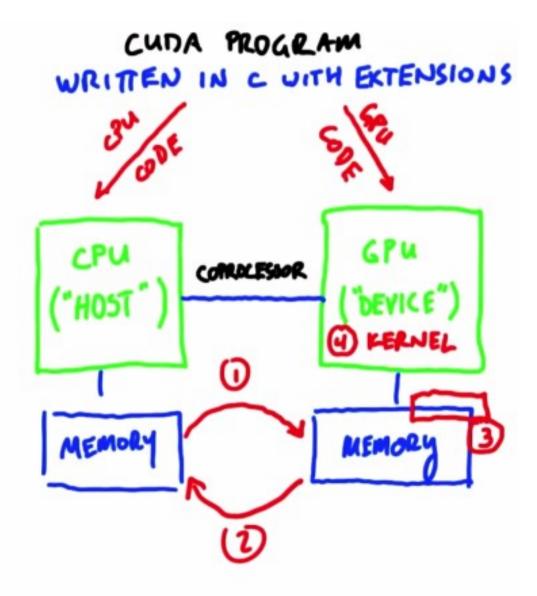
General-purpose computing on graphics processing units (GPGPU) is the utilization of a graphics processing unit (GPU), which typically handles computation only for computer graphics, to perform computation in applications traditionally handled by the central processing unit (CPU).

CUDA for scientific applications

• Before 2007 it was difficult... (only rendering oriented languages).

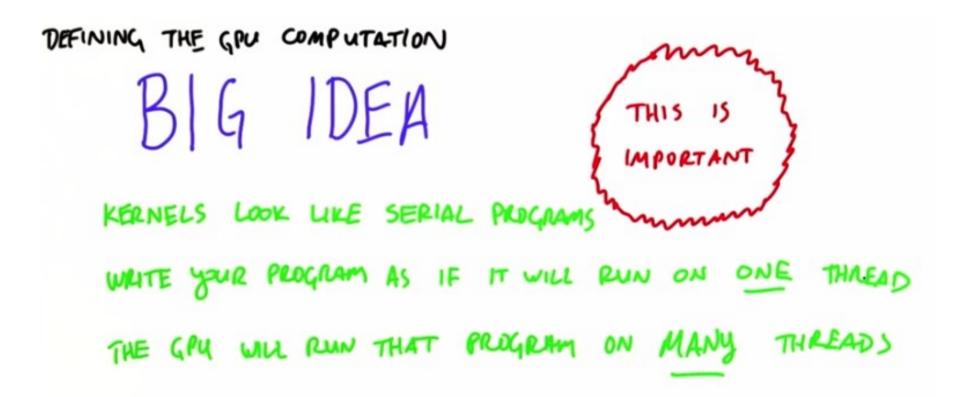
• After 2007:

CUDA, which stands for Compute Unified Device Architecture, is a parallel computing platform and programming model created by NVIDIA and implemented by the graphics processing units (GPUs) that they produce. CUDA gives developers direct access to the virtual instruction set and memory of the parallel computational elements in CUDA GPUs.



Intro to Parallel Programming, Using CUDA to Harness the Power of GPUs [UDACITY]

CUDA for scientific applications



Intro to Parallel Programming, Using CUDA to Harness the Power of GPUs [UDACITY]

CUDA for scientific applications

- Throughput vs Latency
 - Slow but many "workers" or threads...
- Data-parallel computing is the most scalable solution
 - Instead of modifying the code when we change the number of cores (2, 4, 8, 16...)
 - We build the computation around the data, as we will allways have more data than cores → Many threads doing the same computation on different data.
- Scalability !
 - from cell phones to supercomputers.
 - from current to future GPUs.

Very lucky problem for GPGPU!

Simples numerical implementation: explicit finite-difference

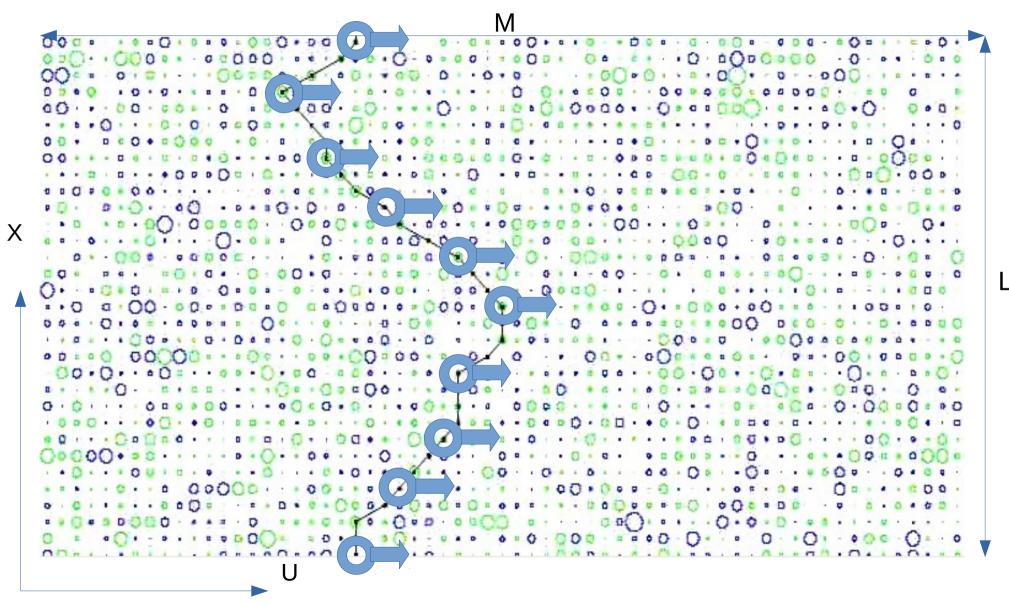
$$u(X, n+1) = u(X, n) + F_{tot}(X, n)\Delta t$$

$$F_{tot}(X,n) = C[u(X-1,n) + u(X+1,n) - 2u(X,n)] + F_p[u(X,n),X] + f + \sqrt{\frac{2T}{\Delta t}}R(X,n)$$

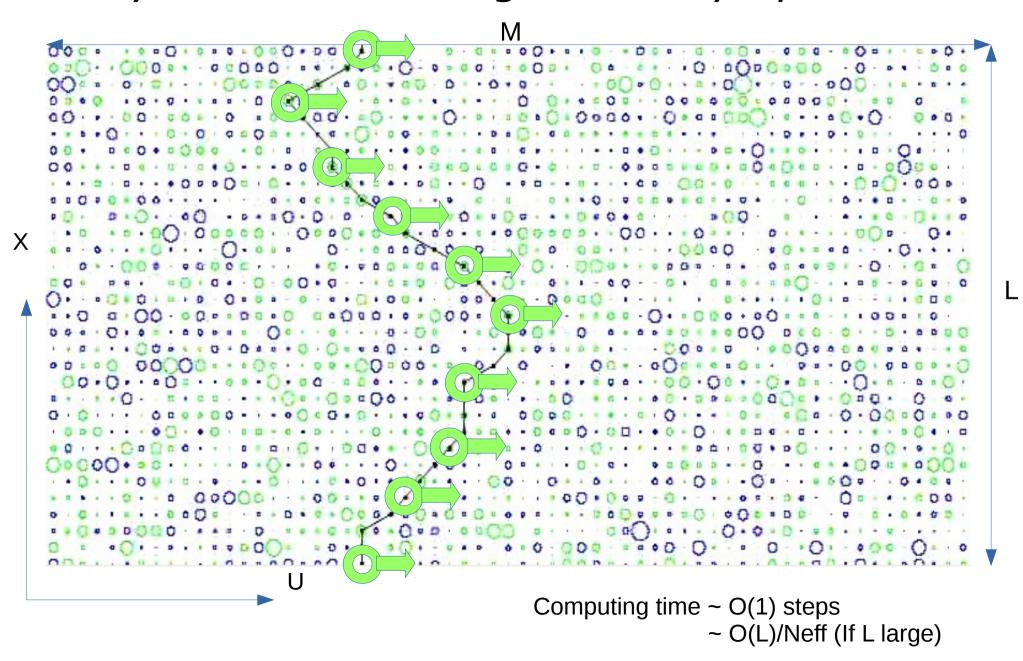
"Embarrasingly parallel problem"

- 1) "*Parallel For*" or "*transform*" (X=0,...,L-1) Thread "X" calculates force in X, Ftot(X) → Array "Ftot" (device memory)
- 2) "*Parallel For*" or "transform" (X=0,...,L-1) Thread "X" calculates $u(X,n+1) \rightarrow New$ "u" array (device memory)
- 3) "parallel reductions" : physical properties (or visualization...)
- 4) Time = Time + Dt ; Go to (1)

