

# The Fractal Dimension of SAT Formulas

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# Known Facts from SAT Community

- **Random** and **industrial** formulas: **distinct nature**.
  - SAT competitions: different tracks.
- SAT solvers **specialize**.
- Many **very large industrial instances solved efficiently** by modern SAT solvers (**CDCL**).
  - Good performance: ability to exploit some **hidden structure**.

# SAT Instances

## ■ Random $k$ -CNF:

- Its **definition** is **clear**.
- Generate  $k$ -CNF of  $n$  vars and  $m$  clauses:  
for  $i$  in  $1..m$   
    Select randomly  $k$  literals among  $n$   
    with random polarity
- **Theoretical** point of view.

## ■ Industrial CNF:

- Problems encodings from **real-world** applications.
- No **precise definition**: crypto, bmc, scheduling, planning, ...
- **Heterogeneity**.

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- **Heterogeneity**.

# [Some] Open Questions in SAT

- **Open Question #1:** What is exactly the **structure** of industrial formulas?
- **Open Question #2:** How SAT solvers (can) **exploit** this structure?

# Complex Networks

## ■ The **classical Erdős-Rényi model**:

- Generate a graph of  $n$  nodes and  $m$  edges:  
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    Select randomly 2 nodes among  $n$
- These networks cannot be used for representing many **real-world** networks.

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- **Features**: Clustering coefficient, Modularity, ...
- **Models**: Small-world, Scale-free, ...
- Methods of **generation**: Preferential attachment, ...

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## ■ Industrial CNF: ?



# Complex Networks vs SAT

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- **Industrial CNF:** ?

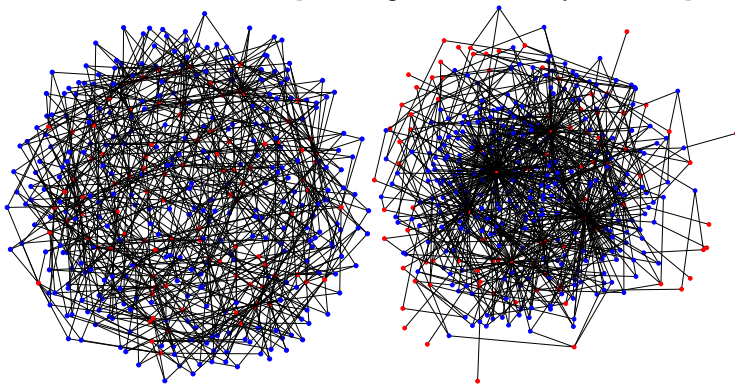
# [Some] Open Questions in SAT

- **Open Question #1:** What is exactly the **structure** of industrial formulas?
- **Open Question #2:** How SAT solvers (can) **exploit** this structure?
- Many works in terms of **complex networks** trying to **answer** these questions.

# Previous Work (I)

- **Open Question #1:** What is exactly the **structure** of industrial formulas?

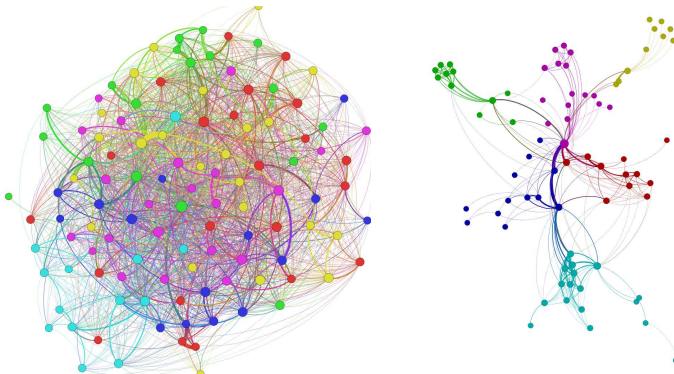
Scale-free Structure [Ansótegui, Bonet, Levy. CP2009]



# Previous Work (II)

- **Open Question #1:** What is exactly the **structure** of industrial formulas?

