



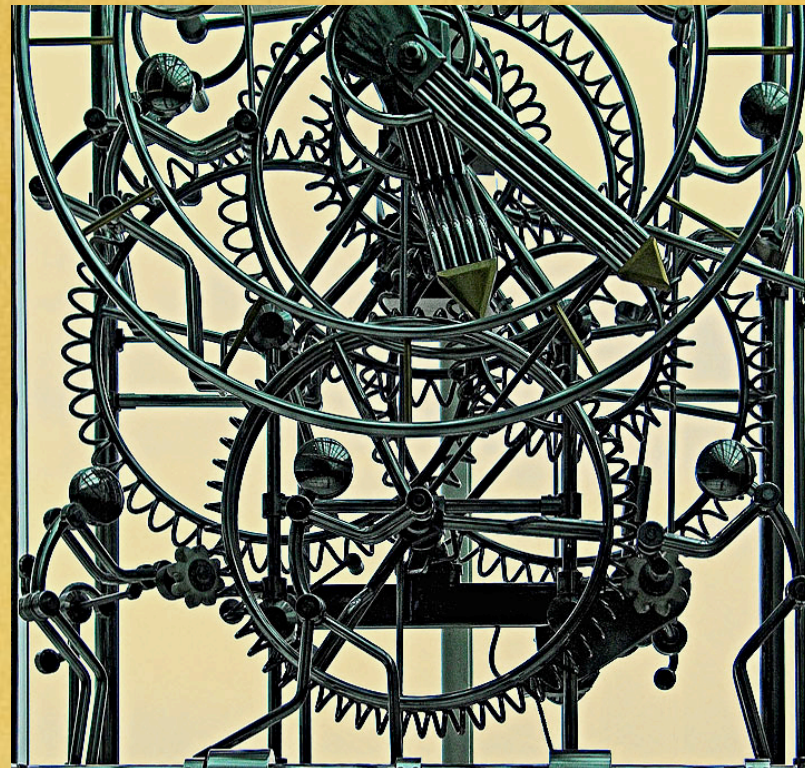
Empirical Methods in Algorithmic Decision Theory

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ADT, Venice, Oct 2009

Why? An example

- Perhaps **complexity** can be an escape from Gibbard-Satterthwaite's theorem?
 - Some voting rules (2nd order Copeland, STV, ..) are NP-hard to manipulate
[Bartholdi, Tovey & Trick 89, Bartholdi & Orlin 91]



Why? An example



- ❧ Is **complexity** an escape from Gibbard-Sattertwhaite's theorem?
- ❧ But recent results raise doubts
- ❧ NP-hardness is only worst case
- ❧ Manipulation might be easy in practice

Why? An example

- ⌘ Is **complexity** an escape from Gibbard-Satterthwaite's theorem?
 - ⌘ For instance, with STV, either a coalition is too small $O(\sqrt{n})$ to change result or so large $\Omega(\sqrt{n})$ they easily can [Xia & Conitzer 08]
 - ⌘ Only question is when coalition is $\Theta(\sqrt{n})$?
 - ⌘ We can run some experiments!



Outline

- ❧ Where are the *hard* problems?
 - ❧ Phase transition behaviour
 - ❧ Sharp/smooth transitions
- ❧ A *close* look at the data
 - ❧ Early mistakes
 - ❧ Discrepancy search & restarts
- ❧ What makes problems *hard*?
 - ❧ Backbones, backdoors, ...
 - ❧ Structure

Where are the really *hard* problems?

☞ Influential IJCAI-91 paper by
Cheeseman, Kanefsky & Taylor

☞ 857 citations on Google Scholar

“... for many NP problems one or more "order parameters" can be defined, and hard instances occur around particular critical values of these order parameters ... the critical value separates overconstrained from underconstrained ...”

Where are the really *hard* problems?

- ∞ Influential IJCAI-91 paper by Cheeseman, Kanefsky & Taylor
 - ∞ 857 citations on Google Scholar

“We expect that in future computer scientists will produce “phase diagrams” for particular problem domains to aid in hard problem identification”

Where are the really *hard* problems?

⌘ 1980s

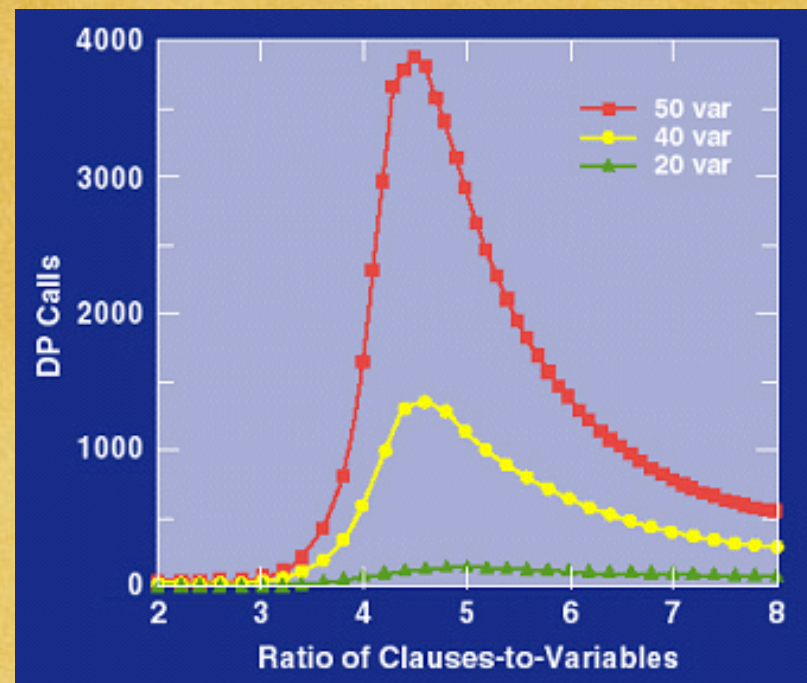
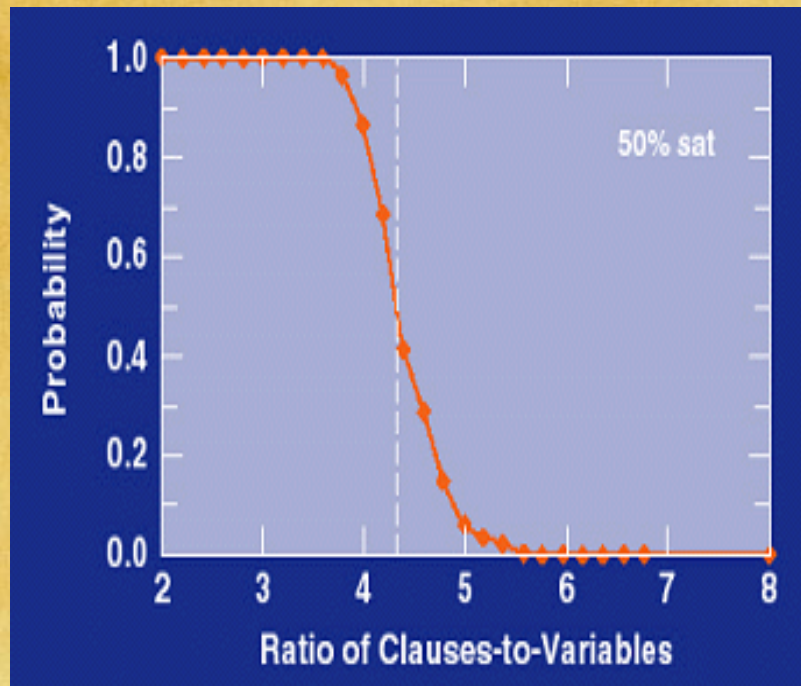
⌘ Difficulty finding hard satisfiability problems using constant probability model

⌘ 1992 (one year after Cheeseman et al)