Serial time step: one "thread" in charge of many operations

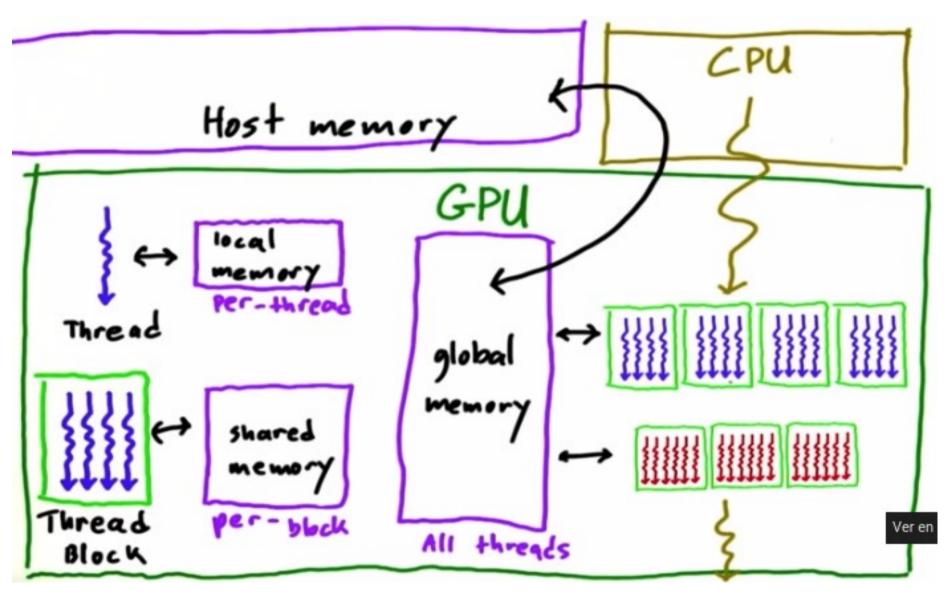


Parallel time step: Many threads in charge of many operations

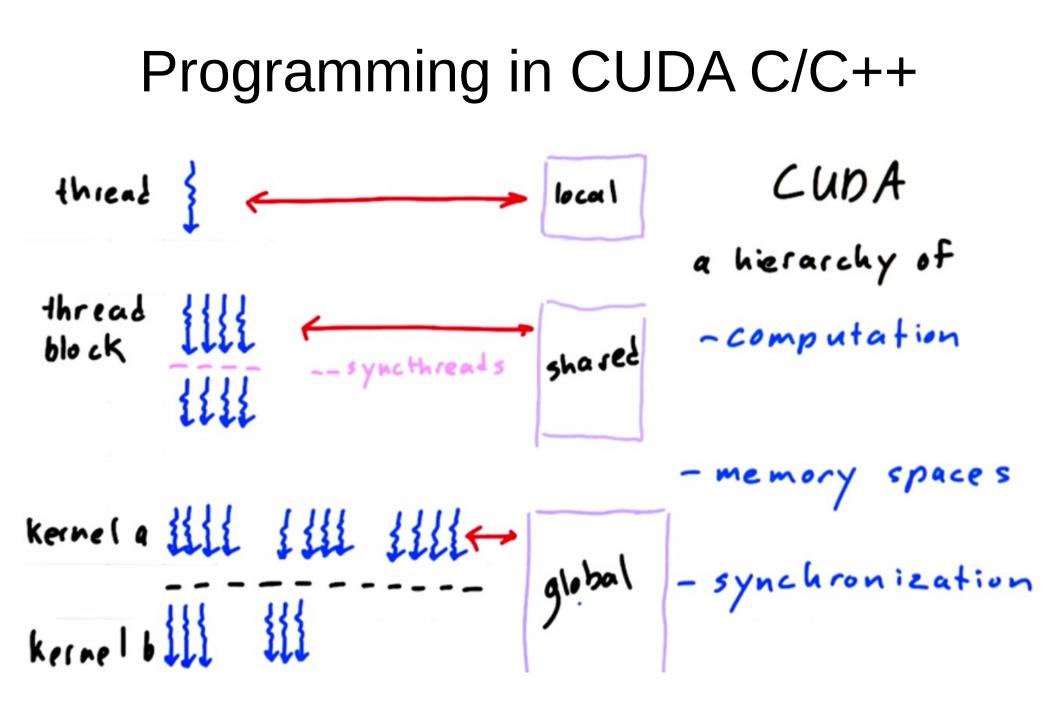


How do we code for many parallel threads in the GPU?

A low level interface: CUDA C/C++



Intro to Parallel Programming, Using CUDA to Harness the Power of GPUs [UDACITY]



Intro to Parallel Programming, Using CUDA to Harness the Power of GPUs [UDACITY]

Be careful with threads and communication patterns!

Example:

- Two threads have to increment by 1 an integer, which is zero initially
- Result must be 2, right?

Thread 1	Thread 2		Integer value	
			0	
read value		←	0	
increase value			0	
write back			1	
	read value	←	1	
	increase value		1	
	write back	→	2	

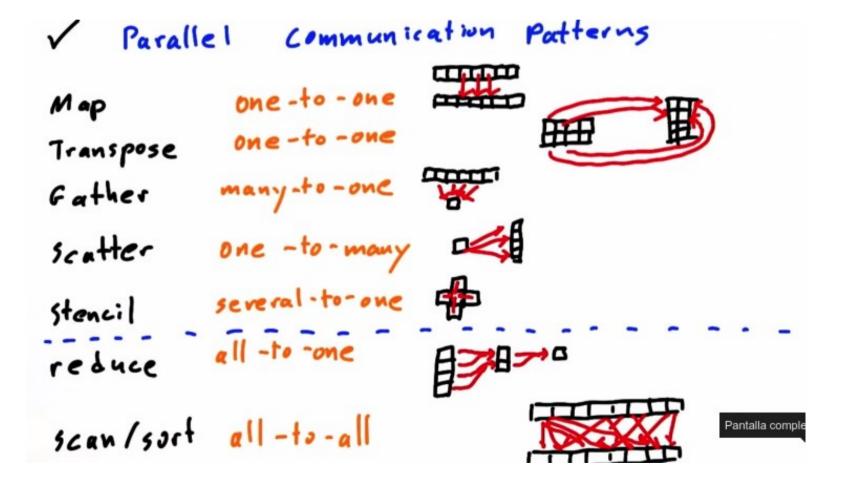
Thread 1	Thread 1 Thread 2		Integer value
			0
read value		←	0
	read value	←	0
increase value			0
	increase value		0
write back		→	1
	write back	\rightarrow	1

RACE CONDITIONS, and other problems we did not think about !! How to solve them, without killing too much the performance gain by parallelism?

Do not be scared!

there is a way to quickly start programming GPUs, and accelerate your old sequential codes :-)

Indentify common parallel patterns in your code....



Intro to Parallel Programming, Using CUDA to Harness the Power of GPUs [UDACITY]

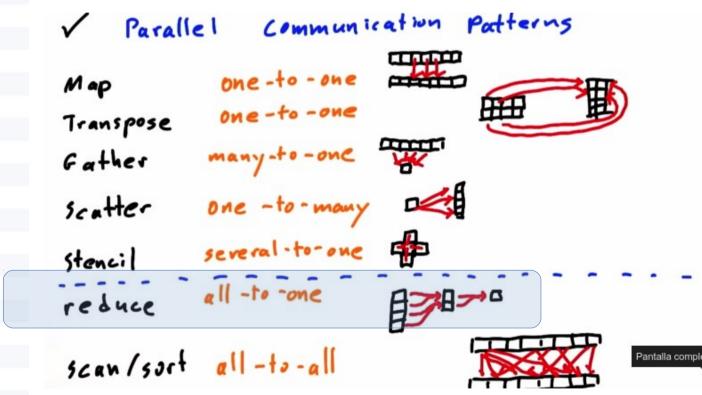
Ex: Thrust library

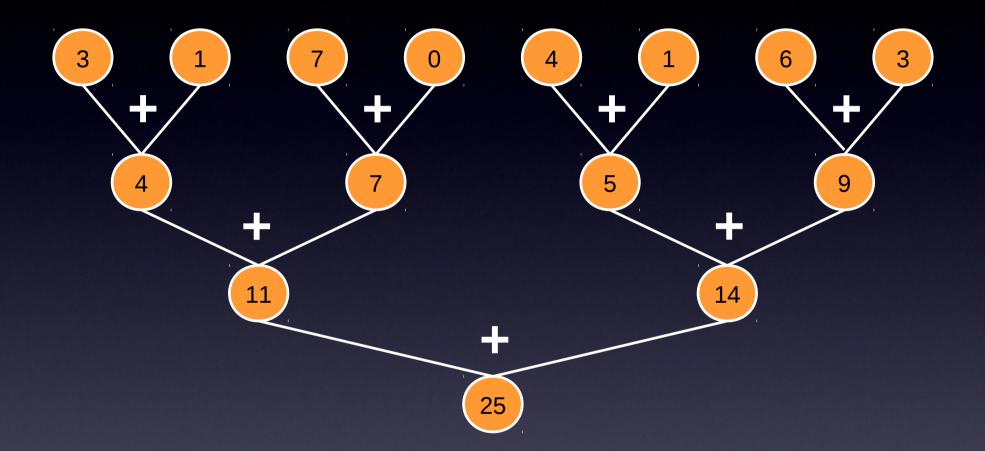
Algorithms Searching Binary Search Vectorized Searches Copying Gathering Parallel Scattering Reductions Map Counting Comparisons Transpose Extrema Father Transformed Reductions Logical Predicates Scatter Merging Reordering stencil Partitioning Stream Compaction Il -to -one reduce Prefix Sums Segmented Prefix Sums scan/sort all-to-all **Transformed Prefix Sums** Set Operations Sorting Transformations