Community Structure [Ansótegui, Giráldez-Cru, Levy. SAT2012]





# Motivations

- **Analysis** of the structure of industrial SAT instances.
- **Generators** of more realistic industrial-like SAT formulas.
- (Possible) **improvements** in SAT solving techniques.

# Outline

1 Introduction

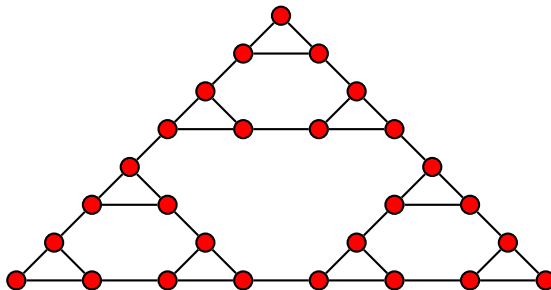
2 The Fractal Dimension of Graphs

3 The Fractal Dimension of SAT Formulas

4 Conclusions

# Intuition

A graph has **fractal dimension** (it is **self-similar**) if it keeps the **same shape** after *rescaling*.





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0.5,0.151.5,1.152.5,0.152.5,2.153.5,3.154.5,2.154.5,0.155.5,1.156.5,

# Intuition

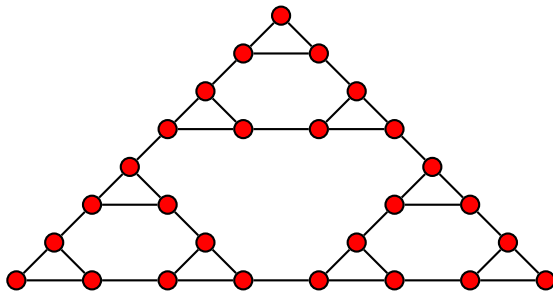
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1.5,0.553.5,2.555.5,0.550.5,0.151.5,1.152.5,0.152.5,2.153.5,3.154.5,

# Computing the Fractal Dimension (I)

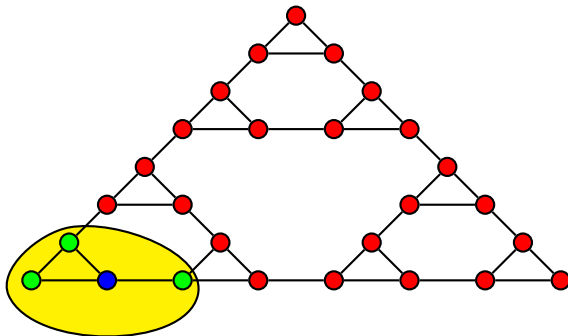
- **[Def.]** A **circle** of radius  $r$  and center  $c$  is a subset of nodes of the graph such that the distance between any of them and the node  $c$  is strictly smaller than  $r$ .
- **[Def.]** Let  $N(r)$  be the minimum number of circles of radius  $r$  required to cover a graph.
  - $N(1) = n$
  - $N(d^{max} + 1) = 1$

# Computing the Fractal Dimension (II)



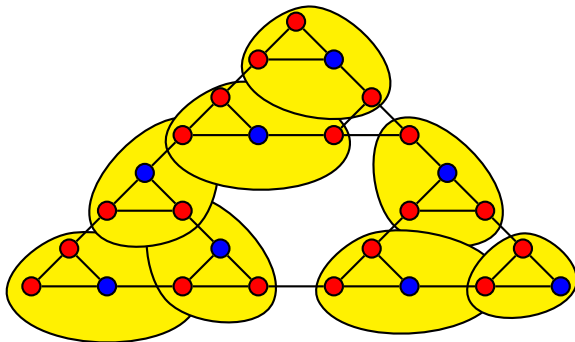
$r$	$N(r)$	
1	27	#nodes
2		
3		
4		
5		
6		
7		
8	1	$d^{max} = 7$

