

# What is this Unique Games Conjecture that everyone keeps talking about?

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# Agenda

- 1 The conjecture
- 2 Implications
  - Assorted graph problems
  - Constraint satisfaction
- 3 Progress?

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- An edge  $e = (v, w)$  is satisfied if  $(l_v, l_w) \in R_e$ .
- We denote by  $\text{Val}(X) \in [0, 1]$  the maximum possible fraction of satisfied edges.

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Thus it is NP-hard to determine if we can satisfy all edges.

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- This is an instance of Label Cover with label set  $L = \mathbb{Z}_n$ .

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Does this seem familiar?

# Two-prover games

- Consider a two-prover game:
  - two provers try to convince us (the verifier) that some 3-CNF formula is satisfiable
  - we send a question to each of the provers, which they have to answer independently

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- Finding a good strategy for the provers is equivalent to solving the label cover problem



# Approximating Label Cover

## Theorem

*For every  $\eta > 0$  it is NP-hard to distinguish Label Cover instances with  $\text{Val}(X) = 1$  from those with  $\text{Val}(X) \leq \eta$*

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(Follows from PCP theorem and Parallel Repetition theorem)

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- Our example with linear equations is a Unique Label Cover problem if  $n$  is prime
- How hard is Unique Label Cover?



# The Unique Games Conjecture

## Conjecture (Khot, 2002)

*For any  $\eta > 0$  there is an  $L$  such that it is NP-hard to distinguish Unique Label Cover instances with  $\text{Val}(X) \geq 1 - \eta$  from those with  $\text{Val}(X) \leq \eta$  for  $X$  with label set  $L$ .*

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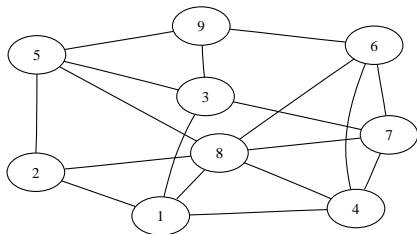
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- UG-hard  $\Rightarrow$  NP-hard assuming the UGC

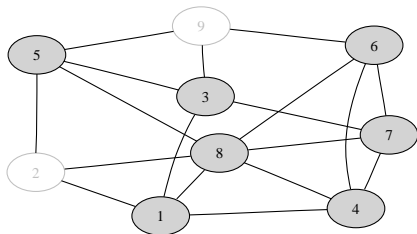
# Min Vertex Cover

- Given a graph  $G = (V, E)$ , a set  $C \subseteq V$  is a vertex cover if for each edge  $\{v, w\}$ , at least one of  $v$  and  $w$  is in  $C$



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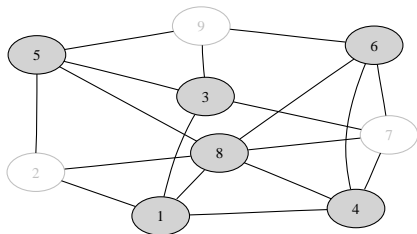
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- Goal: find a vertex cover of minimum cardinality



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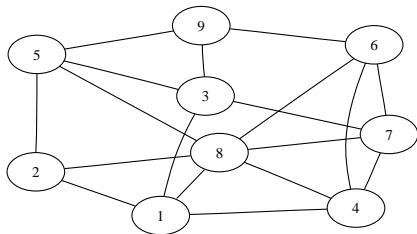
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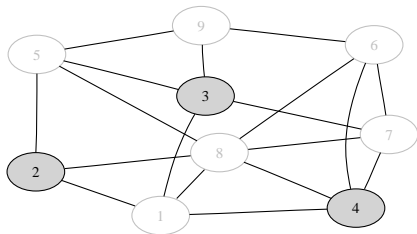
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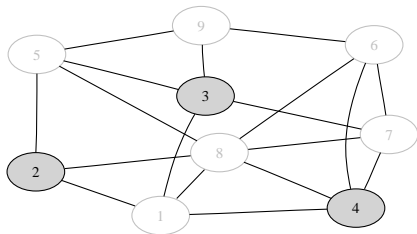
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  - Think of  $\Delta$  as big, but constant
  - NP-hard for any constant  $\Delta \geq 3$