Filling

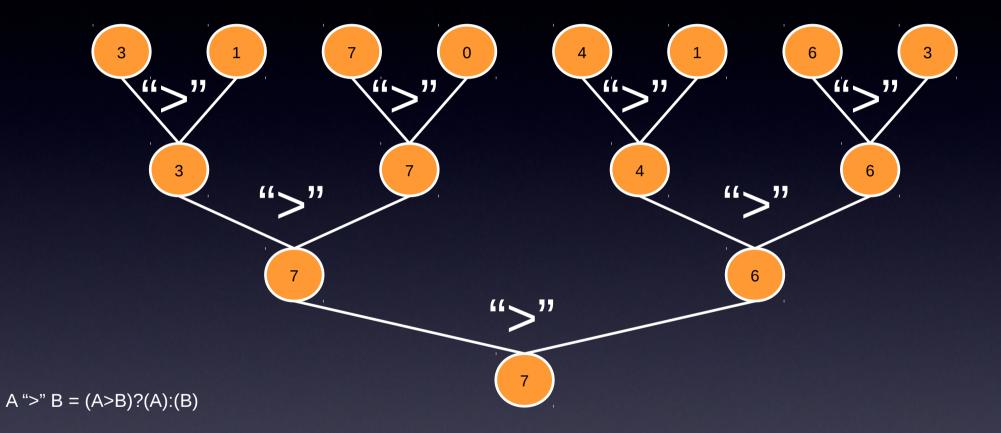
Modifying Replacing Intro to Parallel Programming, Using CUDA to Harness the Power of GPUs [UDACITY]

Look for a flexible, portable and expressive parallel library

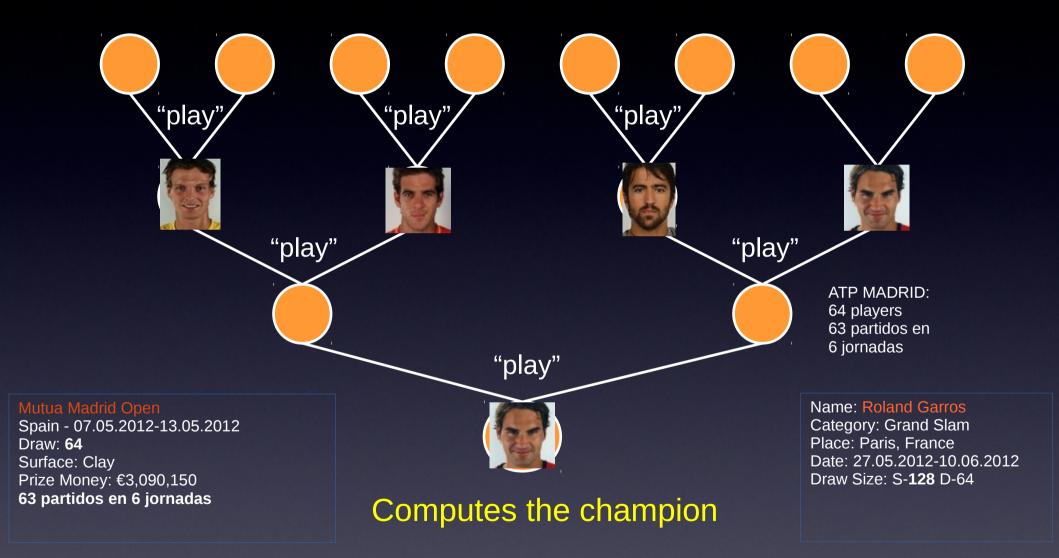




Parallel Reduction ~ $O(\log 2 N)$ steps Serial Reduction ~ O(N) steps (!)

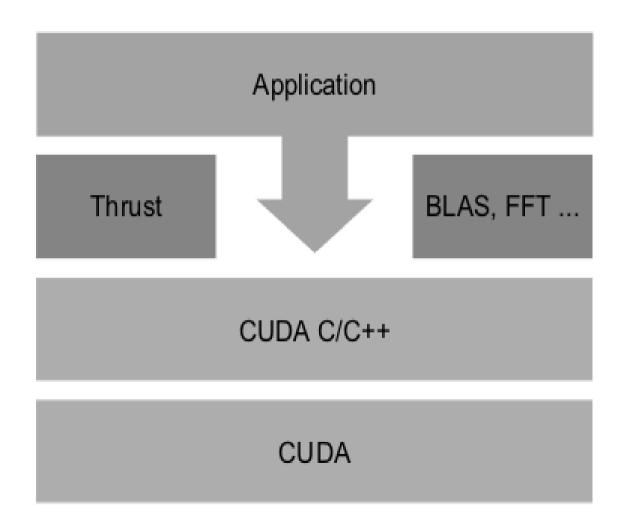


Finds the maximum



log_2(N) operations: Grand-Grand-Slam: $2^20 = 1048576$ players \rightarrow only 20 stages !!!

High Level Interface



http://thrust.github.io/

Fundamental Parallel Algorithms Libraries for scientists

C++ template library



GPU y CPU multicore

Random Numbers

Random123:

a Library of Counter-Based

Random Number Generators

GPU y CPU multicore

+ several others!

http://thrust.github.io/

https://www.deshawresearch.com/resources_random123.html

Thrust

Thrust Content from GTC 2012: Nathan Bell

What is Thrust?

- High-Level Parallel Algorithms Library
- Parallel Analog of the C++ Standard Template Library (STL)
- Performance-Portable Abstraction Layer
- Productive way to program CUDA

Thrust

Thrust Content from GTC 2012: Nathan Bell

GPU

Easy to Use

- Distributed with CUDA Toolkit
- Header-only library
- Architecture agnostic
- Just compile and run!
 - \$ nvcc -02 -arch=sm_20 program.cu -o program

Thrust

Thrust Content from GTC 2012: Nathan Bell

Productivity

- Containers
 - host_vector
 - device_vector

Memory Mangement

- Allocation
- Transfers
- Algorithm Selection
 - Location is implicit

// allocate host vector with two elements
thrust::host_vector<int> h_vec(2);

// copy host data to device memory
thrust::device_vector<int> d_vec = h_vec;

// write device values from the host
d_vec[0] = 27;
d_vec[1] = 13;

// read device values from the host
int sum = d_vec[0] + d_vec[1];

// invoke algorithm on device
thrust::sort(d_vec.begin(), d_vec.end());

// memory automatically released

Thrust High Level Interface

```
#include <thrust/host vector.h>
#include <thrust/device vector.h>
#include <thrust/generate.h>
#include <thrust/sort.h>
#include <thrust/copy.h>
#include <algorithm>
#include <cstdlib>
int main(void)
  // generate 32M random numbers serially
  thrust::host vector<int> h vec(32 << 20);
  std::generate(h vec.begin(), h vec.end(), rand);
  // transfer data to the device
  thrust::device vector<int> d vec = h vec;
  // sort data on the device (846M keys per second on GeForce GTX 480)
  thrust::sort(d vec.begin(), d vec.end());
  // transfer data back to host
  thrust::copy(d vec.begin(), d vec.end(), h vec.begin());
  return 0;
```

http://thrust.github.io/

Simulating a large elastic String in a Random Medium using the GPU

```
// #include<...>
using namespace thrust;
int main()
```



// vectors in GPU global memory
// declaration and allocation:
device_vector<REAL> u(L); // mi pared (discretizada)
device_vector<REAL> Ftot(L); // la fuerza sobre mi pared