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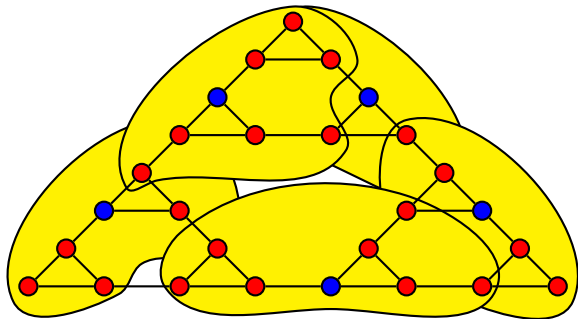
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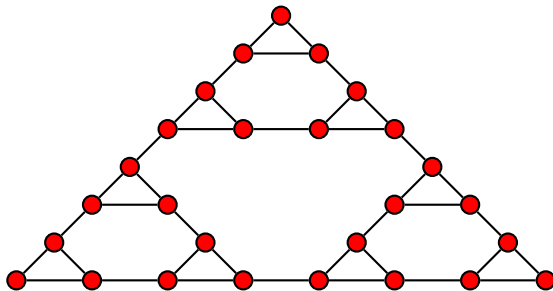
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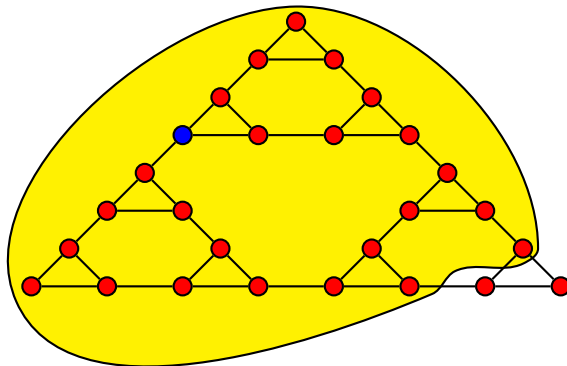
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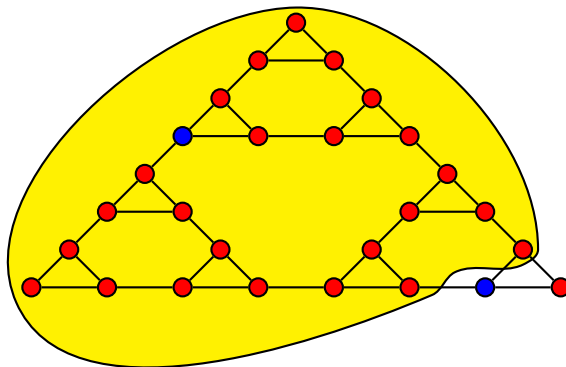


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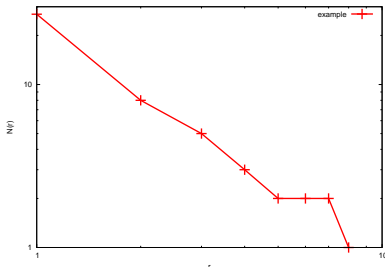
- [Def.] (*Hausdorff*) A graph has the **self-similarity** property if the function  $N(r)$  decreases polynomially.
- I.e.  $N(r) \sim r^{-d}$ , for some value  $d$ .
- In the case, we call  $d$  the **dimension** of the graph.

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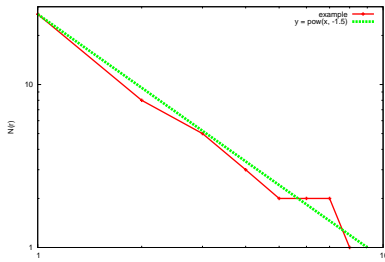
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# Computing the Fractal Dimension (IV)

- **[Lemma]** Computing the function  $N(r)$  is **NP-hard**.
  - Reducing *GraphCOL* to  $N(2)$ .
- **Burning** algorithms:
  - More efficient algorithms (**greedy**).
  - Approximate **upper bounds** of  $N(r)$ .
  - Select the circle that covers (**burns**) the maximal number of uncovered (**unburned**) nodes.
  - Further approximations needed to make the algorithms of **practical use in very large graphs**.
- The **Burning by Node Degree (BND)** algorithm.