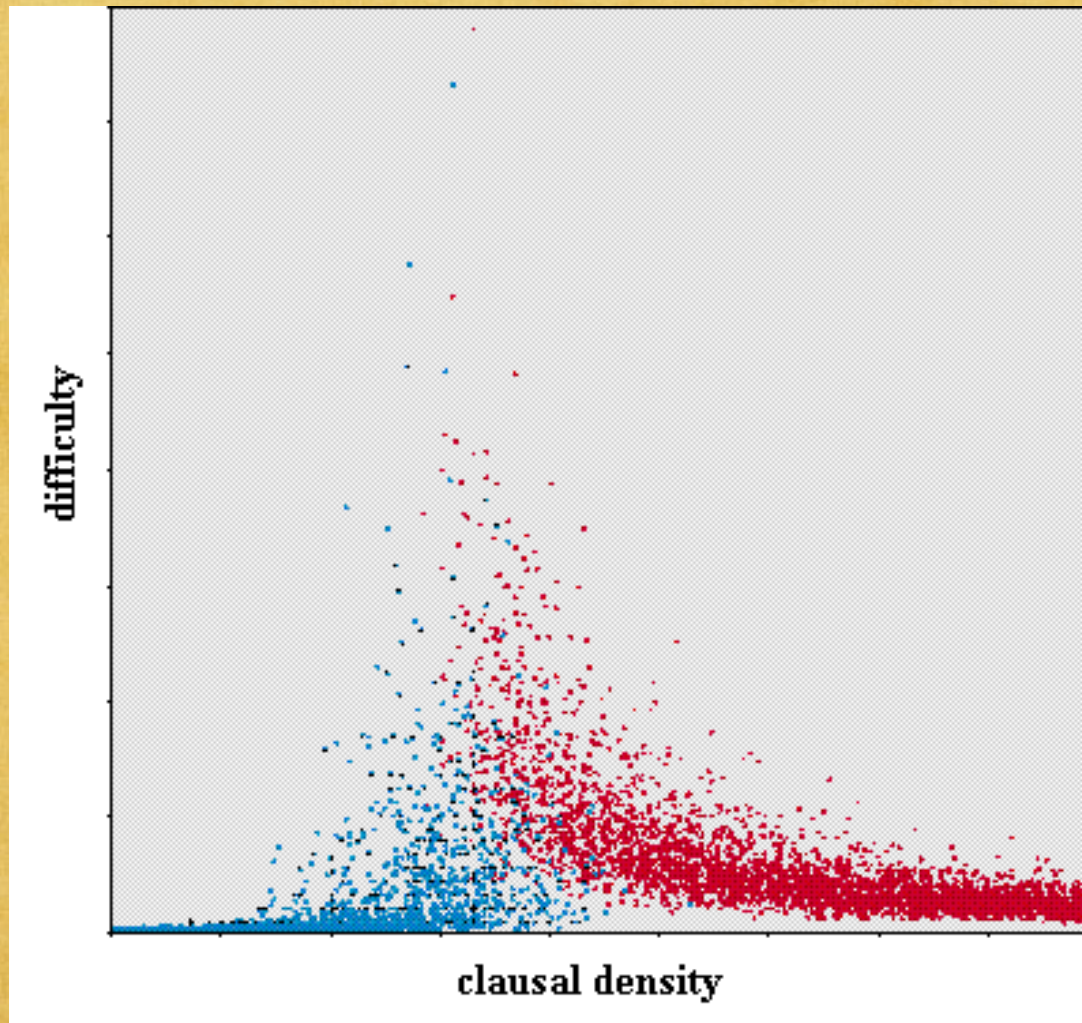
⌘ *Hard & Easy Distributions of SAT Problems*, Mitchell, Selman & Levesque

⌘ 804 citations on Google Scholar

3-SAT phase transition



3-SAT phase transition





Phase transitions

⌘ Polynomial problems

⌘ 2-SAT, arc consistency, ...

⌘ NP-complete problems

⌘ SAT, COL, k-Clique, HC, TSP, number partitioning, ..

⌘ Higher complexity classes

⌘ QBF, planning, ...



**So where are the hard
manipulation problems?**

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[Walsh, IJCAI-09]

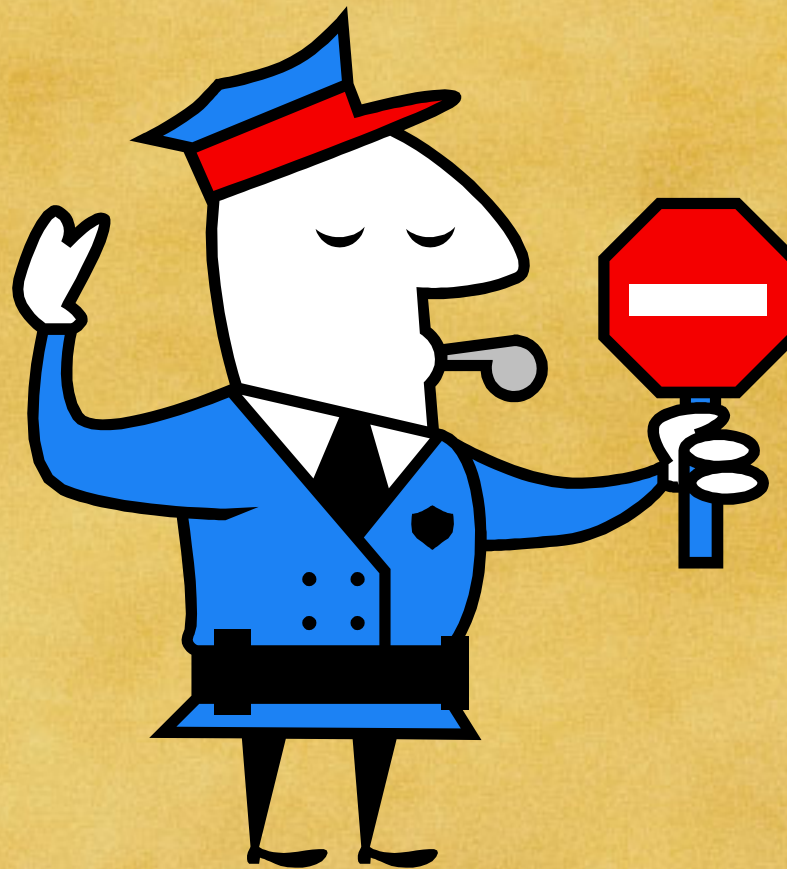
Manipulating the veto rule

- Simple rule to analyse
 - Each voter gets one veto
 - Candidate with least vetoes wins
- But on border of complexity
 - NP-hard to manipulate constructively with 3 or more candidates
 - Polynomial to manipulate destructively



Manipulating veto rule

- ❧ Manipulation not possible with 2 candidates
- ❧ If the coalition want A to win then veto B



Manipulating veto rule

- ❧ Manipulation possible with 3 candidates
- ❧ Voting strategically can improve the result



Manipulating veto rule

- Suppose
 - A has 4 vetoes
 - B has 2 vetoes
 - C has 3 vetoes
- Coalition of 5 voters
 - Prefer A to B to C



Manipulating veto rule

- ❧ Suppose
 - ❧ A has 4 vetoes
 - ❧ B has 2 vetoes
 - ❧ C has 3 vetoes
- ❧ Coalition of 5 voters
 - ❧ Prefer A to B to C
 - ❧ If they all veto C, then B wins



Manipulating veto rule

- Suppose
 - A has 4 vetoes
 - B has 2 vetoes
 - C has 3 vetoes
- Coalition of 5 voters
 - Prefer A to B to C
 - Strategic vote is for 3 to veto B and 2 to veto C



Manipulating veto rule

- With 3 or more candidates
- Unweighted votes
 - Manipulation is polynomial to compute
- Weighted votes
 - Destructive manipulation is polynomial
 - Constructive manipulation is NP-hard (=number partitioning)