// iterators on GPU global memory: they define ranges
// necessary for applying parallel algorithms
device_vector<REAL>::iterator u_begin = u.begin();
device_vector<REAL>::iterator Ftot_begin = Ftot.begin();

```
// temporal evolution
for(long n=0;n<Trun;n++)</pre>
```

```
....

// (1) parallel algorithm in the GPU \rightarrow Ftot(X,n)_____

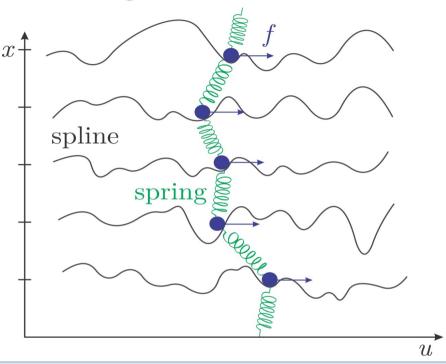
// (2) parallel algorithm in the GPU \rightarrow u(X,n+1) en la GPU

.....

// (3) parallel algorithm in the GPU \rightarrow properties

.....

// (4) Copies GPU -> CPU (reduced to a minimum)
```

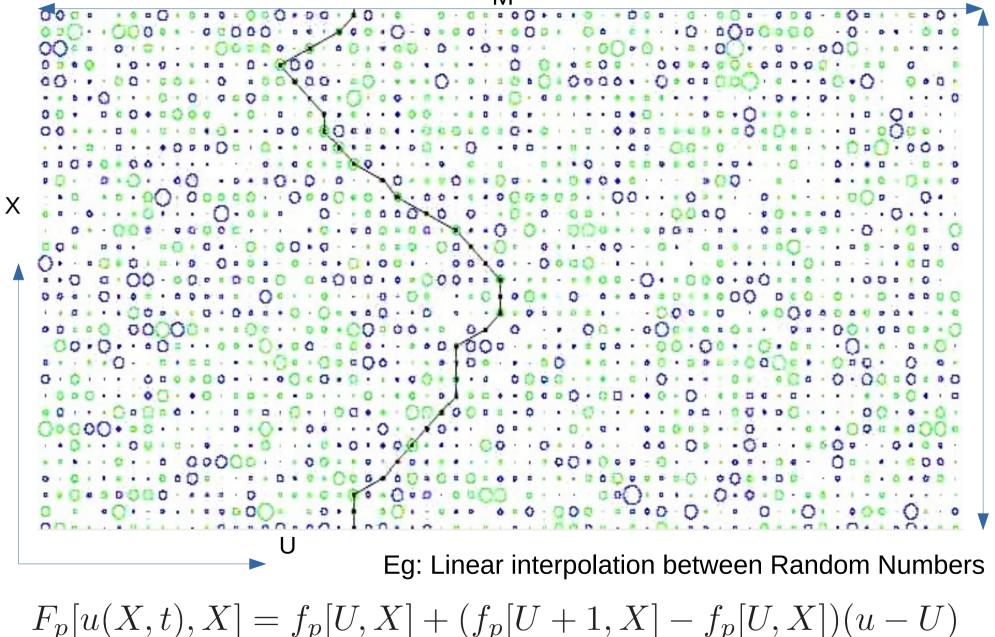


$$F_{tot}(X,n) = C[u(X-1,n) + u(X+1,n) - 2u(X,n)] + F_p[u(X,n),X] + f + \sqrt{\frac{2T}{\Delta t}}R(X,n)$$

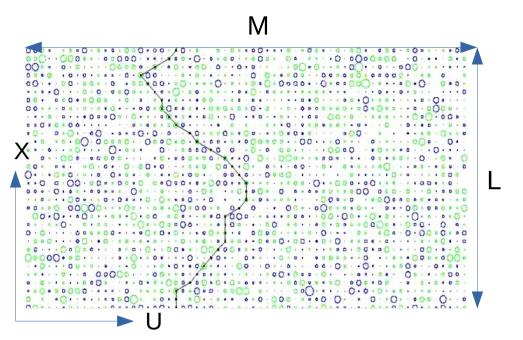
$$u(X, n+1) = u(X, n) + F_{tot}(X, n)\Delta t$$

Disordered landscapes

Μ



The problem with the disorder...



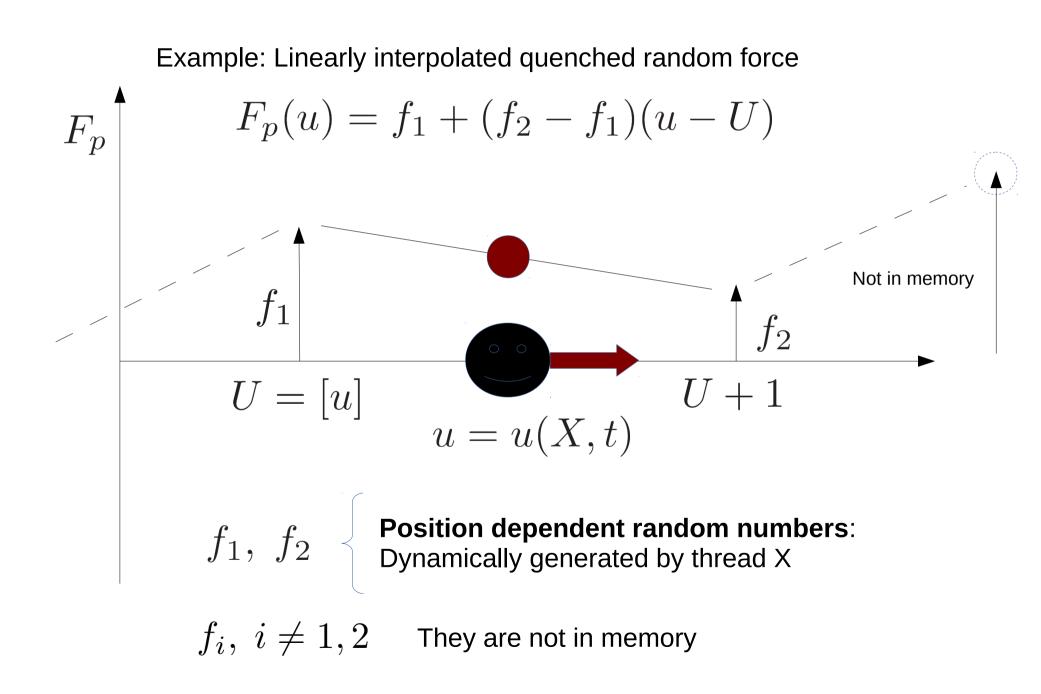
- Need L x M samples, with: L > 2^16 = 65536 y M ~ L^{5/4}~2^20.
- Disorder Matrix > 2^{16+20} floats
 → 256 Gb (!!).

koltona@mostro:/opt/cudasdk/C/bin/linux/release\$./deviceQuery

Device 1: "Tesla C2075" CUDA Driver Version / Runtime Version 4.2 / 4.1 CUDA Capability Major/Minor version number: 2.0 Total amount of global memory: 5375 MBytes (5636554752 bytes) (14) Multiprocessors x (32) CUDA Cores/MP: 448 CUDA Cores

- Save disorder in device memory is not an option!
- Simplest solution → *Dynamically Generate* using a parallel RNG.

Dynamically generated disorder



The European Physical Journal Special Topics August 2012, Volume 210, Issue 1, pp 53-71

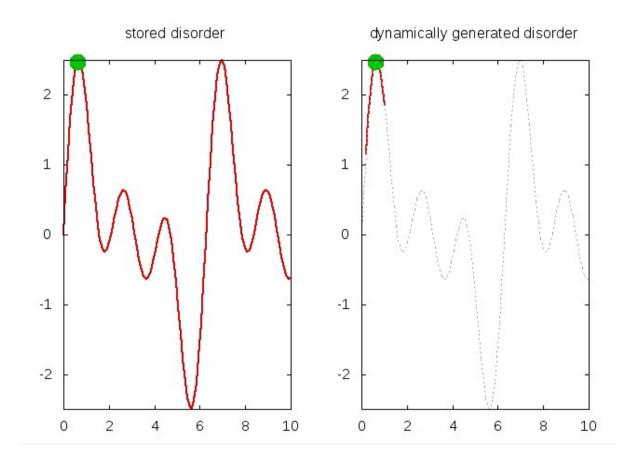
Random number generators for massively parallel simulations on GPU

