

Phase Transitions and Computational Complexity

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NP-complete problems

[Cook, Karp 1970s]

- Literally hundreds of fundamental problems are *NP-hard*:
 - Boolean Satisfiability [model checking, VLSI, ...],
 - Graph Colouring [timetabling/allocation, ...],
 - Travelling Salesman [scheduling problems, ...],
 - ...
- **$P \neq NP$ conjecture**: no efficient algorithm in the *worst case*.
- \rightsquigarrow “Average Case analysis of algorithms”

Phase transitions in discrete structures

- Colouring random graphs or hypergraphs
- Independent sets in random graphs
- Random k -SAT

Phase transitions in discrete structures

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- Conjectured **phase transitions**... *[experiments, physics]*
- ... appear to affect the **combinatorial** and **computational** nature.

- Random k -SAT: “classical” stuff.
- The physics perspective (non-rigorous).
- Physics-inspired rigorous stuff.
- Open problems.

The k -SAT problem

- Fix $k \geq 3$ and let x_1, x_2, \dots, x_n be *Boolean variables*.
- Given an expression of the form

$$\underbrace{(x_1 \vee \bar{x}_{17} \vee \dots \vee x_{29})}_{k\text{-clause}} \wedge \underbrace{(\bar{x}_{11} \vee x_2 \vee \dots \vee \bar{x}_1)}_{k\text{-clause}} \wedge \dots$$

- ... find a *Boolean assignment* that makes the entire formula **true**.

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- ... find a *Boolean assignment* that makes the entire formula **true**.

The naive algorithm

- Check all 2^n *possible assignments*!
- (Prohibitive even for $n = 100$.)
- Yet **no better** worst-case algorithm is known.

Random k -SAT

The random k -SAT model

- Fix $k \geq 3$ and let x_1, \dots, x_n be Boolean variables.
- Generate k -clauses C_1, \dots, C_m *uniformly and independently*.
- Let $\Phi = C_1 \wedge \dots \wedge C_m$.

Random k -SAT

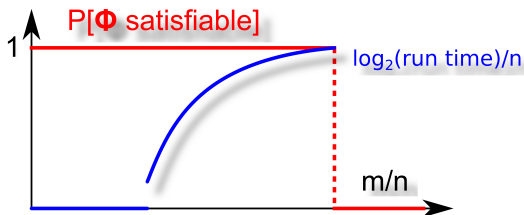
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Kirkpatrick, Selman (experimental)

[Science 1994]

- There occurs a **satisfiability phase transition**.
- *Industrial SAT solvers* require exponential time.



Theorem

[Achlioptas, Peres: JAMS 2004]

The random formula Φ is...

- ... **satisfiable** if $m/n < 2^k \ln 2 - \Theta(k)$.
- ... **unsatisfiable** if $m/n > 2^k \ln 2$.

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Proof

Non-constructive ('2nd moment method').

Hunting the k -SAT threshold

- The **precise** k -SAT threshold is *unknown* for any $k \geq 3$.
- (In fact, it is not known to exist.)

Question

- The *threshold* is $r_k \sim 2^k \ln 2$.
- For what m/n can we *find* satisfying assignments *efficiently*?

Algorithms for random k -SAT

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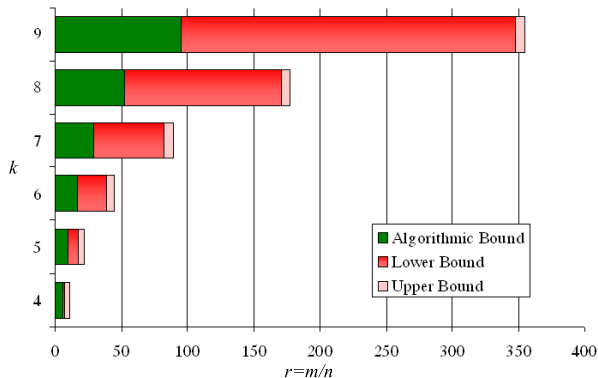
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<i>Algorithm</i>	<i>Density $m/n < \dots$</i>	
Pure Literal	$\sim 2 \ln(k)/k$	BFU 1993
Walksat, proven	$0.04 \cdot 2^k/k$	ACO, Frieze 2012
Walksat, conjectured	$2^k/k$	Monasson, Semerijan 2003
Shortest Clause	$\frac{e^2}{8} \cdot 2^k/k$	Chvatal, Reed 1992
Unit Clause	$\frac{e}{2} \cdot 2^k/k$	Chao, Franco 1990
SC+backtracking	$1.817 \cdot 2^k/k$	Frieze, Suen 1996
Fix	$\sim 2^k \ln(k)/k$	ACO 2009

Algorithms for random k -SAT

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The statistical mechanics perspective

- “Mean-field models of disordered systems” (such as glasses).
- Phase transitions in **glasses** hypothesized by *Kauzmann* (1948).