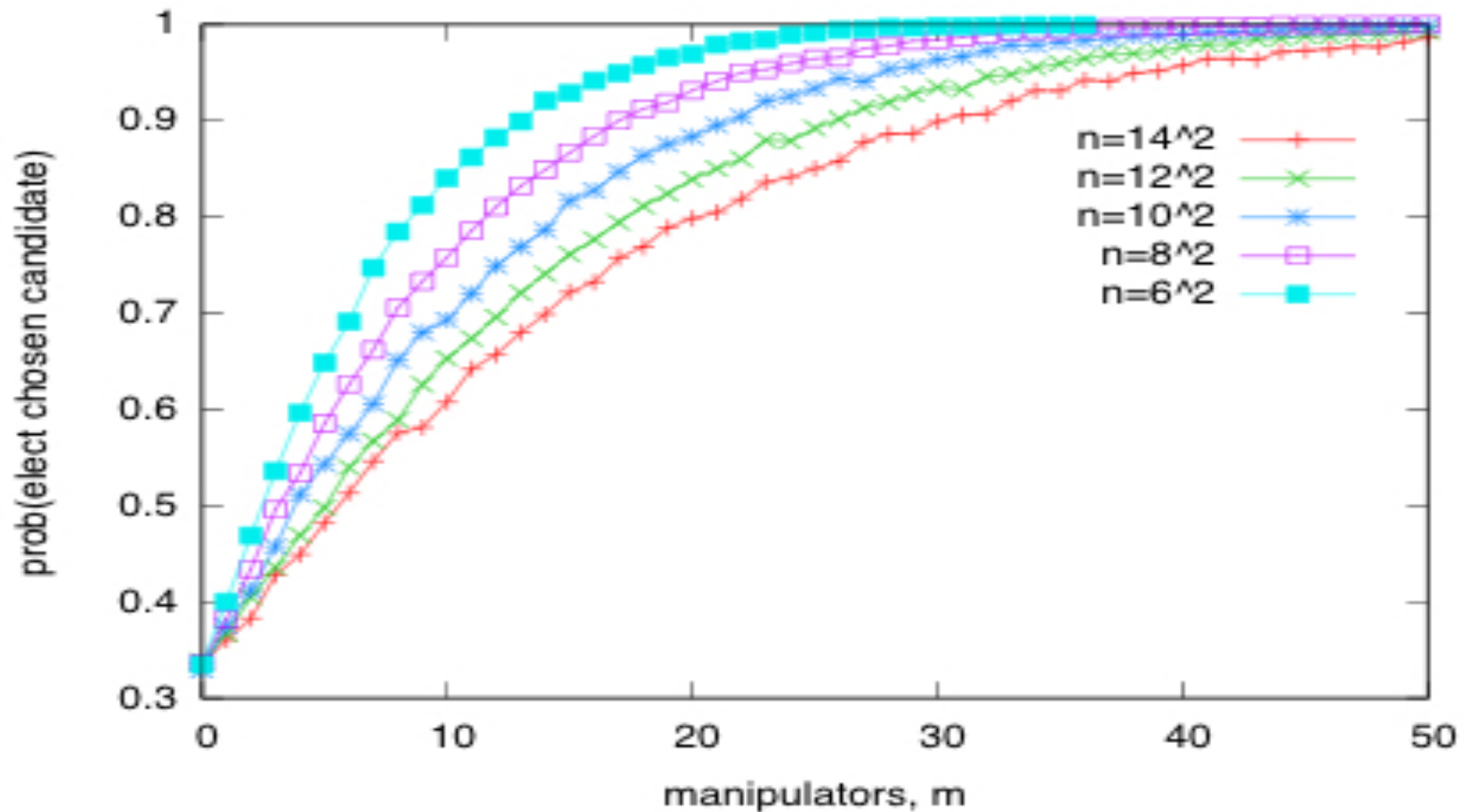
Uniform votes

- ⌘ n agents
- ⌘ 3 candidates
- ⌘ coalition of size m
- ⌘ weights from $[0, k]$

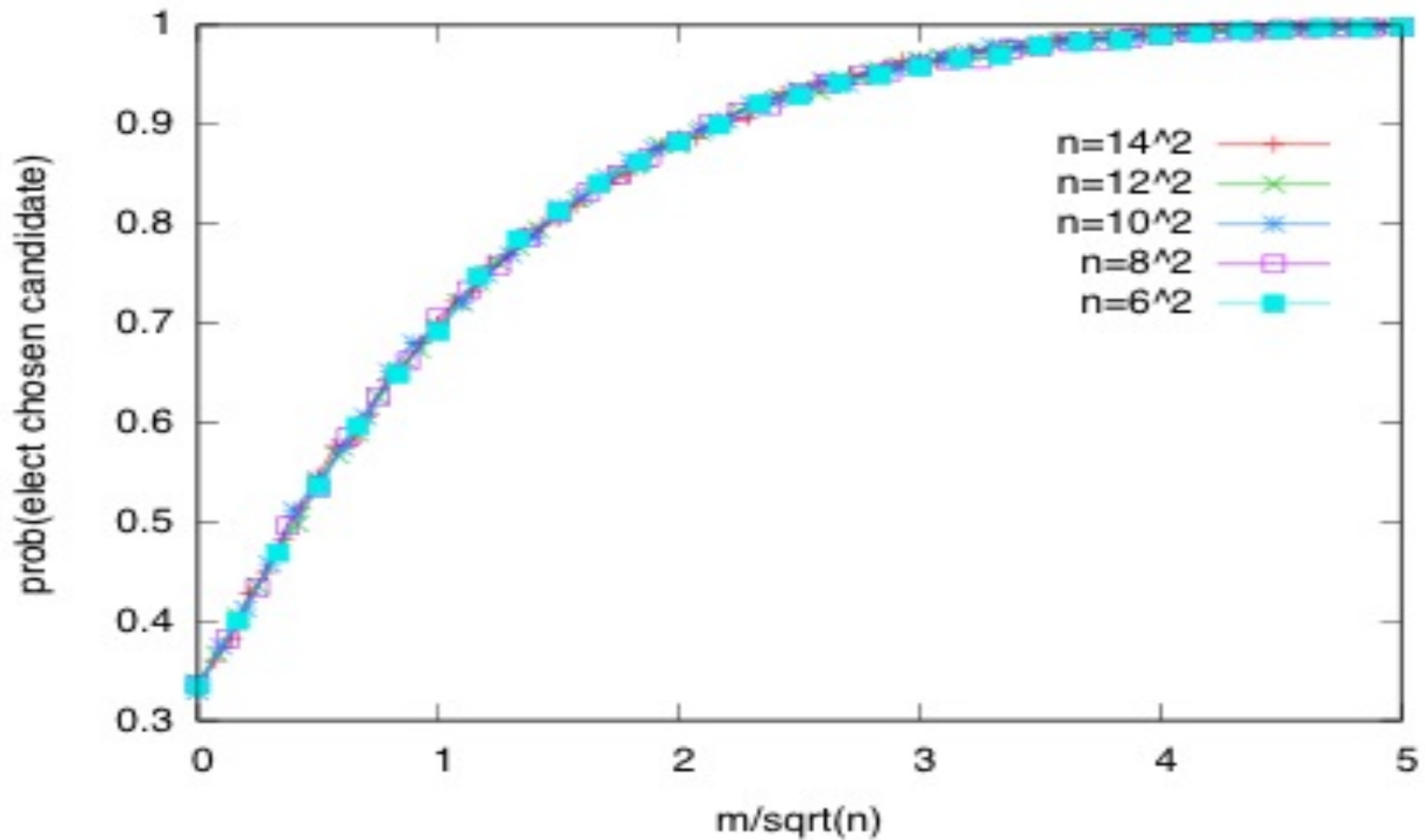
aka “Impartial Culture”