M. Manssen, M. Weigel, A. K. Hartmann

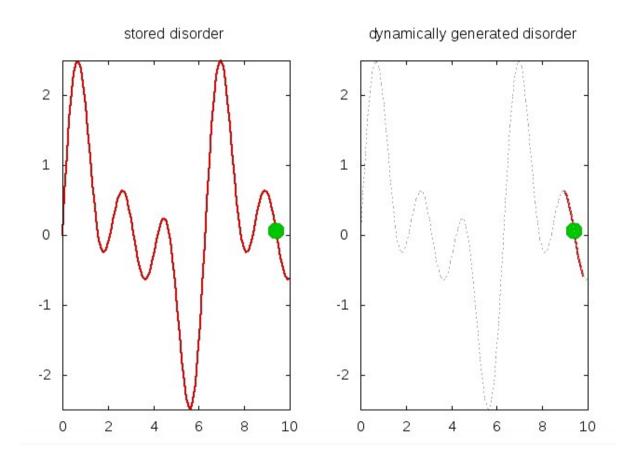
generator	bits/thread	failures in TestU01			Ising test	perf.
2746		SmallCrush	Crush		$\times 10^9/s$	
LCG32	32	12	10. TO	<u></u>	failed	58
LCG32, random	32	3	14	197 - 182)	passed	58
LCG64	64	None	6		failed	46
LCG64, random	64	None	2	8	passed	46
MWC	64 + 32	1	29		passed	44
Fibonacci, $r = 521$	≥ 80	None	2	<u> 20 - 20</u>	failed	23
Fibonacci, $r = 1279$	≥ 80	None	(1)	2	passed	23
XORWOW (cuRAND)	192	None	None	1/3	failed	19
MTGP (cuRAND)	≥ 44	None	2	2		18
XORShift/Weyl	32	None	None	None	passed	18
Philox4x32_7	(128)	None	None	None	passed	41
Philox4x32_10	(128)	None	None	None	passed	30

Which one?

Dynamically Generated quenched disorder



Dynamically Generated quenched disorder



Need for a reversible random generator

Climate (thermal noise) might not, but the landscape (quenched disorder) must be reversible!



Random number generators for massively parallel simulations on GPU

M. Manssen, M. Weigel, A. K. Hartmann

Regeneración sencilla de números ya generados?

generator	bits/thread	failu	failures in TestU01			perf.
			$\times 10^{9}/$			
LCG32	32	12			failed	58
LCG32, random	32	3	14		passed	58
LCG64	64	None	6		failed	46
LCG64, random	64	None	2	8	passed	46
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Fibonacci, $r = 521$	≥ 80	l o e	. 2	78	failed	23
Fibonacci, $r = 1279$	≥ 80	None	(1)	2	passed	23
XORWOW (cuRAND)	192	None	None	1/3	failed	19
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Philox4x32_10	(128)	None	None	None	passed	30

Which one?

Counter-Based RNGs



Parallel Random Numbers: As Easy as 1, 2, 3

John K. Salmon, Mark A. Moraes, Ron O. Dror, David E. Shaw

".. We demonstrate that independent, **keyed transformations of counters** produce a large alternative class of PRNGs with excellent statistical properties... are ideally suited to modern multicore CPUs, GPUs, clusters, and special-purpose hardware because they vectorize and parallelize well, and require little or no memory for state. ... All our PRNGs pass rigorous statistical tests and produce at least 2^64 unique parallel streams of random numbers, each with period 2^128 or more. In addition to essentially unlimited parallel scalability, our PRNGs offer excellent single-chip performance: Philox is faster than the CURAND library on a single NVIDIA GPU.

Counter-Based RNGs

Overview

counter-based RNGs are stateless functions whose arguments are a counter, and a key and whose return value is the same type as the counter.

value = CBRNGname(counter, key)

The returned value is a *deterministic function of the key and counter*, i.e. a unique (counter, key) tuple will always produce the same result. The result is highly sensitive to small changes in the inputs, so that the sequence of values produced by simply incrementing the counter (or key) is effectively indistinguishable from a sequence of samples of a uniformly distributed random variable.

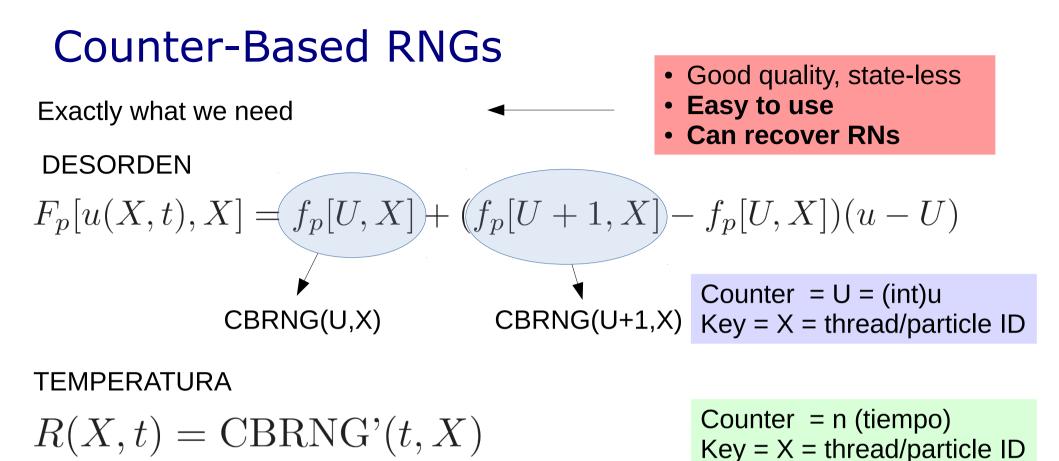
Exactamente lo que necesitamos!

$$F_p[u(X,t),X] = f_p[U,X] + (f_p[U+1,X]) - f_p[U,X])(u-U)$$

$$CBRNG(U,X) CBRNG(U,X)$$

$$Counter = U = (int)u$$

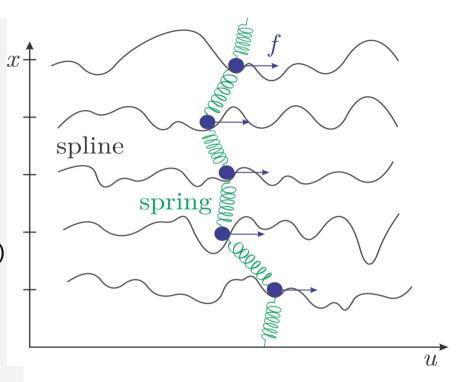
$$Key = X = thread/particle ID$$



Force update and Euler step on each particle

GPU fast implementation

```
// #include<...>
using namespace thrust;
int main()
  device vector<REAL> u(L);
  device vector<REAL> Ftot(L);
  device vector<REAL>::iterator u begin = u.begin();
  device vector<REAL>::iterator Ftot begin = Ftot.begin()
  for(long n=0;n<Trun;n++)</pre>
  ł
     // (1) parallel stencil \rightarrow Ftot(X,n) en la GPU
     ....
     // (2) parallel map \rightarrow u(X,n+1) en la GPU
     // (3) parallel reductions \rightarrow properties (GPU)
     // GPU -> CPU copies (reduced to a minimum)
```



GPU and multicore CPU parallel implementation

```
// #include<...>
using namespace thrust;
int main()
{
```

```
device_vector<REAL> u(L);
device_vector<REAL> Ftot(L);
```

```
device_vector<REAL>::iterator u_begin = u.begin();
  device_vector<REAL>::iterator Ftot_begin =
Ftot.begin();
```

```
for(long n=0;n<Trun;n++)
{
```

```
// (1) calculo Ftot(X,n) en la GPU
```

```
// (2) calculo u(X,n+1) en la GPU
```

```
// (3) calculo propiedades (GPU)
```

```
// GPU -> CPU copies (reduced to a minimum)
```

Same code!!

In GPU

nvcc -o stringCUDA string.cu

CPU-MULTICORE

- cp string.cu string.cpp
- g++ -o stringOMP string.cpp -O2 -fopenmp -DTHRUST_DEVICE_BACKEND=THRUST_ DEVICE_BACKEND_OMP -lgomp

Thrust Device backends http://thrust.github.io/

PORTABILITY any GPU, CPU Portable parallel performance

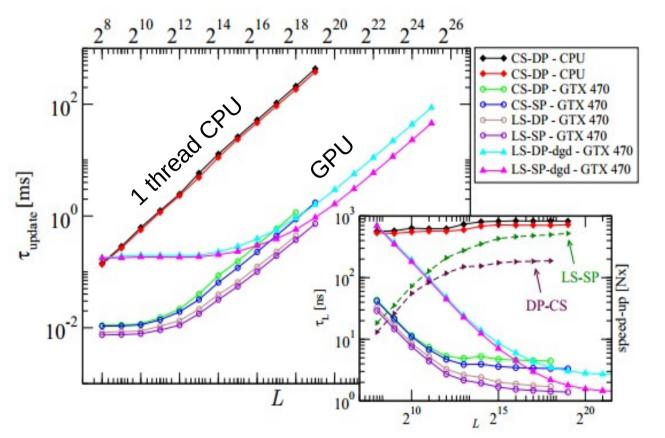
Real Time Simulation on this laptop

CUDA Device Query (Runtime API) version (CUDART static linking) Detected 1 CUDA Capable device(s)

Device 0: "GeForce GT 620M" CUDA Driver Version / Runtime Version 6.0 / 5.5 CUDA Capability Major/Minor version number: 2.1 Total amount of global memory: 1024 MBytes (1073479680 bytes) (2) Multiprocessors, (48) CUDA Cores/MP: 96 CUDA Cores GPU Clock rate: 1250 MHz (1.25 GHz) Memory Clock rate: 900 Mhz Memory Bus Width: 64-bit

ASUS Zenbook, ubuntu 14.04 Intel(R) Core(TM) i5-3317U CPU @ 1.70GHz

Performance !!!