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# The Burning by Node Degree (BND) Algorithm

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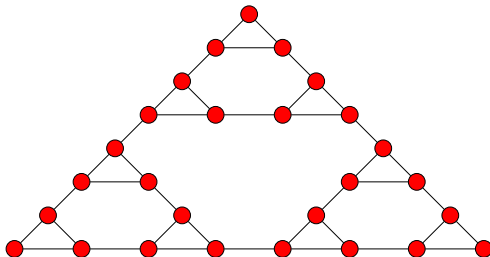
## Algorithm 1 Burning by Node Degree (BND)

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```
1: Input: Graph  $G = (V, E)$ 
2: Output: vector[int]  $N$ 
3:  $N[1] := |V|$ 
4: int  $i := 2$ 
5: while  $N[i - 1] > \text{connectedComponents}(G)$  do
6:   vector[bool]  $\text{burned}(|V|)$ 
7:    $N[i] := 0$ 
8:    $\text{burned} := \{\text{false}, \dots, \text{false}\}$ 
9:   while  $\text{existsUnburnedNode}(\text{burned})$  do
10:     $c := \text{highestDegreeUnburnedNode}(G, \text{burned})$ 
11:     $S := \text{circle}(c, i)$ 
12:    for all  $x \in S$  do
13:       $\text{burned}[x] := \text{true}$ 
14:    end for
15:  end while
16:   $i := i + 1$ 
17: end while
```

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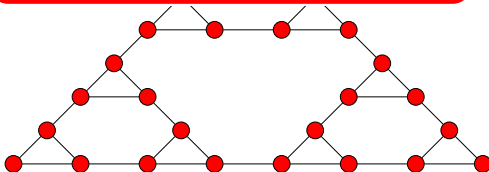
# Example



$r$	$N^{real}(r)$	$N^{BND}(r)$
1	27	27
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# Example

**BND gives upper bounds of  $N(r)$**

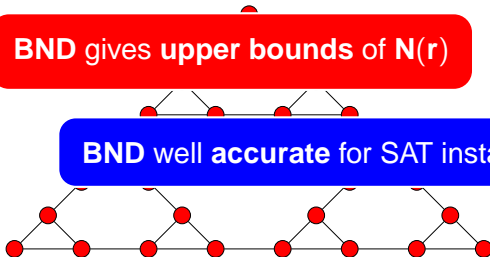


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**BND well accurate for SAT instances**



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# Fractal Dimension vs Diameter

- Determines the **maximal radius**  $r^{max}$ .
- Related to the **diameter**:  $r^{max} \leq d^{max} \leq 2r^{max}$
- **Diameter**:
  - **Dependent** on the size of the graph.
  - Quite **expensive** to compute in practice.
- **The fractal dimension**:
  - **Independent on the size**. Families with similar  $N(r)$  function shape.
  - It can be **computed more efficiently** than the diameter.