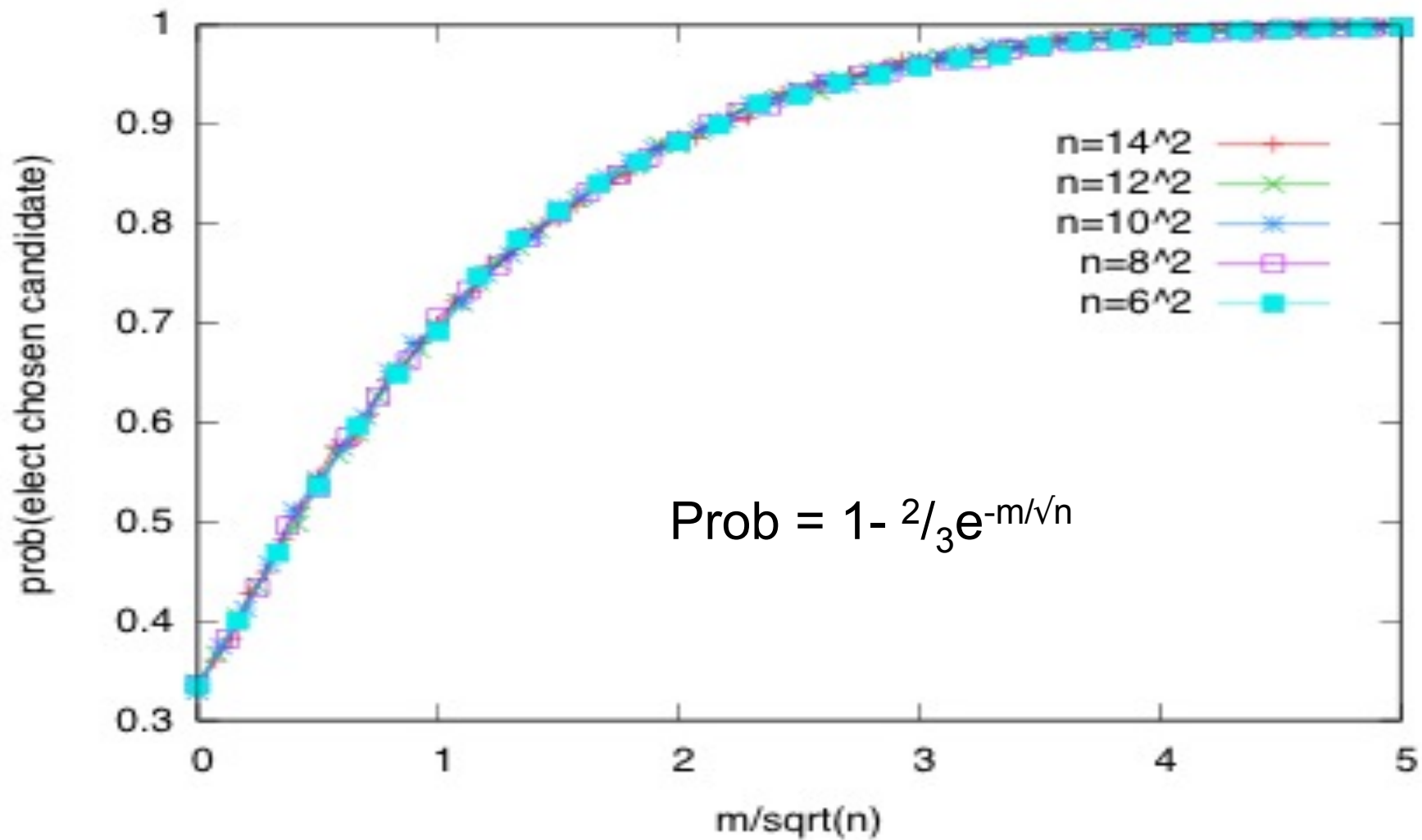
Phase transition



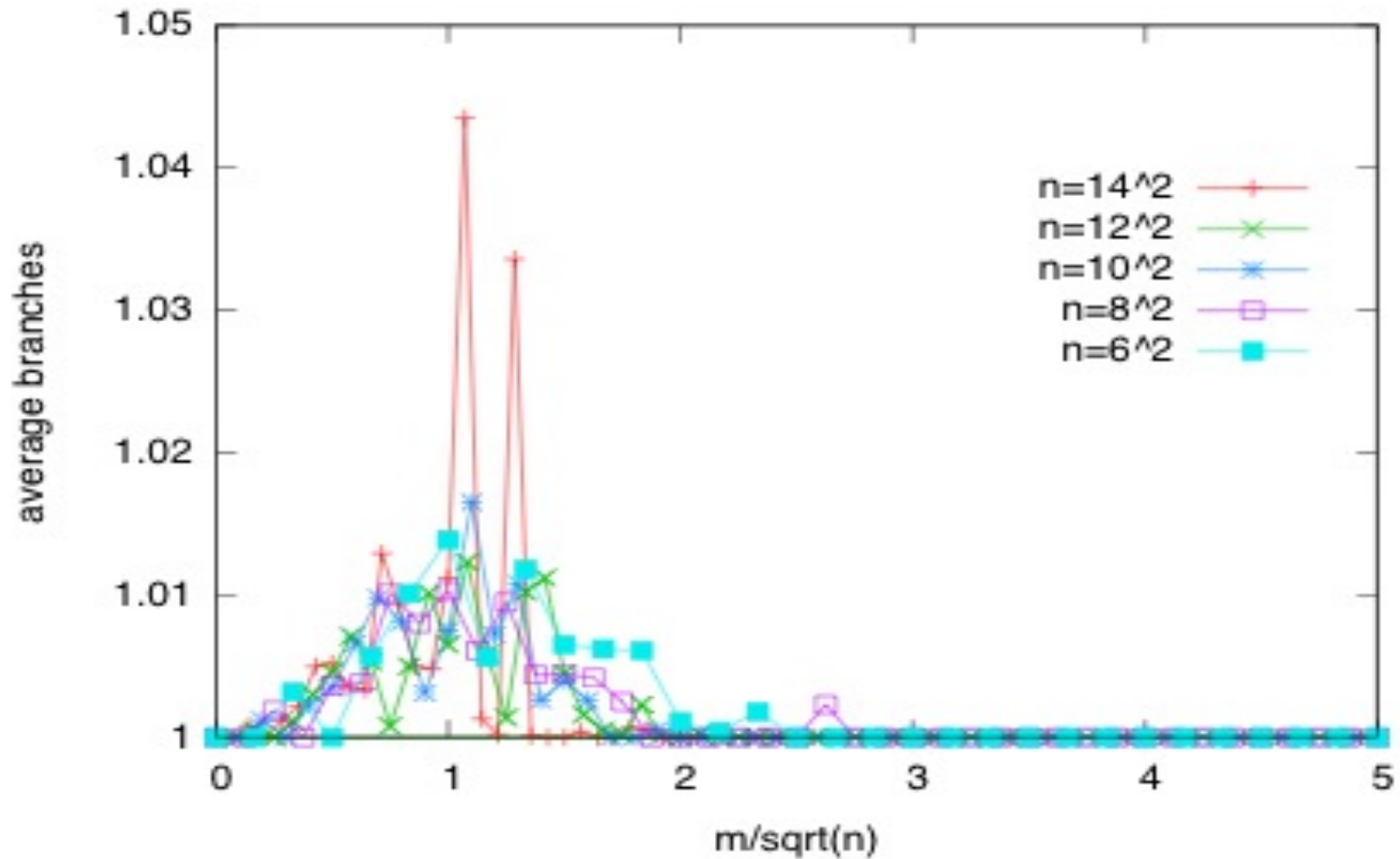
Phase transition



Phase transition



Phase transition



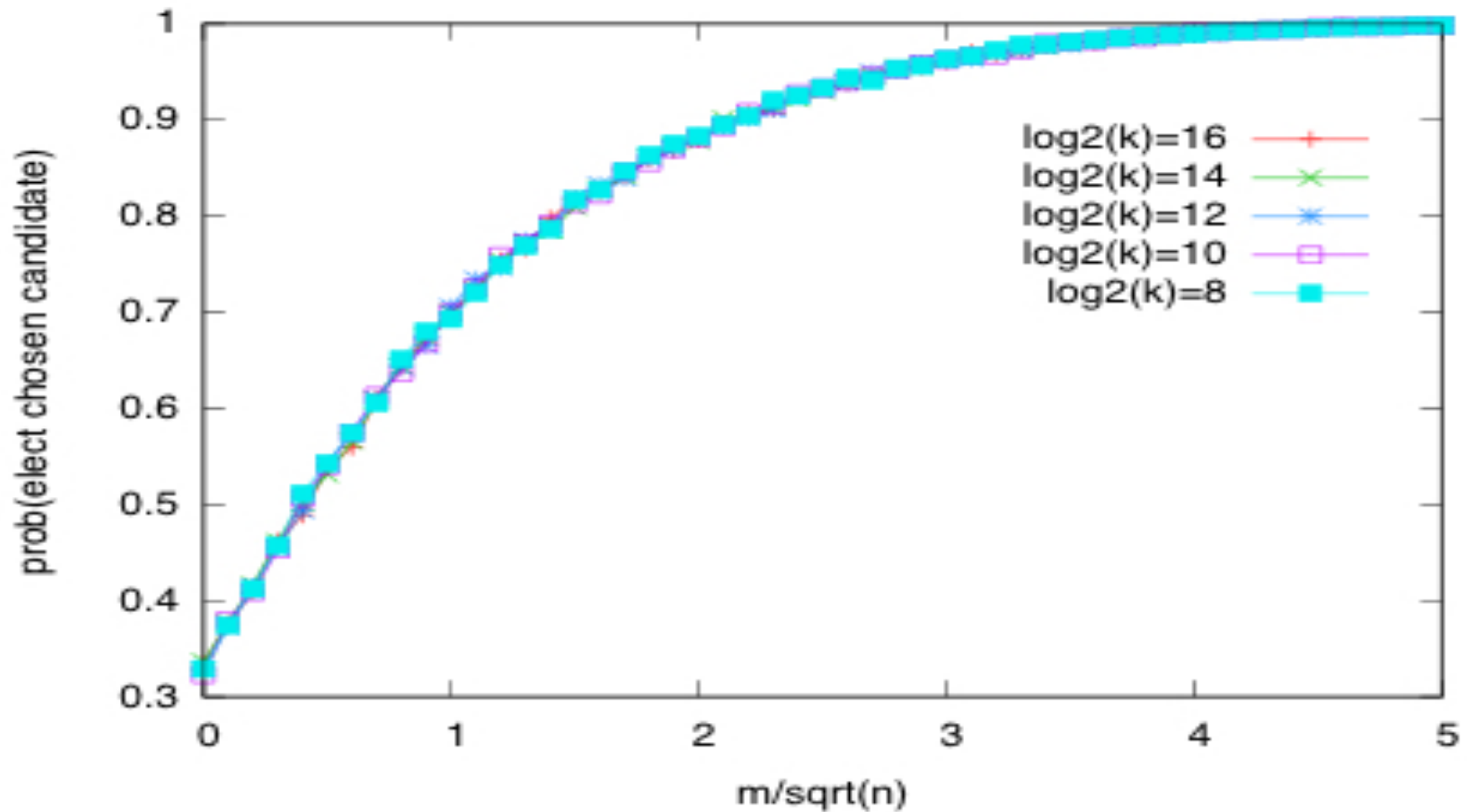


Phase transition



- ☞ Same result with other distributions of votes
 - ☞ Different size weights
 - ☞ Normally distributed weights
 - ☞ ..

Phase transition



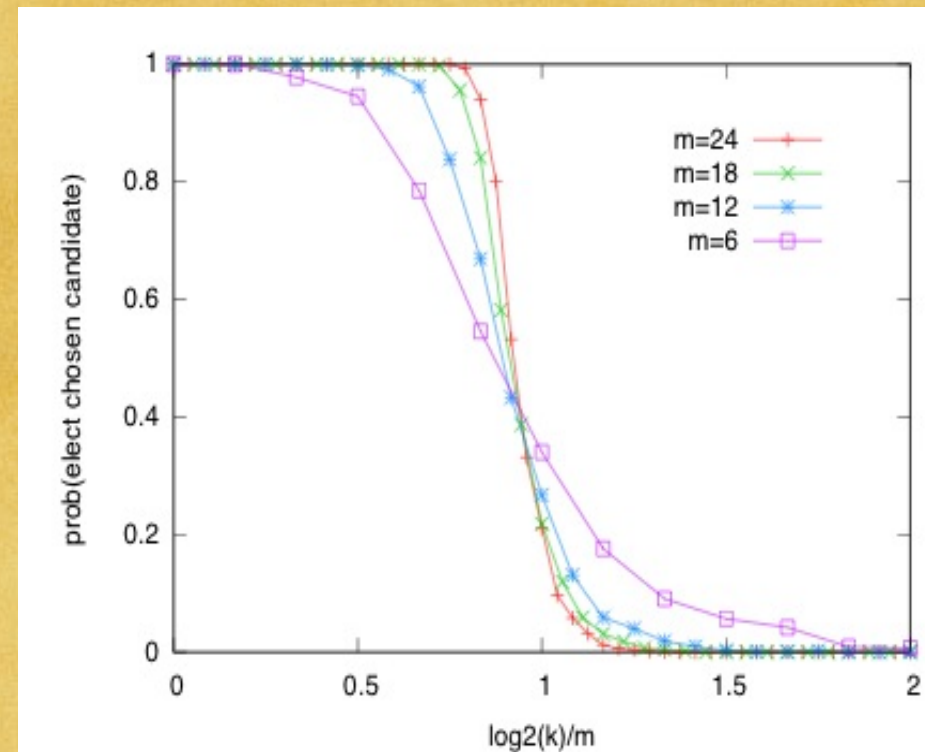
Why is manipulation easy?

