On the impact of small-world on local search

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Goal

Previous work [Walsh, 1999]

 CSP instances defined over 'small-world' graphs are harder to solve for complete algorithms

Question: What about local search behavior on small-world instances?

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Motivation

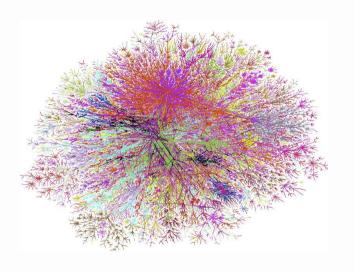
The impact of *structure* – whatever it is – on search algorithms is dramatically relevant

- Identify most difficult instances (for a given technique)
- Understand why an instance is difficult
- Exploit this bit of information to choose the best solver, or a combination of solvers
- Evaluate the quality of benchmarks

Outline

- Background: Complex networks
- Structure in CSPs
- Small-world SAT instances
- Experimental results
- Discussion

Complex networks



Graphs as structure abstraction

- Entities represented as graph nodes
- Node: either one entity or an entire subsystem

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Main characteristics

- Node degree (distribution, average, etc.)
- Diameter, characteristic path length et similia
- Clustering (i.e., cliquishness tendency)

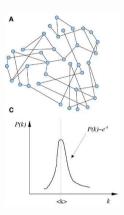
Complex networks

- System topology is crucial for understanding its dynamics
- Graph theory provides useful models
- Complex networks: emerging research field

Random graphs

- First developed model for system structure
- Several important applications
- Random graphs fail to represent social and biological systems

Random graphs



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Random graphs

- Node degree distribution: Poissonian (approx Normal)
- Characteristic path length: low
- Clustering: low

Scale-free networks

- Relations among individuals in a society (e.g., scientific collaborations)
- Web pages structure
- Internet structure
- **.**...

Scale-free networks

- Node degree distribution: nodes with degree $k \sim k^{-\gamma}$ (γ parameter)
- Very few hubs (but not negligible) and many nodes with few connections
- Robust wrt random failures
- Sensitive to attacks

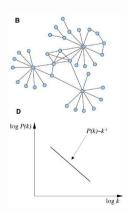
Scale-free networks formation

- Growth: older nodes has on average a higher number of connections
- Preferential attachment: new nodes are more likely to connect to nodes with higher degree (probability proportional to the degree)
- Model variants that take into account also the fitness

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Scale-free networks

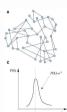


Small-world

- Any pair of nodes connected by few hops (short characteristic path length)
- High degree of cliquishness (high clustering coefficient)
- Examples:
 - Social networks
 - World Wide Web
 - Scientific collaboration network
 - C.Elegans worm neural network

Characteristic length

Informally: average path length between any pair of nodes.



Random graphs \rightarrow short



Grid graphs → long

Structure

- Diverse meanings
- Structure vs. random
- Usually real world problems are said to be structured
- Attempts to define quantitative measures (entropy, compression ratio, etc.)
- Graph representation of relations among problem entities

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Clustering



Informally: it quantifies the probability that, given node a connected to b and c, there is an edge between b and c



Random graphs → low

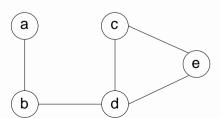


Grid graphs → high

SATgraphs

$$(a \vee \neg b) \wedge (b \vee d) \wedge (c \vee \neg d \vee \neg e)$$





Remember the initial goal...

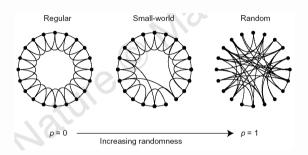
Previous work [Walsh, 1999]

 CSP instances defined over 'small-world' graphs are harder to solve for complete algorithms

Question: What about local search behavior on small-world instances?

Small-world SAT

► Morphing between a lattice SAT instance and a random SAT instance. [Gent et al., 1999]



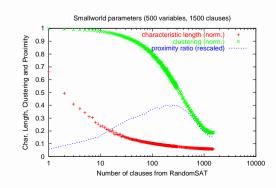
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Experimental issues

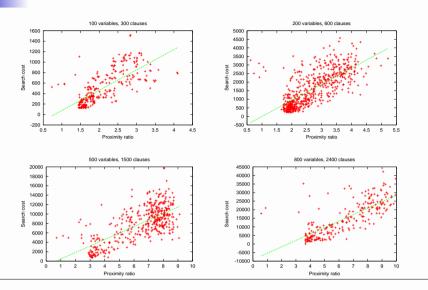
- Small-world SAT instances
 - Procedure to generate instances
 - Measuring 'small-world' property
- Attacking the benchmark with local search algorithms
 - GSAT
 - WalkSAT
 - ILS-SAT

Small-world SAT

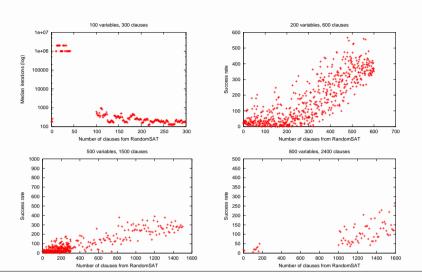
Length, clustering and proximity ratio (normalized ratio clustering/length)



Complete algorithm



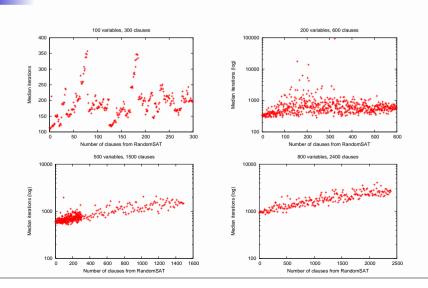
GSAT



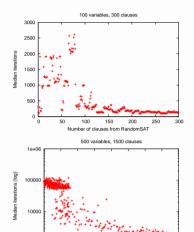
Outline of the results

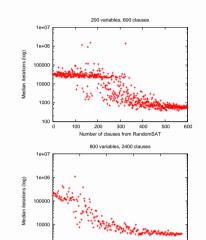
- No common behavior across different algorithms
- 'Mild' tendency of small-world and hardness correlation

WalkSAT



ILS-SAT





1000

Number of clauses from RandomSAT

1500

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Discussion

200 400 600 800 1000 1200 1400 1600

- Many small-world/lattice SAT instances are harder for GSAT and ILS-SAT
- WalkSAT exhibits a peculiar behavior
- The relation between SATgraph and search landscape plays a very important role

Future work

- Connections between constraint graph properties and search space characteristics
- Exploring strengths and weaknesses of the heuristics w.r.t. constraint graph properties
- Relation between problem encoding and graph properties
- Alternative formulations to study the structure of a problem can be used (e.g., weighted graphs)