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We propose the use of the **Fractal Dimension**

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2 The Fractal Dimension of Graphs

3 The Fractal Dimension of SAT Formulas

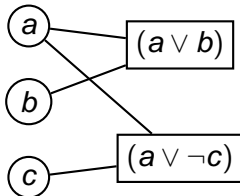
4 Conclusions



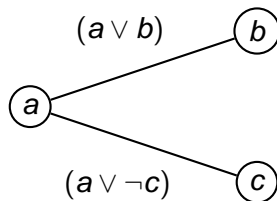
# SAT Formulas as Graphs

$$\sigma = (a \vee b) \wedge (a \vee \neg c)$$

**Clause-Variable Incidence Graph (CVIG)**



**Variable Incidence Graph (VIG)**



# The Relation between VIG and CVIG

**VIG** $N(r)$ **CVIG** $N^b(r)$ **[Lemma]**

$$\text{If } N(r) \sim r^{-\mathbf{d}} \implies N^b(r) \sim r^{-\mathbf{d}}$$

$$\text{If } N(r) \sim e^{-\beta r} \implies N^b(r) \sim e^{-\frac{\beta}{2} r}$$

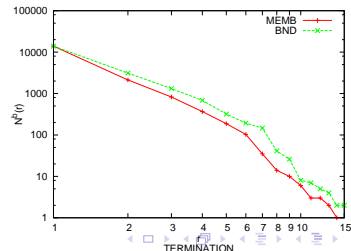
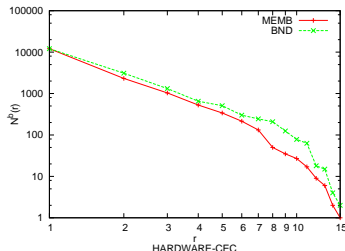
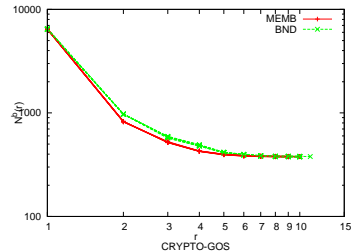
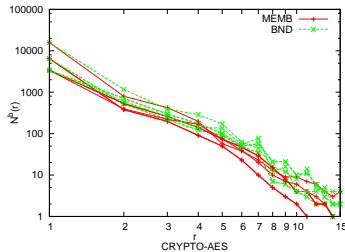
# The Accuracy of the BND Algorithm (I)

	BND	MEMB <sup>1</sup>
#solved	300	17
av. of runtime	0.11sec	10min 7.2sec
$N^b(r)$	Very similar values	

Set: 300 industrial instances of the SAT Competition 2013

<sup>1</sup>[Song et al. Journal of Statical Mechanics (2007)]

# The Accuracy of the BND Algorithm (II)



# Known Results for Random 2CNF Formulas

- **Phase transition point** at  $m/n = 1$ .
- VIG's of **random 2CNF** formulas = **Erdős-Rényi** graphs.
- **Percolation threshold** at  $m/n = 0.5$ .
  - In this point, **self-similar** ( $d = 2$ ).
  - **Above** this point  $N(r)$  decays **exponentially**.
- To the best of our knowledge, there is **no known result for random 3CNF instances**.

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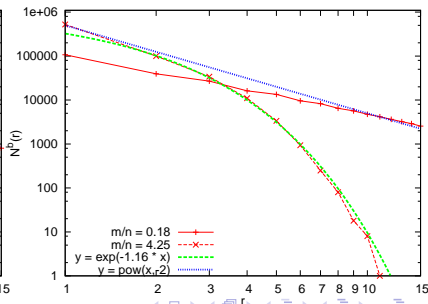
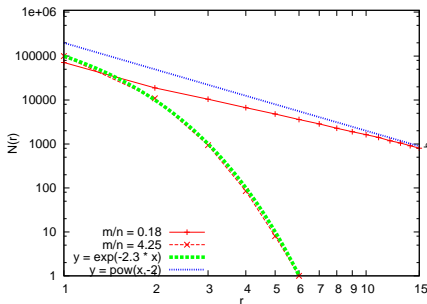