- ☞ Coalition needs to be large enough to be able to change result
 - ☞ Coalition size $m = O(\sqrt{n})$
- ☞ But if the coalition is large
 - ☞ Variance in number of vetoes is large, $O(m)$
- ☞ Easy to find a partition of votes or to prove none exists
 - ☞ Greedy heuristic solves problem
 - ☞ Or simple bound proves it is impossible

So where are the hard manipulation problems?

⌘ Hung election

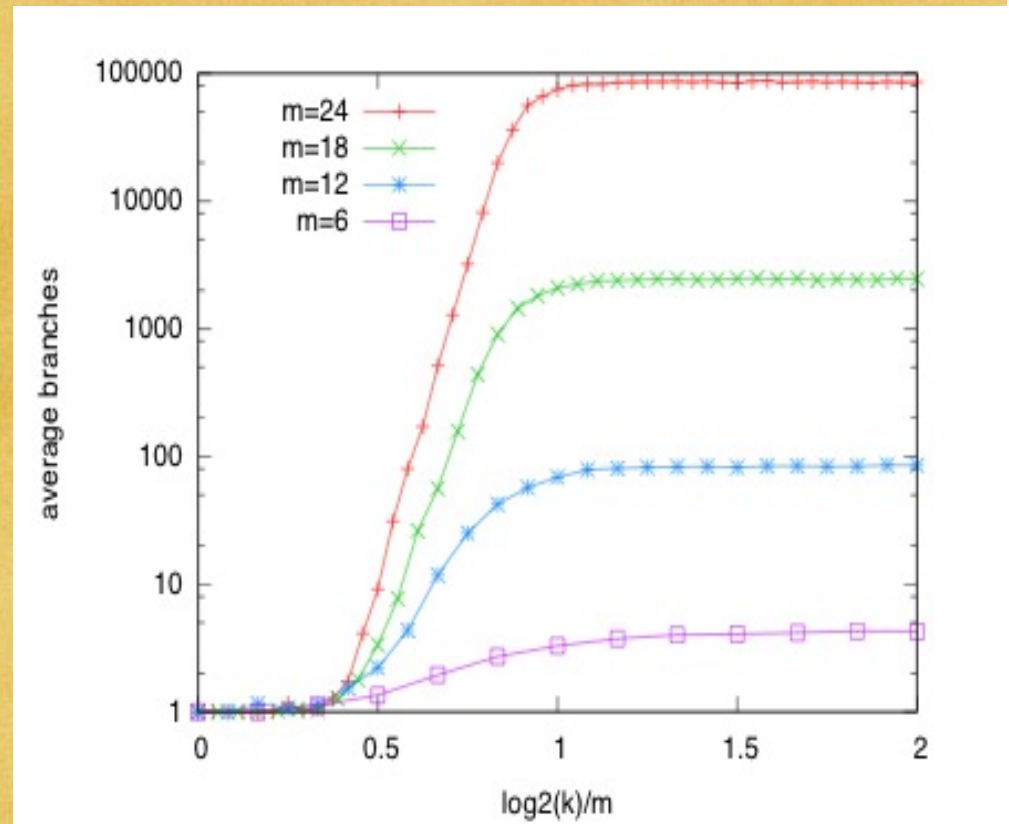
- ⌘ n voters have vetoed one candidate
- ⌘ coalition of size m has twice weight of these n voters



So where are the hard manipulation problems?

⌘ Hung election

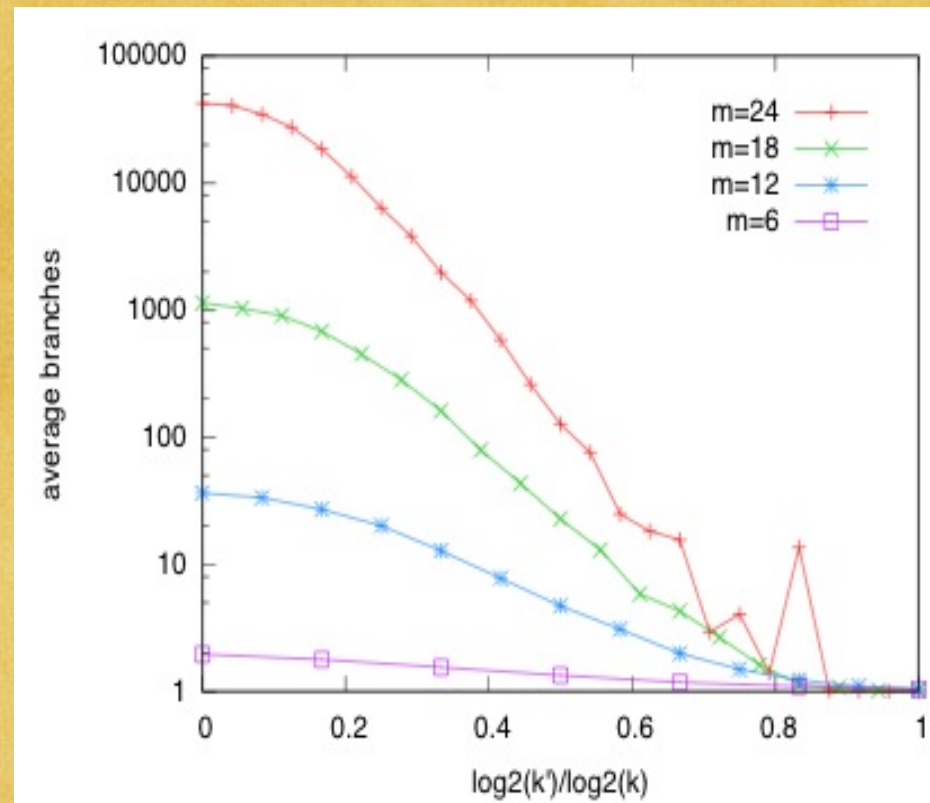
- ⌘ n voters have vetoed one candidate
- ⌘ coalition of size m has twice weight of these n voters



So where are the hard manipulation problems?

☞ Hung election

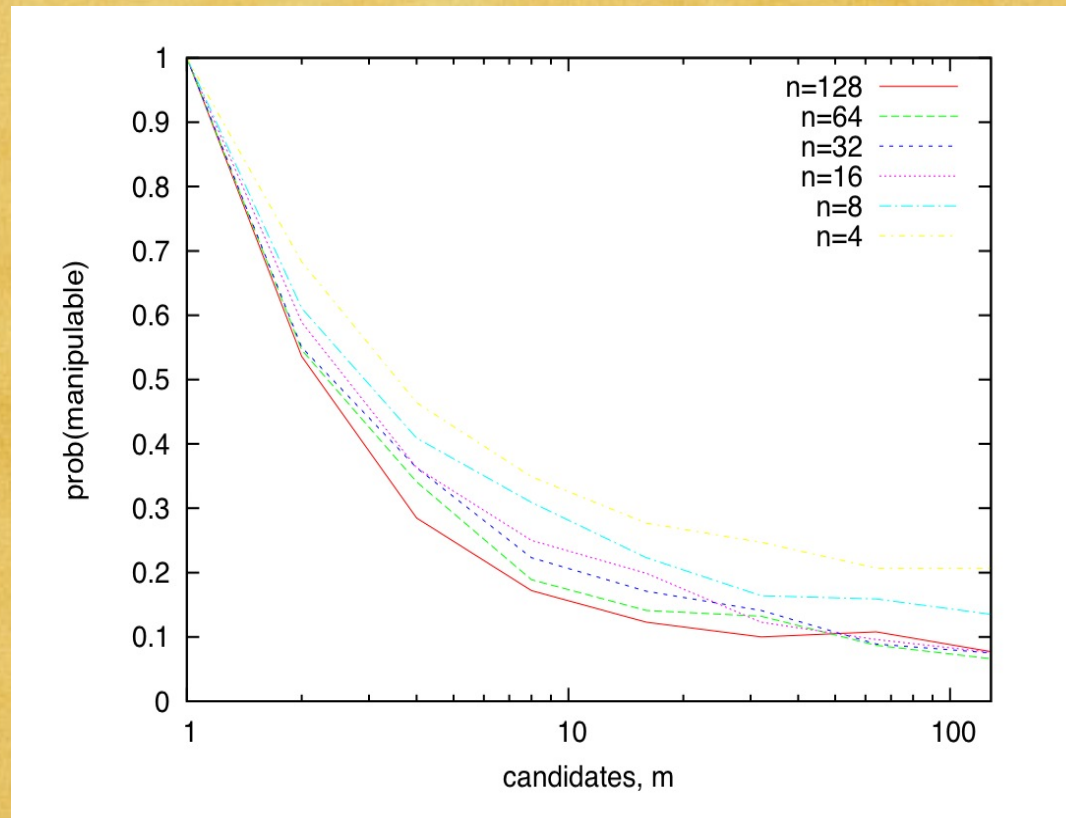
- ☞ n voters have vetoed one candidate
- ☞ coalition of size m has twice weight of these n voters
- ☞ But one random voter with enough weight makes it **easy**



What about unweighted votes?