Mezard, Parisi, Zecchina

[Science 2002]

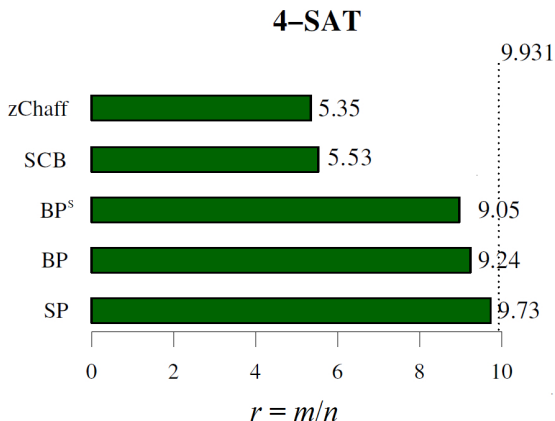
- The (non-rigorous) **cavity method**.
- “Replica symmetry breaking”.
- New algorithms: *Belief/Survey Propagation* guided decimation.

Enter the physicists

Experiments

[Kroc, Sabharwal, Selman 2009]

For **small** k , BP/SP guided decimation fare *extremely* well.



The cavity method

A thought experiment

- Generate a random formula Φ . Fix it.

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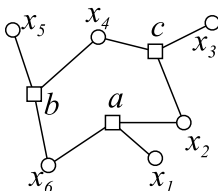
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The factor graph $G(\Phi)$

- *Vertices*: clauses **and** variables.
- *Edges* between clauses and the variables they contain \rightsquigarrow **metric**.



Replica symmetry/non-reconstruction

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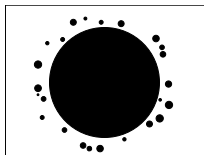
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- Geometrically, $\mathcal{S}(\Phi) = \{\text{all satisfying assignments}\}$ is a “giant ball”.



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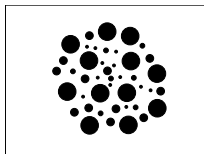
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- Geometrically, $\mathcal{S}(\Phi) = \{\text{all satisfying assignments}\}$ “shatters”.



The cavity method

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Still, so long as

$$m/n < r_{\text{cond}} = 2^k \ln 2 - \frac{3}{2} \ln 2 + o_k(1),$$

for y_1, \dots, y_l **mutually far apart** we have

$$\mathbb{E} \|\sigma(y_1, \dots, y_l) - \sigma(y_1) \otimes \dots \otimes \sigma(y_l)\|_{\text{tv}} \rightarrow 0.$$

The cavity method

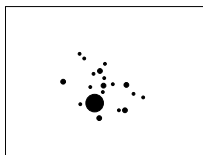
Condensation

$$[r_{\text{cond}} < m/n < r_{\text{sat}}]$$

- A few **large clusters** dominate.
- In particular,

$$\mathbb{E} \|\sigma(y_1, \dots, y_l) - \sigma(y_1) \otimes \dots \otimes \sigma(y_l)\|_{\text{TV}} \not\rightarrow 0.$$

- Still, there is an abundance of **small clusters**.



- Suppose that

$$\mathbb{E} \|\sigma(y_1, \dots, y_l) - \sigma(y_1) \otimes \dots \otimes \sigma(y_l)\|_{tv} \rightarrow 0.$$

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- Suppose that

$$\mathbb{E} \|\sigma(y_1, \dots, y_l) - \sigma(y_1) \otimes \dots \otimes \sigma(y_l)\|_{tv} \rightarrow 0.$$

- Locally, the factor graph is a *tree*.
- Therefore, for a *fixed* Φ , there is a *recurrence* for

$$\mu(y) = \mathbb{E}_{\sigma \in \mathcal{S}(\Phi)} [\sigma(y)].$$

- \rightsquigarrow **Belief Propagation equations.**

Survey propagation

- The *Survey propagation distribution* on $\mathcal{S}(\Phi)$:
 - Pick a **cluster** \mathcal{C} uniformly.
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- \rightsquigarrow **Survey Propagation equations.**

- Yields a prediction on the *k-SAT threshold*...
- ...in the form of an infinite-dimensional variational problem.

Belief Propagation guided decimation

- For $i = 1, \dots, n$ do
 - use **BP** to approximate $\mu(x_i)$.
 - set x_i to *true* with the resulting probability.
 - simplify the formula accordingly.

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A similar scheme yields *Survey Propagation guided decimation*.

Proving the physics hypotheses

- Hunting the k -SAT threshold
- Shattering/Condensation
- Message passing algorithms

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- **Best current bounds:** 2nd moment method [Achlioptas, Peres 2004]
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 - Generalises to other problems [Panchenko, Talagrand 2004]
 - *Sherrington-Kirkpatrick* model [Talagrand 2006]

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Theorem

[ACO, Panagiotou 2012]

The threshold for 2-coloring k -uniform hypergraphs is

$$r_{2-col} = 2^{k-1} \ln 2 - \frac{\ln 2}{2} - \frac{1}{4} + \tilde{O}(2^{-k}).$$

Loose/frozen variables

Let Φ be a k -CNF, σ a *satisfying assignment*, and x a variable.