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# Random 3CNF Formulas

- Experimentally,  $N(r)$  (and  $N^b(r)$ ) only depends on the clause/variable ratio  $m/n$  (and not on the number of variables  $n$ ).
- Phase transition point ( $m/n \approx 4.25$ ):
  - $N(r) \sim e^{-2.3r}$  and  $N^b(r) \sim e^{-1.16r}$
  - Hence, the decay of CVIG is just half of the decay of VIG (as expected)
- (Experimentally) Percolation threshold at  $m/n \approx 0.17$ ,  $d = 2$

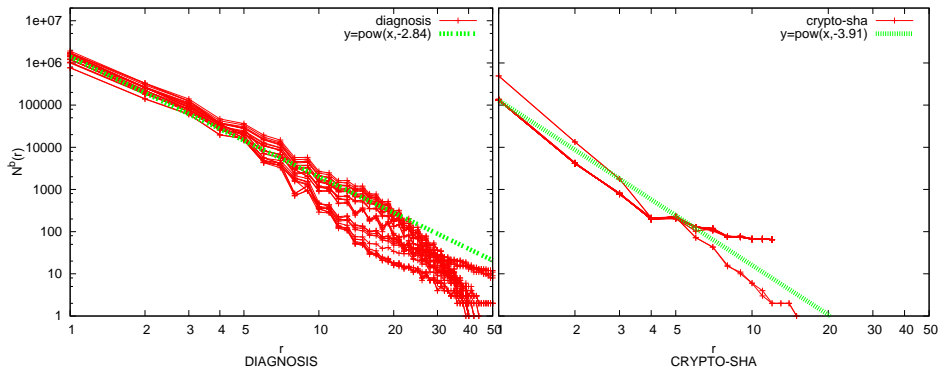






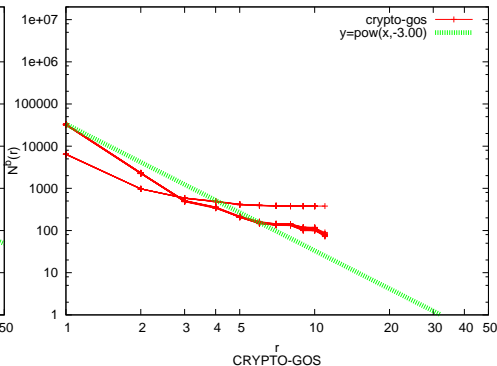
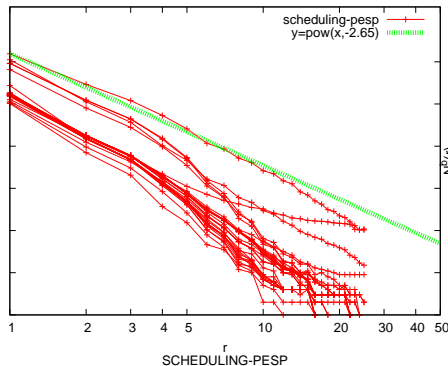


# Industrial SAT Formulas (II)



- Family *diagnosis*:  $d \approx 2.84$  (26 instances)
- Family *crypto-sha*:  $d \approx 3.91$  (30 instances)

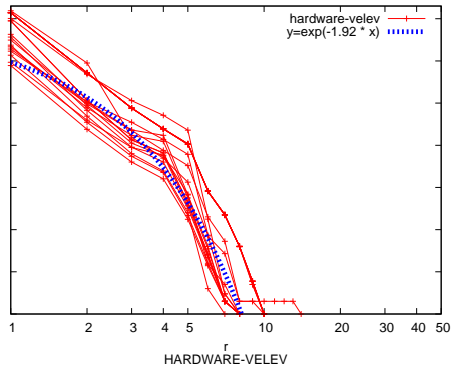
# Industrial SAT Formulas (III)



- Family *scheduling-pesp*:  $d \approx 2.65$  (30 instances)
- Family *crypto-gos*:  $d \approx 3.00$  (30 instances)

# Industrial SAT Formulas (IV)

- In some families, all instances have a  $N(r)$  function with exponential decay, i.e. they are **not self-similar**.

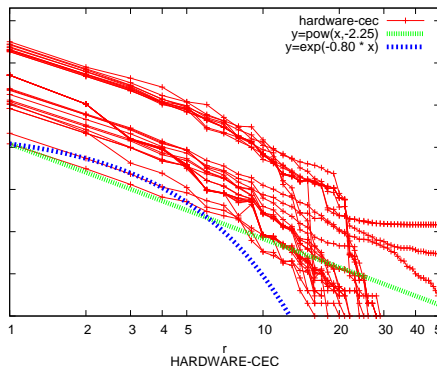


# Analyzing the Fractal Dimension (I)

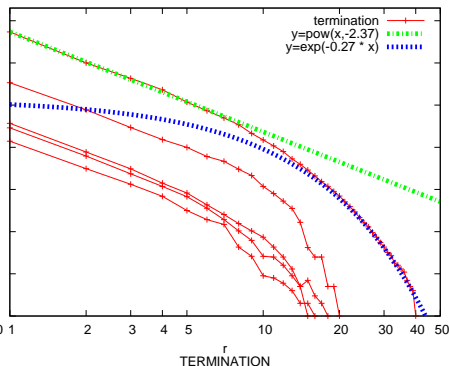
We identify **two phenomena** (only in some cases):

- 1 **Abrupt decay** (but no exponential function).

# Analyzing the Fractal Dimension (II)



HARDWARE-CEC



TERMINATION

- Family *hardware-cec*:  $d \approx 2.25$  (30 instances)
- Family *termination*:  $d \approx 2.37$  (5 instances)

# Analyzing the Fractal Dimension (III)

We identify **two phenomena** (only in some cases):

- 1 **Abrupt decay** (but no exponential function).
  - **Small number of edges connecting distant areas** of the graph.