- STV is one of the few voting rules where manipulation by a single agent is NP-hard **without** weights
 - Unbounded number of candidates
- Conitzer gives an $O(n1.62^m)$ procedure to compute this
 - How does this perform in practice?*

STV phase transition

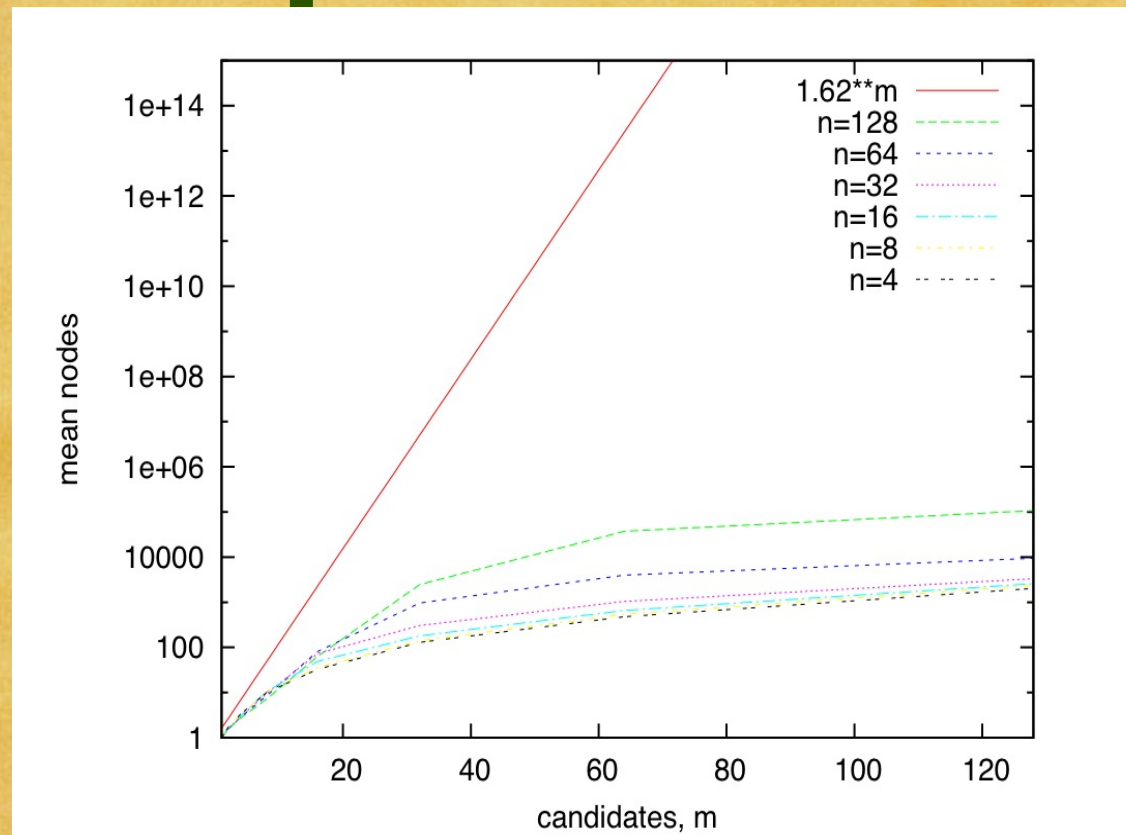


Smooth not sharp?

Other smooth transitions: 2-COL, 1in2-SAT, ...

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STV phase transition



☞ Fits 1.008^m with coefficient of determination $R^2=0.95$



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A close look at the data ...

☞ Often worth having a look at the data

(1054 SAT 10 0.01)

(1055 SAT 11 0.01)

☞ Sometimes interesting items

(1056 SAT 10 0.01)

(1067 SAT 21 0.01)

jump out

(1067 SAT 17059238 52653)

☞ Outliers are always interesting!

(1068 SAT 10 0.01)

(1069 SAT 10 0.01)

☞ Where complexity can hide

...

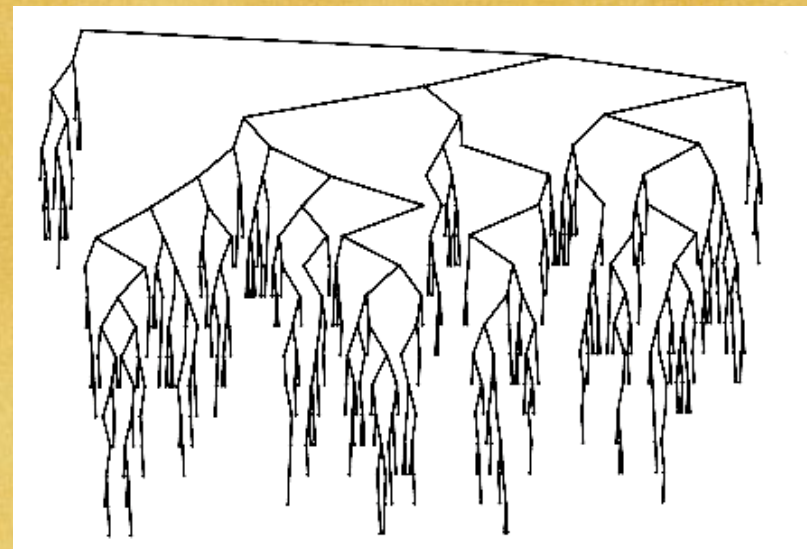
Easy problems are sometimes hard

❧ Under-constrained

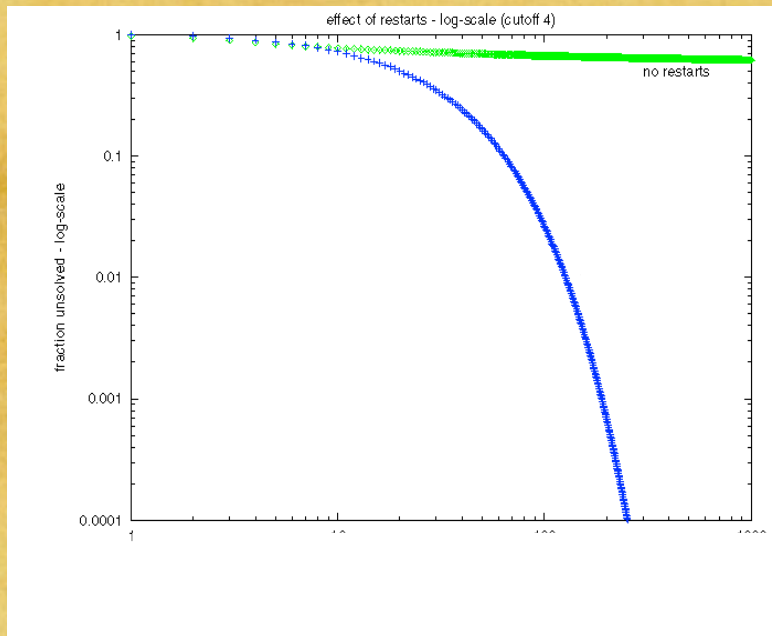
❧ Little information
for branching

❧ Little pruning

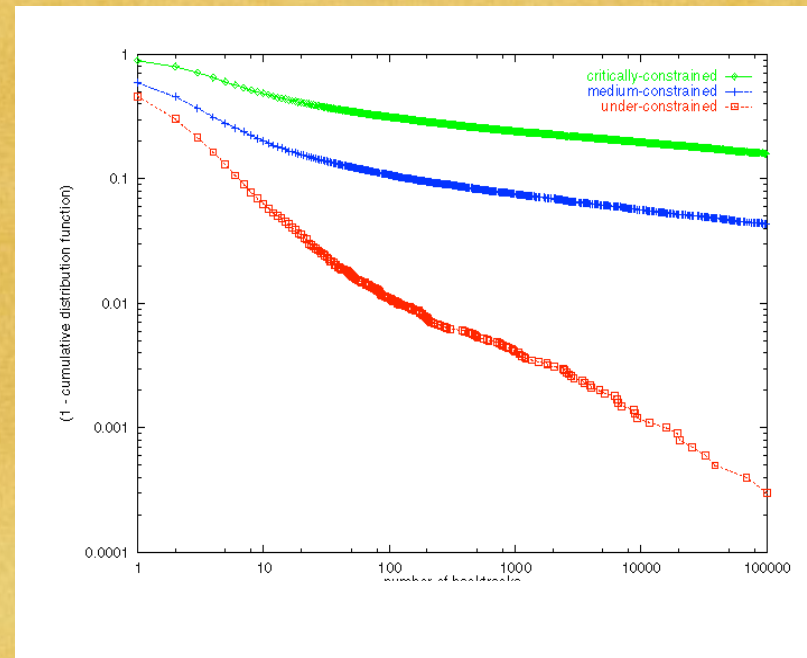
❧ Early branching
mistakes can be
very costly



Easy problems are sometimes hard



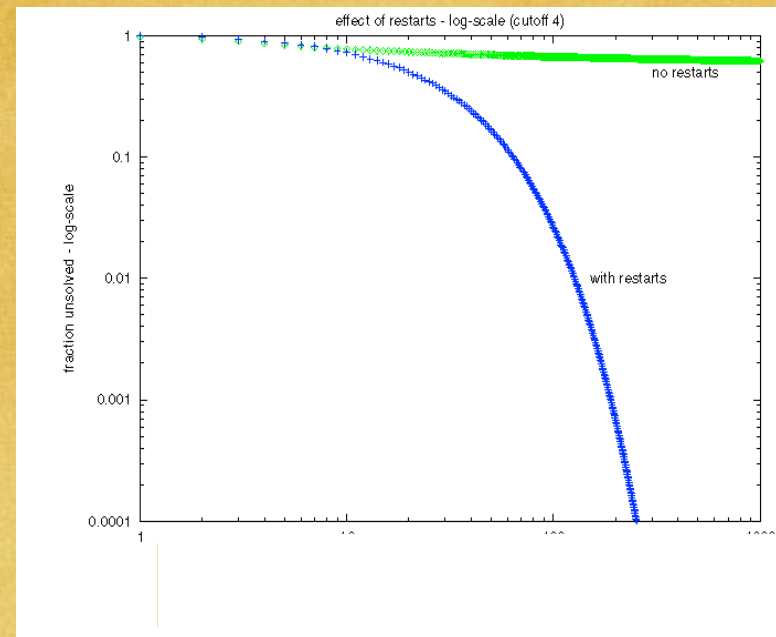
Normal exponential distribution



Heavy-tailed distribution

Two attacks on this problem!

