

Testing Graph Properties with the Container Method

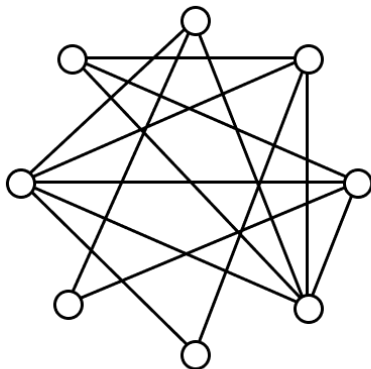
Eric Blais Cameron Seth

June 5 2024



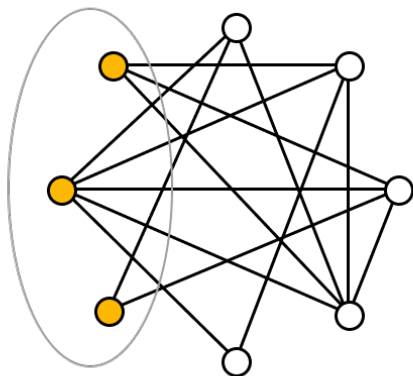
Independent Set Problem

Does a graph on n vertices have an independent set of size ρn ?



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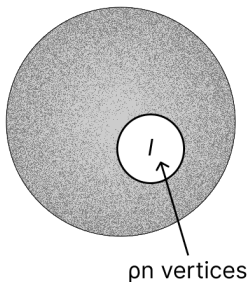
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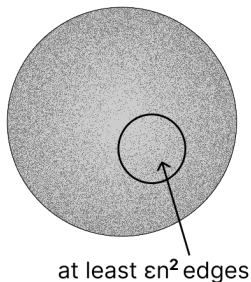
Testing Independent Sets

Testing Problem: Distinguish between the cases:

- (i) G has a ρn independent set, and
- (ii) every induced subgraph of size ρn in G has at least ϵn^2 edges.



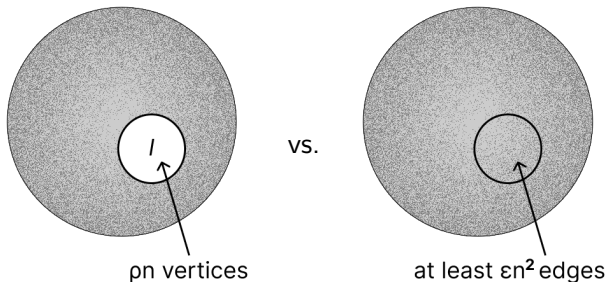
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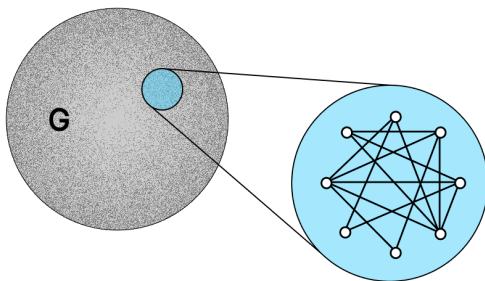
Theorem: Inspecting a random subgraph on $\tilde{O}(\rho/\epsilon^4)$ vertices is sufficient to distinguish between (i) and (ii).

[Goldreich, Goldwasser, Ron '98]

Definitions

An ϵ -tester for ρ -INDEPSET is an algorithm that samples a set S of s random vertices, examines the induced subgraph $G[S]$, and distinguishes between the cases (with high probability):

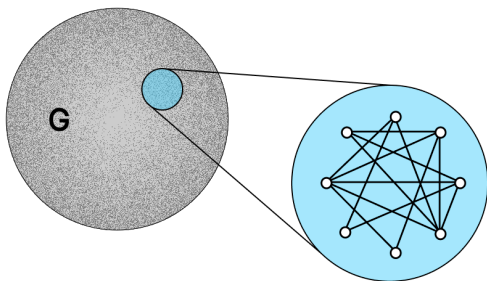
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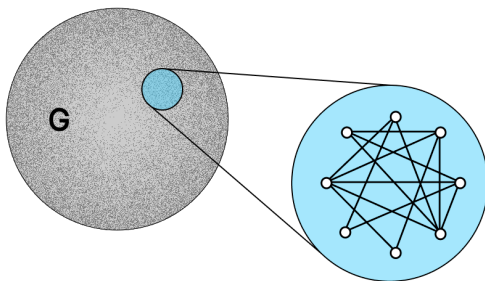


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Question: What is the minimum sample complexity $\mathcal{S}_{\rho\text{-INDEPSET}}(n, \epsilon)$ of ϵ -testing the ρ -INDEPSET property?

Results

Prior Results: The sample complexity $\mathcal{S}_{\rho\text{-INDEPSET}}(n, \epsilon)$ for ϵ -testing $\rho\text{-INDEPSET}$ is bounded above by:

- ▶ $\tilde{O}(\rho/\epsilon^4)$ [Goldreich, Goldwasser, Ron '98]
- ▶ $\tilde{O}(\rho^4/\epsilon^3)$ [Feige, Langberg, Schechtman '04]

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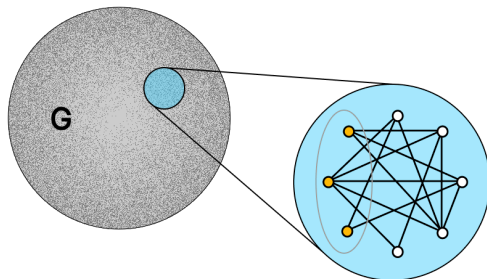
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Our Result: $\mathcal{S}_{\rho\text{-INDEPSET}}(n, \epsilon) = \tilde{\Theta}(\rho^3/\epsilon^2)$.