- CPU: AMD Phenom(tm) II X4 955 Processor @3.2GHz
- **GPU:** NVIDIA GTX 470.

• 33554432 particles: largest model system to date.

Theoretical predictions:

Scaling corrections unveiled \rightarrow Most Accurate critical exponents for the depinning transition to date.

• Low cost !! (100-900 Euros) Same GPUs used for computer games.

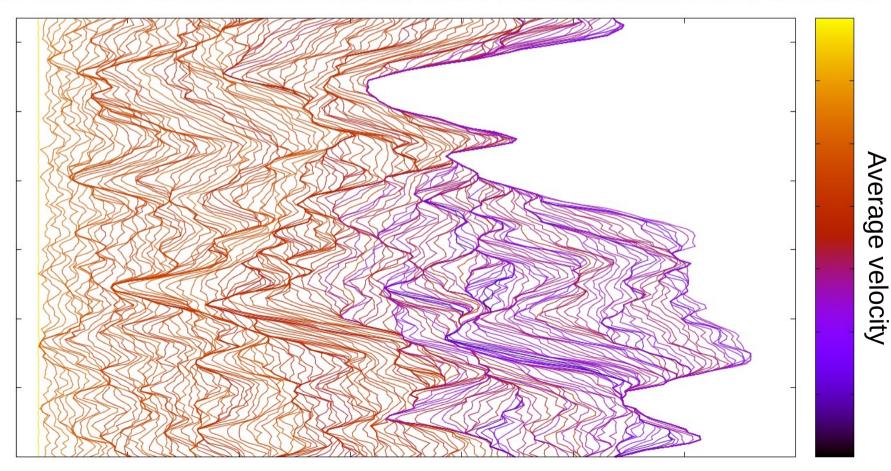
Theoretical Results

PHYSICAL REVIEW E 87, 032122 (2013)

Nonsteady relaxation and critical exponents at the depinning transition

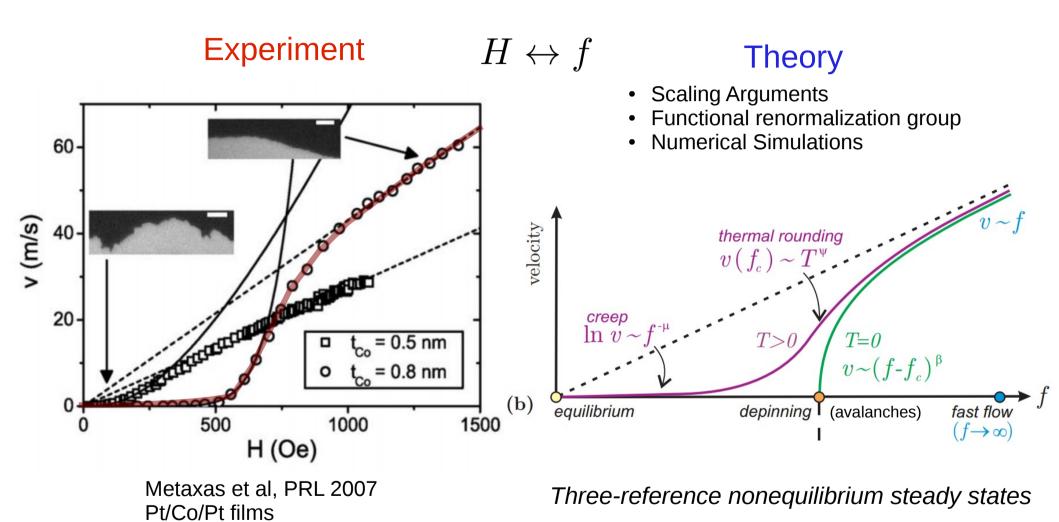
E. E. Ferrero, S. Bustingorry, and A. B. Kolton

CONICET, Centro Atómico Bariloche, 8400 San Carlos de Bariloche, Río Negro, Argentina (Received 28 November 2012; revised manuscript received 4 February 2013; published 11 March 2013)

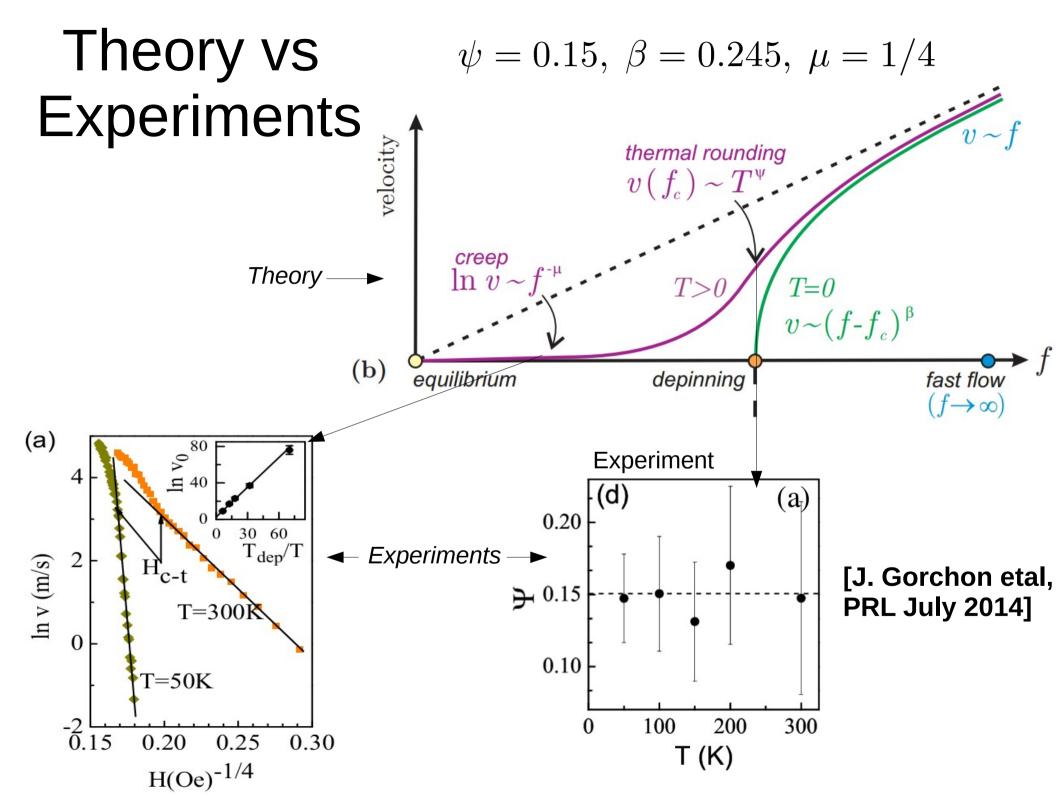


But... do magnetic domain walls in thin films behave as simple elastic strings in disordered media?

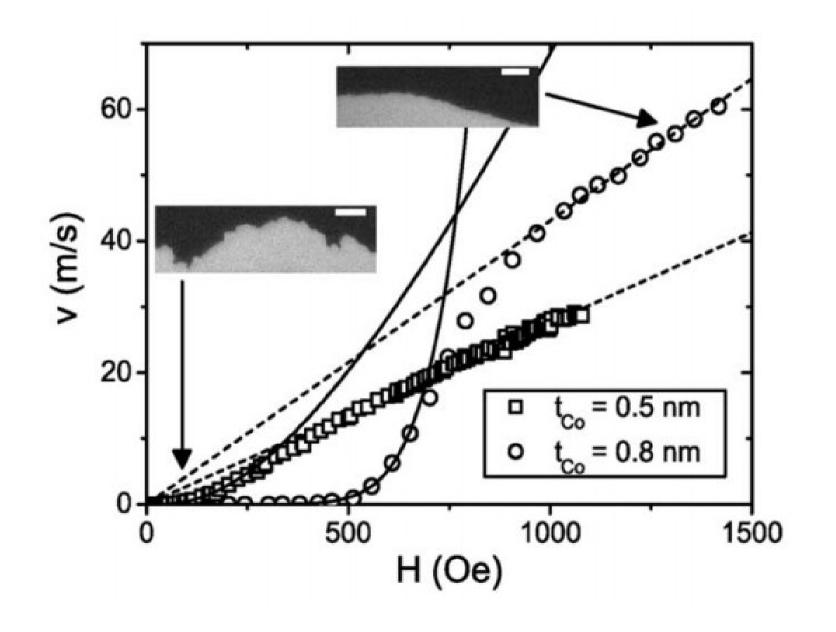
Qualitatively OK!



