

# A Modularity-based Random SAT Instances Generator

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IJCAI 2015  
July 31, 2015

# Industrial SAT Instances

- The **Boolean Satisfiability Problem (SAT)**.
- Many **real-world** problems are **efficiently** solved by **modern SAT solvers**.
  - **Conflict-Driven Clause Learning (CDCL)** SAT solvers.
- **Industrial SAT Instances**:
  - Problems encodings from **real-world** applications.
  - No **precise definition/model**: crypto, bmc, scheduling, planning, ...
  - **Heterogeneity**.
  - **Finite** and **reduced** number of instances.

# Realistic Pseudo-Industrial SAT Instances Generators

- The **generation** of **realistic** random **pseudo-industrial** SAT instances:
  - [SelmanKautzMcAllester97]:  
**CHALLENGE 10:** *Develop a **generator** for problem instances that have **computational properties** that are **more similar to real world instances**.*
  - [KautzSelman03]
  - [Dechter03]
- **Need:** **Testing** and **Debugging** Purposes.
  - Desired **size**.
  - Desired **hardness**.
  - Desired **properties**.
- **Approach:** **Analysis** of SAT instances.
  - General **common properties**.
  - **Isolate** some of them.

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# The Community Structure of Graphs

- A graph has clear **community structure** if its nodes can be grouped into communities such that its edges mostly connect nodes of the same community.
- The **modularity**  $Q$  [NewmanGirvan04] of a graph  $G$  and a partition  $C$  of its nodes measures the *fraction of internal edges* (w.r.t. to a random graph with same nodes and same degrees).

$$Q(G, C) = \sum_{C_i \in C} \frac{\sum_{x, y \in C_i} w(x, y)}{\sum_{x, y \in V} w(x, y)} - \left( \frac{\sum_{x \in C_i} \text{deg}(x)}{\sum_{x \in V} \text{deg}(x)} \right)^2$$

where  $G = (V, w)$  and  $\text{deg}(x) = \sum_{y \in V} w(x, y)$ .

- The **modularity** of a graph is the **maximal** modularity for **any possible partition**:  $Q(G) = \max\{Q(G, C) | C\}$ .
- The (**optimal**) modularity ranges in the interval **[0, 1]**.

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# The Community Structure of Industrial SAT Instances

- **SAT Instances** as Graphs.
  - The **Variable Incidence Graph (VIG)**:
    - **Nodes** are variables.
    - **Edges** between two variables in the same clause.
    - **Weights** to give the same relevance to all clauses.
  
  - **Industrial SAT instances** have a **clear community structure**.
  - Their **modularity** has values **greater than 0.7** in most cases (**random** SAT instances have a modularity **smaller than 0.3**).
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# Community Attachment Generator (I)

- **Classical Random**  $k$ -CNF model:  $F_k(n, m)$ 
  - $n$ : #variables,  $m$ : #clauses,  $k$ : clause size
- **Community Attachment** model:  $F_k(n, m, c, P)$ 
  - $c$ : #communities (each community has  $n/c$  variables)
  - Probability  $P$
- For *each clause*, **repeat**:
  - With probability  $P$ : all literals in the **same community**.
  - With probability  $1 - P$ : all literals in **distinct communities**.
  - Communities randomly chosen among  $c$ .
  - Variables randomly chosen among  $n/c$ .
  - Polarities randomly assigned.
  - **Avoiding trivial clauses** (repeated literals or tautologies).
- **Restriction**:  $k \leq c \leq n/k$ 
  - There always exists at least **one possible clause to select**.

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- **Community Attachment** model:  $F_k(n, m, c, P)$
- **Theorem.** Given a SAT instance  $\Gamma \in F_k(n, m, c, P)$ , let  $G$  be its VIG. The **average modularity** of  $G$  is bounded as:

$$E[Q(G)] \geq P - \frac{1}{c}$$

- When  $P$  is **big enough**, the modularity is very close to this lower-bound. Therefore, we simply take:

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# Community Attachment Generator (III)

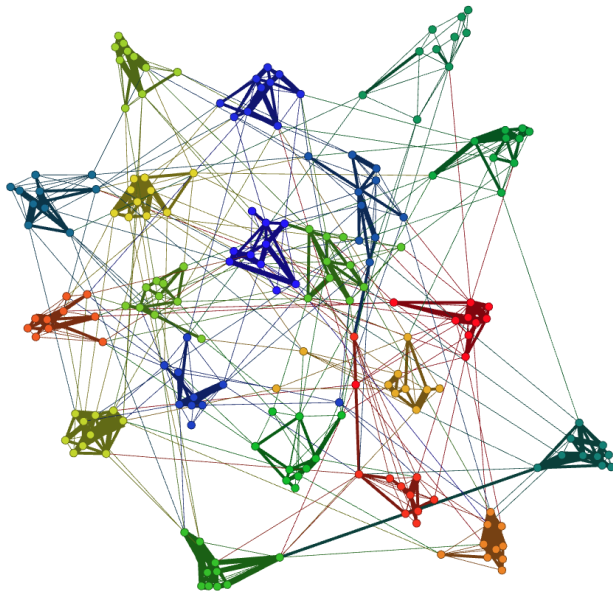
- **High modularity:**  
more adequate to model **real-world** problems.
- **Low modularity:**  
more similar to classical **random** problems.