- ⌘ Randomization & Restarts
 - ⌘ Give up and start again from the root
 - ⌘ Provably eliminates heavy-tails



[Gomes, Selman, Kautz AAI-98]

Two attacks on this problem!



- ❧ Discrepancy search
 - ❧ Discrepancy = branch against heuristic
- ❧ Several flavours
 - ❧ LDS=search according to #discrepancies
 - ❧ DDS=search according to deepest discrepancy
 - ❧ ...

[Ginsberg IJCAI 95, Walsh IJCAI 97]

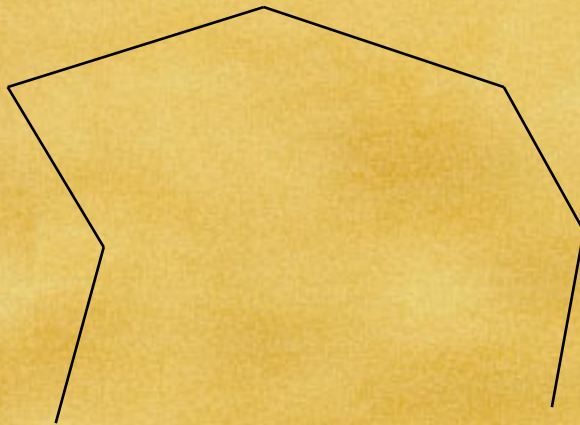
Two attacks on this problem!



Deepest Discrepancy=1

- ❧ Discrepancy search
 - ❧ Discrepancy = branch against heuristic
- ❧ Several flavours
 - ❧ LDS=search according to #discrepancies
 - ❧ DDS=search according to deepest discrepancy
 - ❧ ...

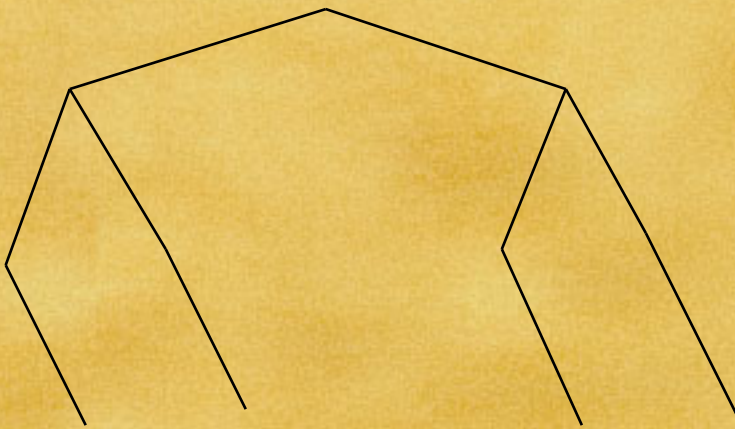
Two attacks on this problem!



Deepest Discrepancy=2

- ❧ Discrepancy search
 - ❧ Discrepancy = branch against heuristic
- ❧ Several flavours
 - ❧ LDS=search according to #discrepancies
 - ❧ DDS=search according to deepest discrepancy
 - ❧ ...

Two attacks on this problem!



Deepest Discrepancy=3

- ❧ Discrepancy search
 - ❧ Discrepancy = branch against heuristic
- ❧ Several flavours
 - ❧ LDS=search according to #discrepancies
 - ❧ DDS=search according to deepest discrepancy
 - ❧ ...



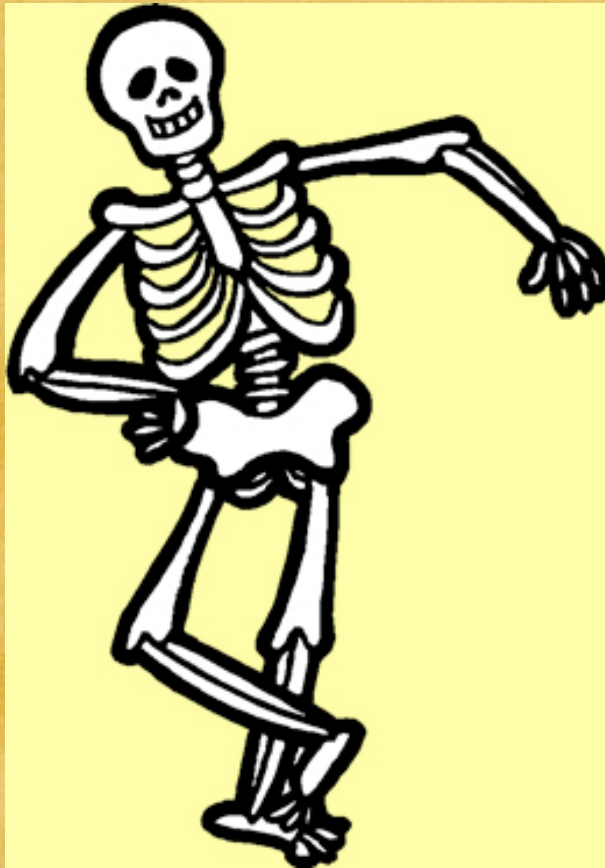
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What makes problems hard?

- ❧ Large backbones
- ❧ Absence of backdoors
- ❧ Presence of certain structures
 - ❧ Balance
 - ❧ Small worldiness
 - ❧ High degree nodes
 - ❧ ...

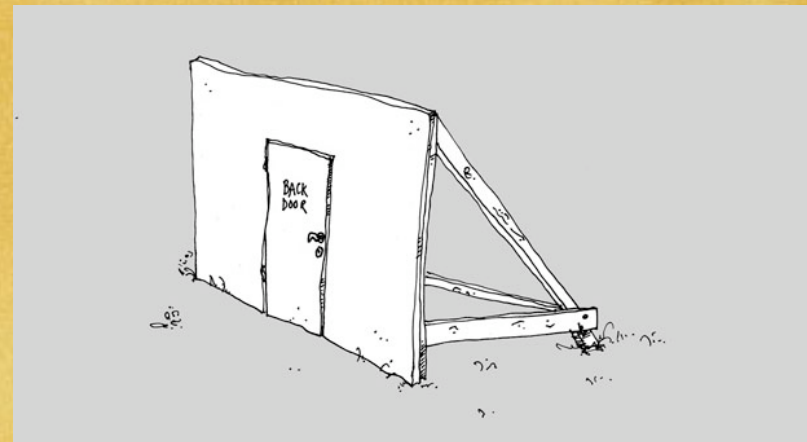
Backbone



- Fixed decisions in all solutions
 - Large backbone = high chance heuristic will get one wrong
- Problem hardness correlated with backbone size

Backdoor

- ⌘ Decisions that give polynomial subproblem
 - ⌘ Weak = particular decisions
 - ⌘ Strong = all possible decisions for a set of vars
- ⌘ Problem hardness exponential in size of backdoor
 - ⌘ c.f. fixed parameter tractability



Balance

☞ SAT

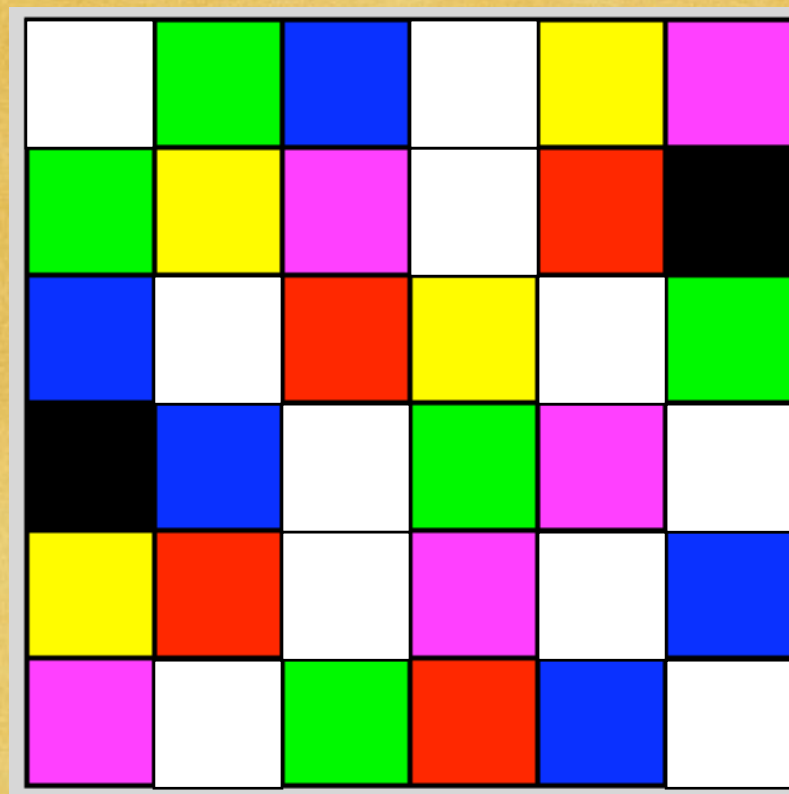
- ☞ All literals occur with same frequency