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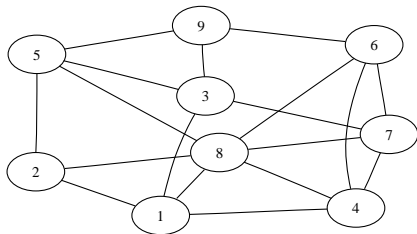
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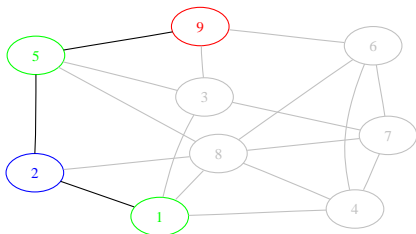
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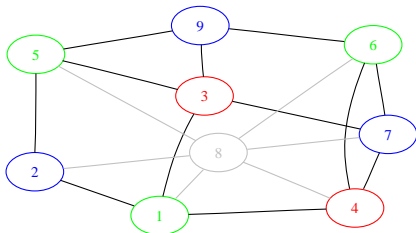
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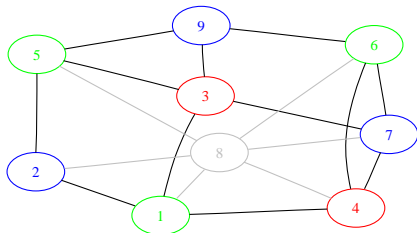
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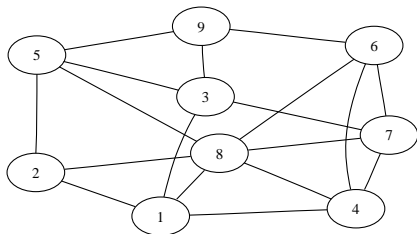
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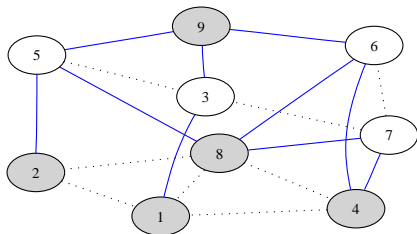
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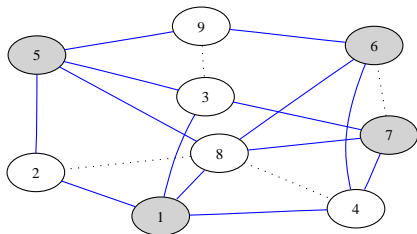
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- Given a 2-CNF formula, find an assignment which satisfies as many clauses as possible

$$\begin{aligned}(\overline{x_1} \vee x_2) \wedge (x_1 \vee \overline{x_3}) \wedge (x_1 \vee x_4) \wedge \\(\overline{x_1} \vee x_8) \wedge (\overline{x_2} \vee x_5) \wedge (\overline{x_2} \vee \overline{x_8}) \wedge \\(x_3 \vee x_5) \wedge (x_3 \vee x_7) \wedge (\overline{x_3} \vee \overline{x_9}) \wedge \\(\overline{x_4} \vee x_6) \wedge (x_4 \vee \overline{x_7}) \wedge (\overline{x_4} \vee \overline{x_8}) \wedge \\(\overline{x_5} \vee x_8) \wedge (x_5 \vee x_9) \wedge (\overline{x_6} \vee \overline{x_7}) \wedge \\(x_6 \vee \overline{x_8}) \wedge (\overline{x_6} \vee x_9) \wedge (x_7 \vee \overline{x_8})\end{aligned}$$



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- Austrin, 2006: UG-hard to approximate within  $\alpha_{LLZ} + \epsilon$  for any  $\epsilon$

# Max $k$ -CSP

- In the Max  $k$ -CSP problem we are given a set of constraints, each of which acts on at most  $k$  variables

$$\begin{aligned} ((x_1 \vee \overline{x_2}) \wedge x_4) & \wedge (x_3 \oplus x_4 \oplus x_5) & \wedge \\ (x_1 \vee x_4) & \wedge (x_2 \wedge \overline{x_4} \wedge x_1) & \wedge \\ (x_2 \vee (x_3 \oplus x_5)) & \wedge (\overline{x_1} \wedge \overline{x_3}) \end{aligned}$$

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  - For  $k = 2^d - 1$ , hard to approximate within  $(k + 1)/2^k$

# Random Max $k$ -CSP

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# Agenda

- 1 The conjecture
- 2 Implications
  - Assorted graph problems
  - Constraint satisfaction
- 3 Progress?

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- Charikar, Makarychev and Makarychev, 2006: The UGC is *not* true for  $|L| \leq (\Omega(1/\eta))^{2/\eta}$ .

# Inapproximability of Unique Label Cover

## Theorem (Feige and Reichman)

*For any  $\eta > 0$  there is a  $0 < \gamma < 1$  and  $L$  and such that it is NP-hard to distinguish between Unique Label Cover instances with  $\text{Val}(X) \geq \gamma$  and  $\text{Val}(X) \leq \eta \cdot \gamma$  for  $X$  with label set  $L$ .*

## Proof (1/3)

Recall

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- Take a prime  $p \geq 1 + |L|^2/\eta$ , and identify the elements of  $L$  with elements of  $\mathbb{Z}_p$  in some arbitrary way.

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# Questions!