**This generator is publicly available at:**

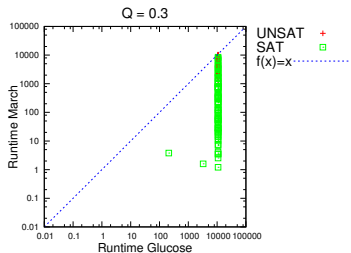
**<http://www.iiia.csic.es/~jgiraldez/software>**

# Example

200 variables  
850 clauses  
 $Q = 0.8$   
20 communities

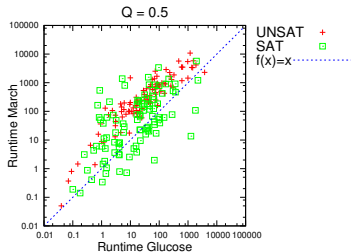
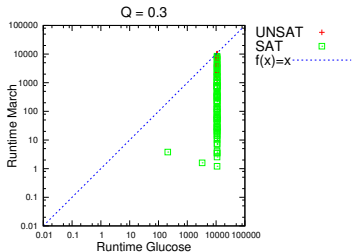


Glucose (*industrial*) vs March (*random*):



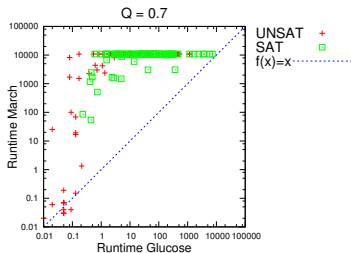
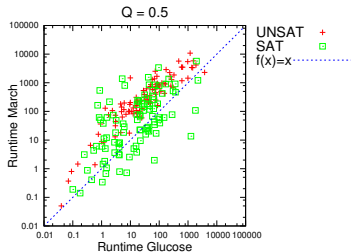
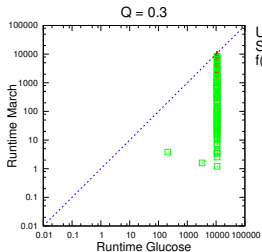
# SAT Solvers Performance

Glucose (*industrial*) vs March (*random*):



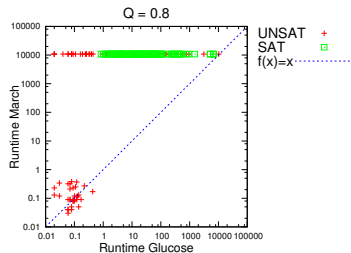
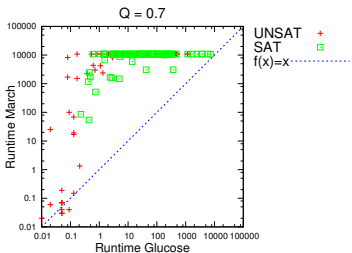
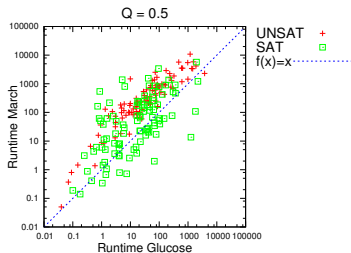
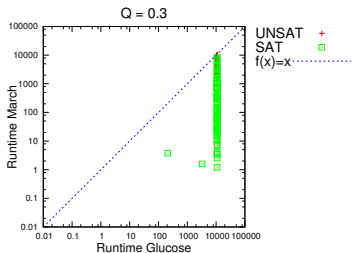
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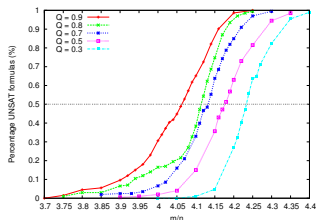
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# The Phase Transition Point

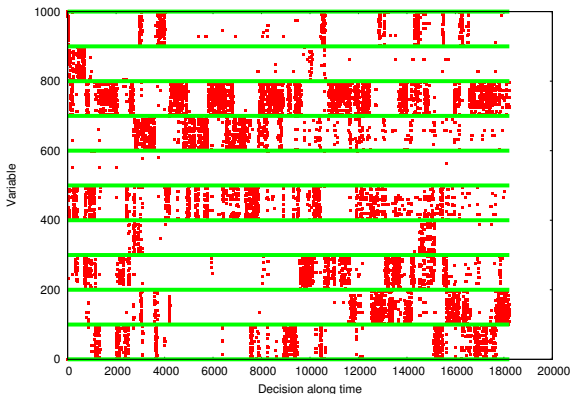
- Finite size (*Observations*):  
The **phase transition point** **SAT-UNSAT**, if exists, **decreases** for higher values of **modularity**.



- Infinite size (*Theorem*):  
The **phase transition point** **SAT-UNSAT**, if exists, does **not depend** on the **modularity**.
  - Proof.

# Analyzing the SAT Solver

- 1000 variables and 10 communities.



- Using **community structure** to **improve** the performance of CDCL SAT solvers:  
[AnsóteguiGiráldezLevySimon] to *appear* in **SAT 2015**.



# Thank you for your attention!

## Poster Panel #49

I am looking for postdocs opportunities  
`jgiralde@iia.csic.es`