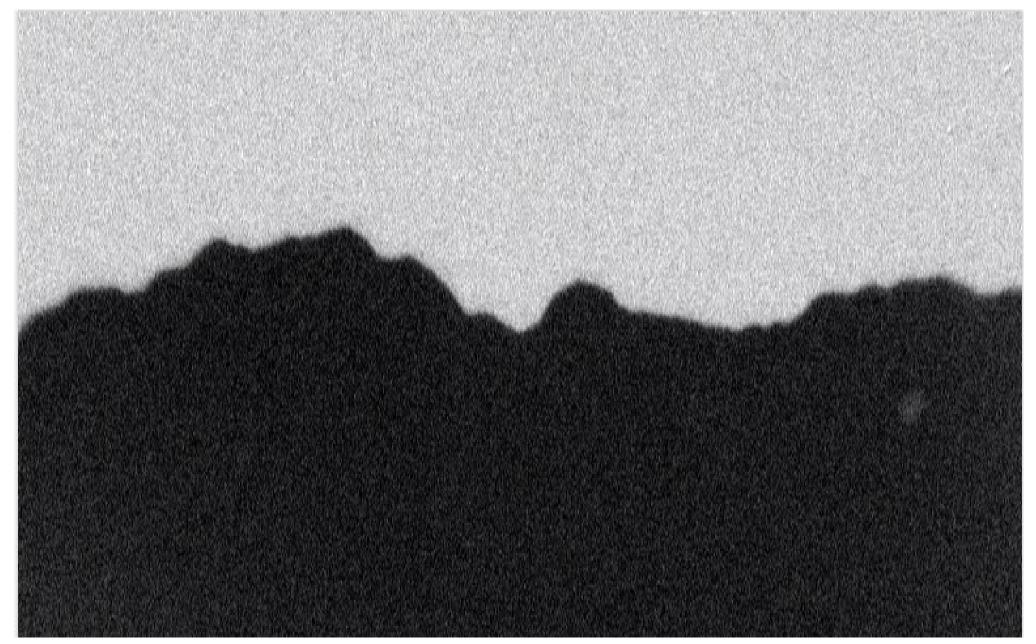
... and quantitatively?



Transport and Rough geometry



Is it moving?

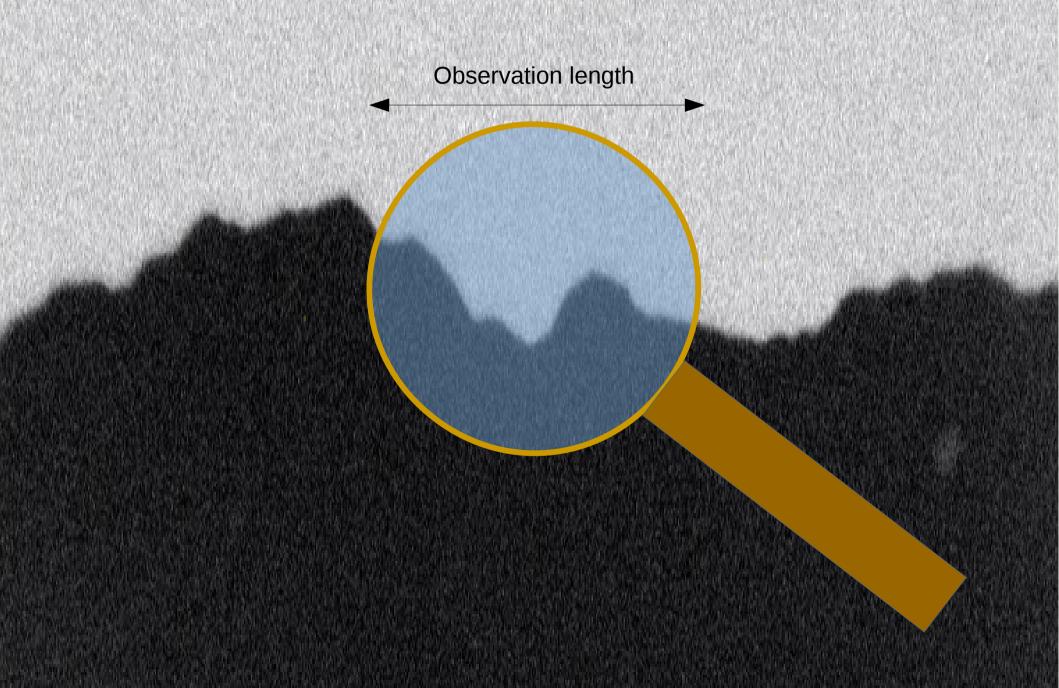


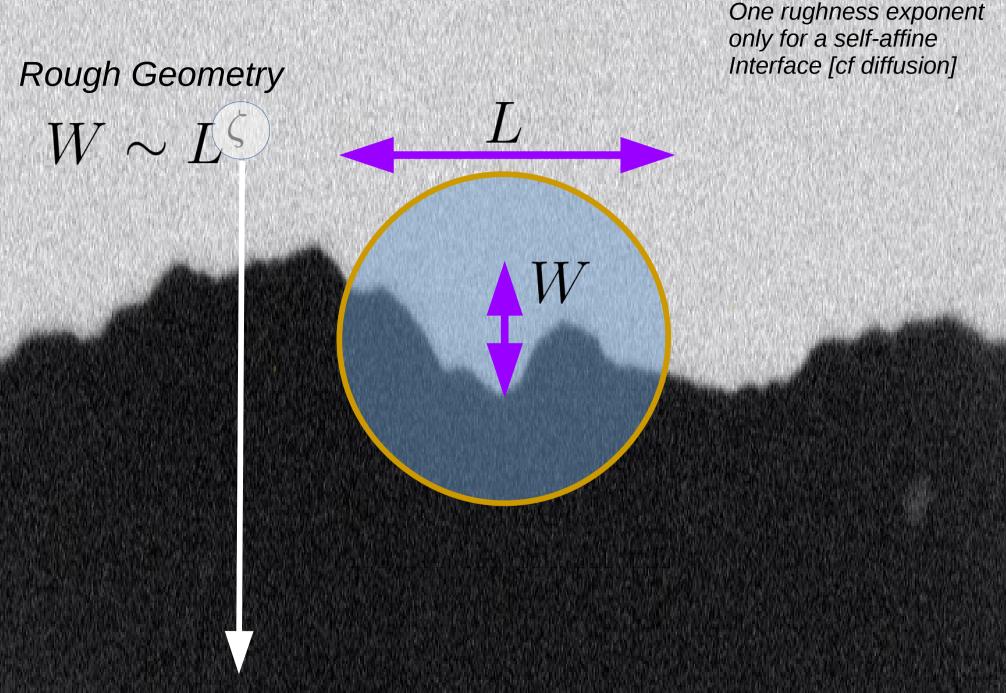
Is it moving?