☞ COL

- ☞ All nodes have same degree

☞ Quasigroup completion problems

- ☞ Same number of holes on each row/column



Balance

☞ SAT

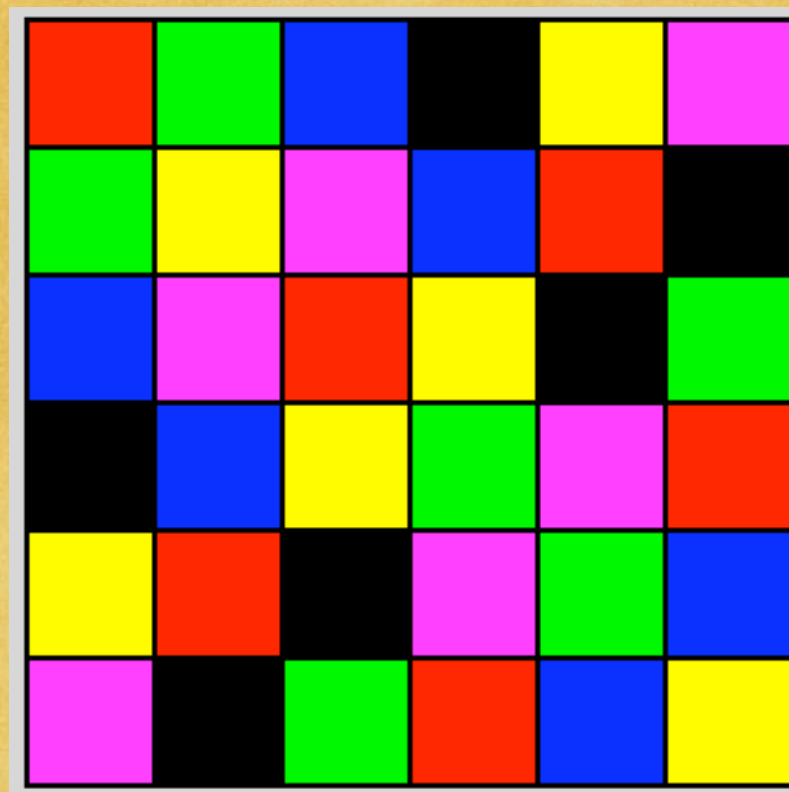
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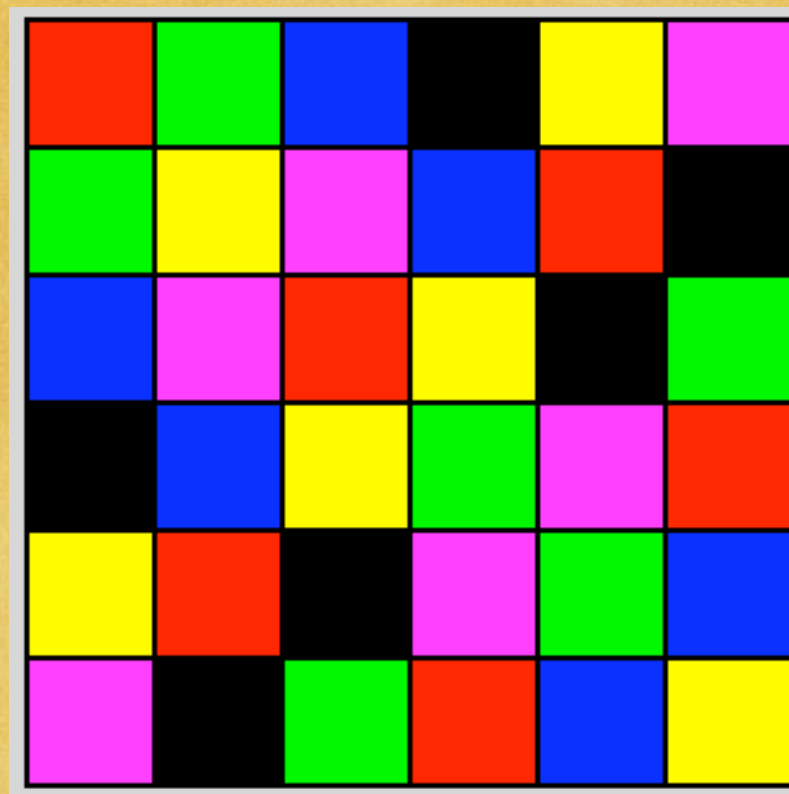
☞ Quasigroup completion problems

- ☞ Same number of holes on each row/column



Balance

- Balance can make problems harder
 - No information for branching heuristics
 - Often necessary to make problems with “hidden” solutions hard
 - Hiding solutions often makes problems very easy

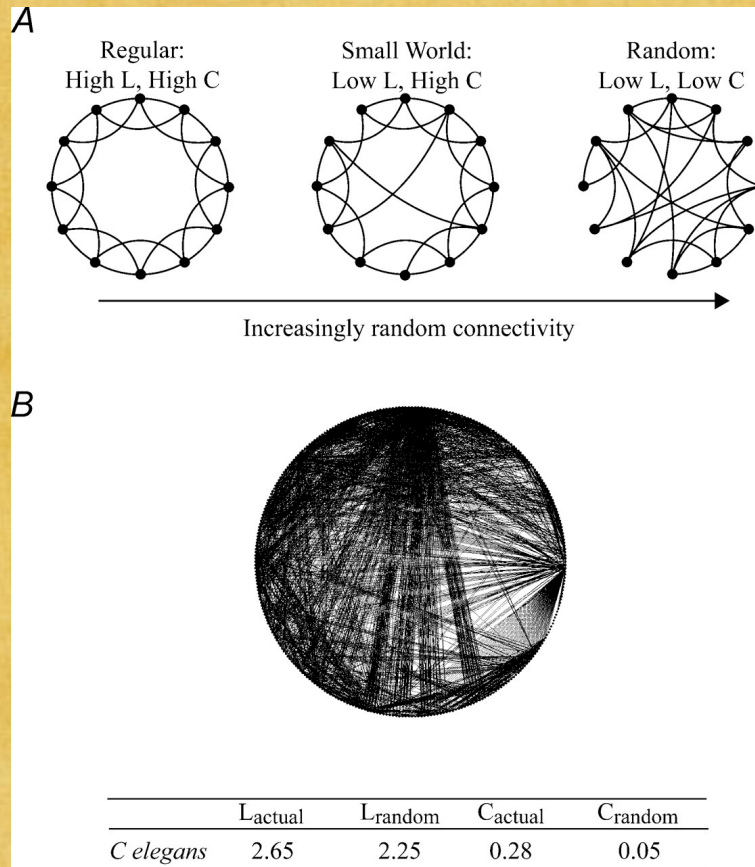


Small worlds



- ❧ Real problems are not random
 - ❧ Structure can make problems hard/easy
- ❧ For instance, graphs often have **small world** structure
 - ❧ Short path lengths
 - ❧ Highly clustered nodes

Small worlds



- ❧ Random graphs
 - ❧ Short paths
 - ❧ Not clustered
- ❧ Ring lattice
 - ❧ Long paths
 - ❧ Clustered
- ❧ Small world graph
 - ❧ Morph between ring lattice & random graph



Small worlds



☞ Impact on search

- ☞ For example, consider colouring small world graphs
- ☞ Heavy tailed distribution in search cost
- ☞ Randomization and restarts appears empirically to eliminate heavy tail!

[Walsh IJCAI-99]

Morphing

- ❧ Often have only a few real data sets
 - ❧ E.g. few voting records available
 - ❧ One such data set is voting record of 10 scientific teams for 32 Mariner trajectories
- ❧ How do we run large experiments from such data?
 - ❧ Sample this voting record
 - ❧ More/fewer voters, more/fewer candidates
 - ❧ Morph it with some randomness
 - ❧ Apply random permutations to these votes

Structure in social choice

☞ Single peakedness

- ☞ Makes manipulation harder

- ☞ Escapes GS theorem

- ☞ Fewer manipulations

☞ Balance

- ☞ Hung elections

...



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Conclusions

- ⌘ Empirical studies can complement theoretical results
 - ⌘ Hidden constants
 - ⌘ Finite size effects
 - ⌘ Theoretical conjectures
- ⌘ Empirical studies can be fun!
 - ⌘ Knife edge graph

Constrainedness knife-edge