- x is **loose** if there is a satisfying assignment τ such that

$$\sigma(x) \neq \tau(x) \text{ and } \text{dist}(\sigma, \tau) \leq \ln(n).$$

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$$\sigma(x) \neq \tau(x) \text{ and } \text{dist}(\sigma, \tau) \leq \ln(n).$$

- x is **frozen** if for any satisfying assignment τ

$$\sigma(x) \neq \tau(x) \Rightarrow \text{dist}(\sigma, \tau) = \Omega(n).$$

The solution space geometry

Theorem

[Achlioptas, ACO 2008]

Choose $\sigma \in \mathcal{S}(\Phi)$ uniformly at random.

- 1 $r < (1 - o_k(1))2^k \ln k/k \Rightarrow$ all but $o_k(1)n$ vars are *loose* w.h.p.
- 2 $r > (1 + o_k(1))2^k \ln k/k \Rightarrow$ all but $o_k(1)n$ vars are *frozen* w.h.p.

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A geometric perspective

For $r > (1 + o_k(1))2^k \ln k/k \dots$

- \dots there is *shattering*.
- $\dots \mathcal{S}(\Phi)$ resembles an *error correcting code*.

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Molloy 2012: *exact* threshold for freezing in random graph colouring.

Theorem

[ACO, Zdeborova 2012]

There is a *condensation transition* in random hypergraph 2-coloring at

$$r_{cond} = 2^{k-1} \ln 2 - \ln 2 + o_k(1).$$

- Large clusters dominate.
- Belief Propagation breaks.

Message passing algorithms

- The best “combinatorial” algorithm succeeds for $m/n < 2^k \ln(k)/k$.
- Do BP/SP perform better?

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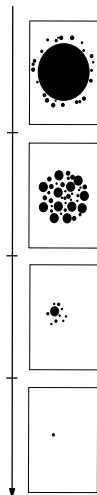
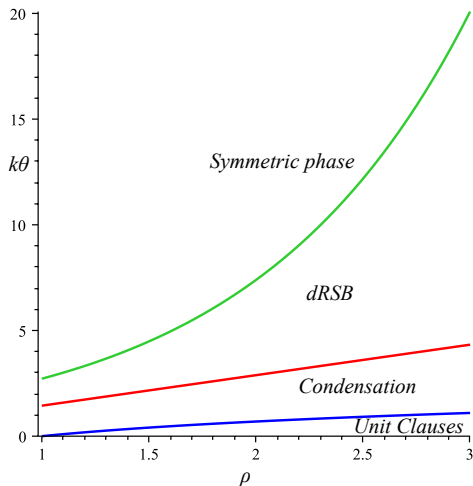
[ACO 2011]

BP guided decimation **fails** for

$$m/n > c \cdot 2^k / k$$

for some constant $c > 0$ w.h.p.

Let $m/n = \rho \cdot 2^k/k$, $\theta = 1 - t/n$ = fraction of *unassigned* vars.



Open problems

- The BP/SP equations.
- Simulated annealing and sampling.
- Local computations.
- Hiding solutions.

The BP/SP equations

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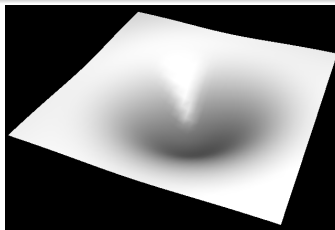
Montanari, Shah 2007: BP works for $m/n < \ln(k)/k$.

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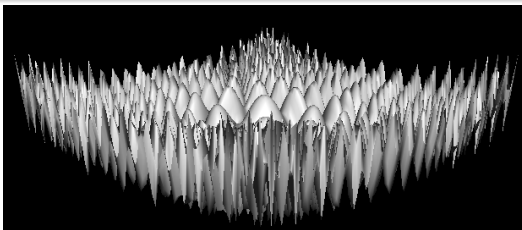
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- *Symmetric regime*: $\mathcal{S}(\Phi)$ is the bottom of a smooth “valley”.
- **Conjecture**: simulated annealing succeeds in *polynomial time*.
- In fact, simulated annealing should **sample**.

Question

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- *Shattered regime*: $\mathcal{S}(\Phi)$ is the bottom of a ragged landscape.
- **Conjecture**: simulated annealing requires *exponential time*.
- (Very) partial results: Jerrum 1992; ACO, Efthymiou 2011.

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Can such *local algorithms* succeed beyond $c \cdot 2^k/k$?

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- \rightsquigarrow identification scheme based on SAT.

- Physics hypotheses \rightsquigarrow *blueprint* for a very nice theory.
- Precise thresholds, message passing algorithms, ...
- **Open problems:**
 - a rigorous *cavity method*,
 - understanding *message passing algorithms*,
 - analysing “recurrent” algorithms.
- Are random problems *easy* or *hard*?