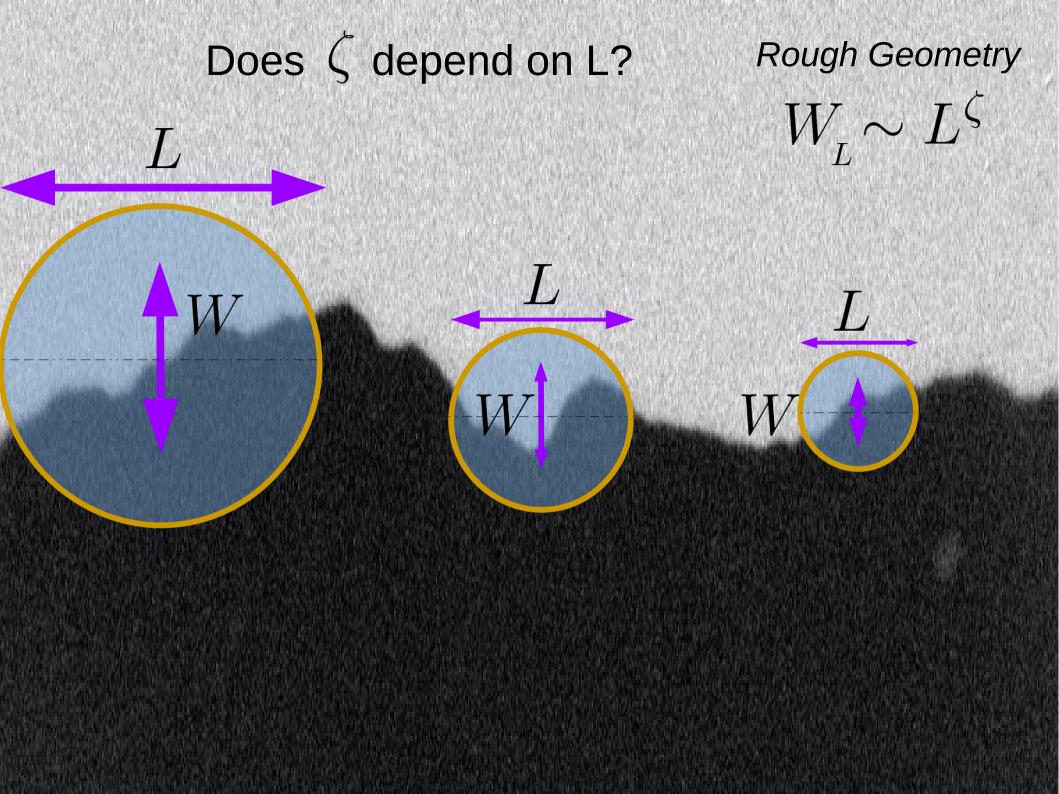
Top Secret! (1984)

Looking for a "hair-like" clue....

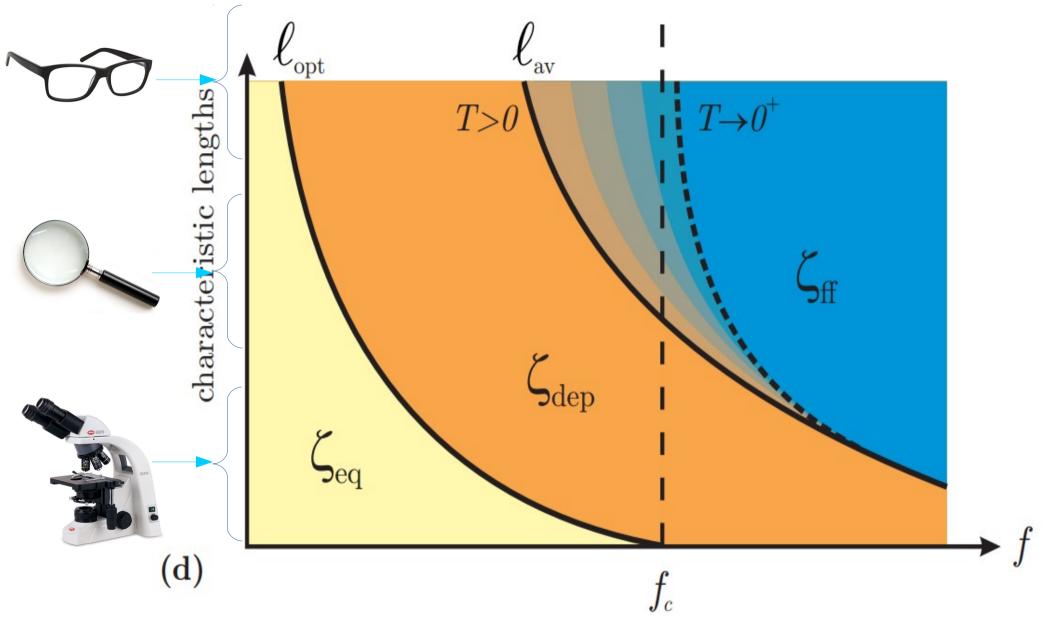




UNIVERSAL ROUGHNESS EXPONENTS



The "hair" of the moving elastic string



ABK, A. Rosso, W. Krauth, T. Giamarchi, PRL (2006), PRB (2009) E. Ferrero, S. Bustingorry, A. B. Kolton, A. Rosso, Comptes Rendus (2013)

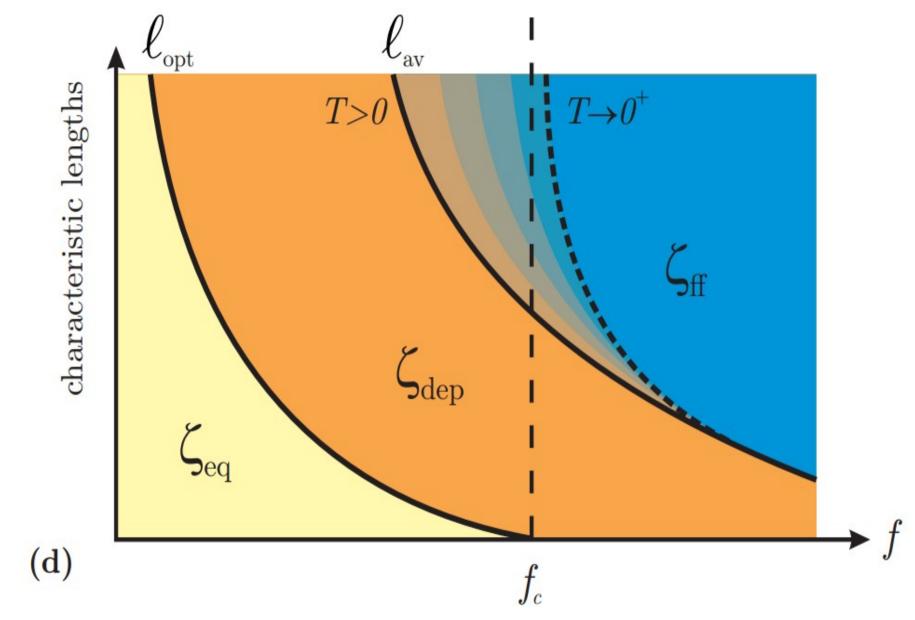
Is it moving?



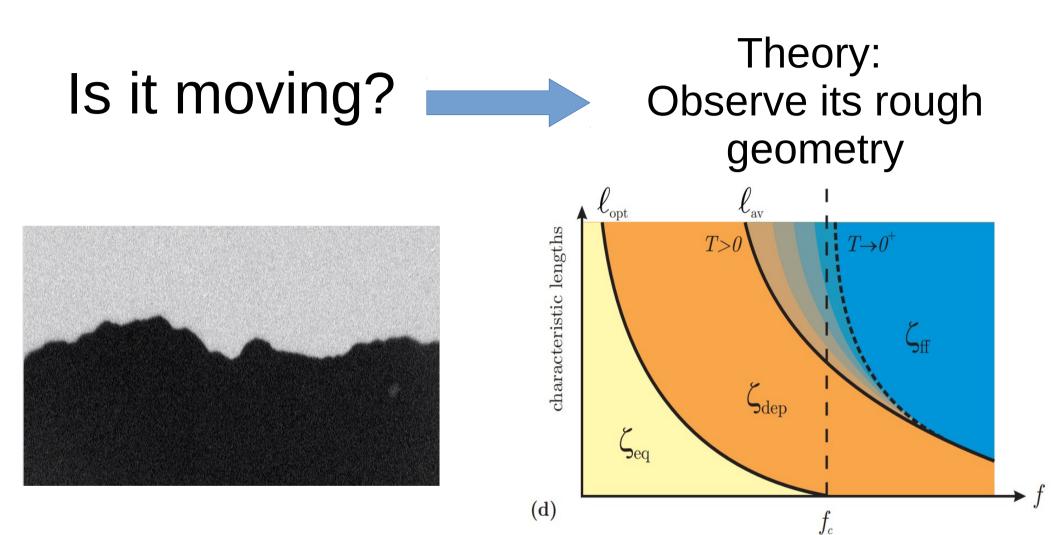
Top Secret! (1984)

We forgot to eliminate memory effects!

Steady-State dynamical diagram



ABK, A. Rosso, W. Krauth, T. Giamarchi, PRL (2006), PRB (2009) E. Ferrero, S. Bustingorry, A. B. Kolton, A. Rosso, Comptes Rendus (2013)





Roughness is the clue for understanding the motion !!

Conclusions

- Motivation:
 - Elasticity vs Disorder: many disordered systems display a similar effective physics \rightarrow minimal models allow us to describe them all.
- Concepts:
 - Universality, collective transport, roughness.
- Playing the game:
 - Parallel finite difference scheme
 - Why in GPU?
 - Quenched disorder tricks and other details.
- Comparison with Experiments:
 - We can interpret measurements...
 - we can make quantitative predictions!

Thank you for the attention!

Questions?

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