    - No effect for small values of  $r$ .
    - They may drop down the number of circles for big values of  $r$ .
  - Existence of **non-local** dependencies.
- 2 **Long tail**.

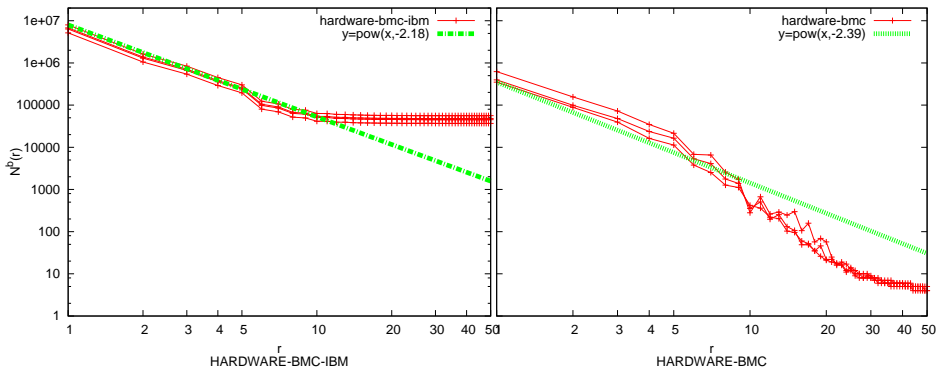


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  - **Small number of edges connecting distant areas** of the graph.
    - No effect for small values of  $r$ .
    - They may drop down the number of circles for big values of  $r$ .
  - Existence of **non-local** dependencies.
- 2 **Long tail**.

# Analyzing the Fractal Dimension (IV)



- Family *hardware-bmc-ibm*:  $d \approx 2.18$  (4 instances)
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# Analyzing the Fractal Dimension (V)

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- Existence of (small) **unconnected components**.
- Removed after preprocessing.

# Outline

- 1 Introduction
- 2 The Fractal Dimension of Graphs
- 3 The Fractal Dimension of SAT Formulas
- 4 Conclusions

# Summary

- **FD** related to **diameter**, but **more stable** (independent on the size).
- **BND**: **efficient** computation of FD in very **large graphs** (as SAT instances).
- Most industrial SAT instances are **self-similar**:  $2 \leq d \leq 4$ .
- **Random** SAT formulas are clearly **not self-similar**.
- **Learning** does **not** contribute to **connect distant parts** of the formula (as one could think) [See details in Section 5].
- **Future work**: **Generators** of **more realistic industrial-like SAT instances** take into account the **fractal dimension**.

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# Thank you for your attention!

# The Fractal Dimension of SAT Formulas

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