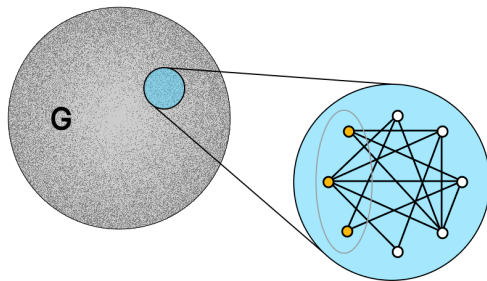
Proof Overview

Testing Algorithm: Take a random sample S of s vertices, check if the induced subgraph $G[S]$ has a ρs independent set.



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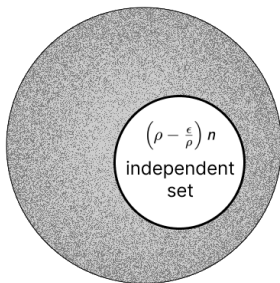
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Key Challenge: Show that if G is ϵ -far from having a ρn independent set then a random induced subgraph on $s \approx \rho^3/\epsilon^2$ vertices has a ρs independent set with only small probability.

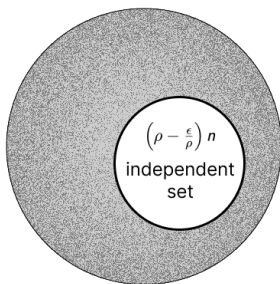
A Naive Approach

Observation: If G is ϵ -far from having a ρn independent set then the largest independent set in G is of size roughly $\left(\rho - \frac{\epsilon}{\rho}\right) n$.



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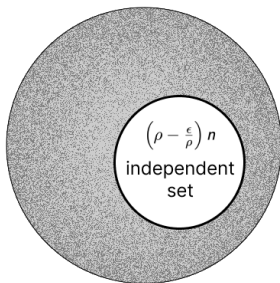


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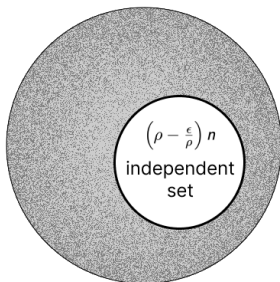


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- ▶ Upper bound the probability of getting ρs vertices in a sample of size s from a single $\left(\rho - \frac{\epsilon}{\rho}\right) n$ independent set.
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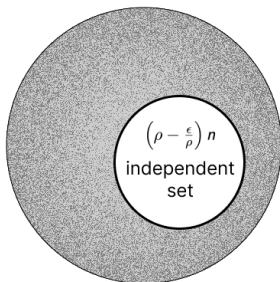


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Solution: The Graph Container Method

Graph Container Method

An Initial Graph Container Lemma

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Lemma. Let $G = (V, E)$ be ϵ -far from ρ -INDEPSET . Then, there exists a collection $\mathcal{C} \subseteq P(V)$ of containers that satisfies:

1. for every independent set I , there exists $C \in \mathcal{C}$ with $I \subseteq C$,
2. for every $C \in \mathcal{C}$, $|C| \lesssim \left(\rho - \frac{\epsilon}{\rho}\right) n$, and
3. $|\mathcal{C}| \lesssim \binom{n}{1/\epsilon}$.

[Kleitman, Winston '82]

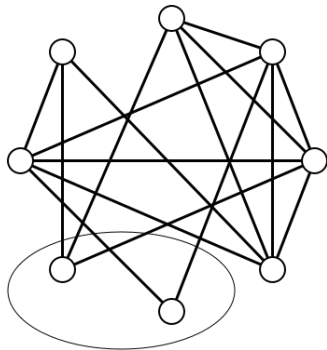
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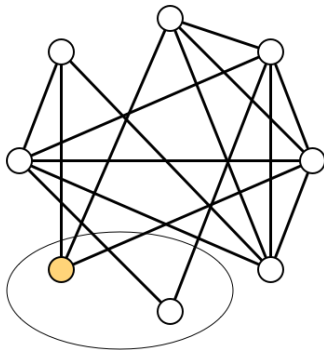
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Graph Container Method – Encoding View



How can I give you the most information about an independent set I by just telling you about **one** of the vertices in I ?

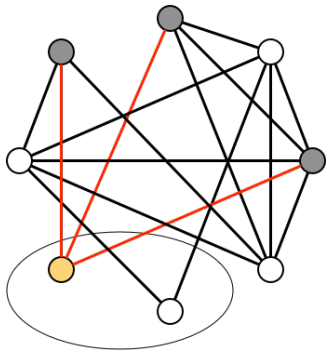
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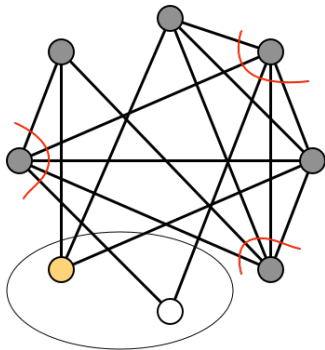
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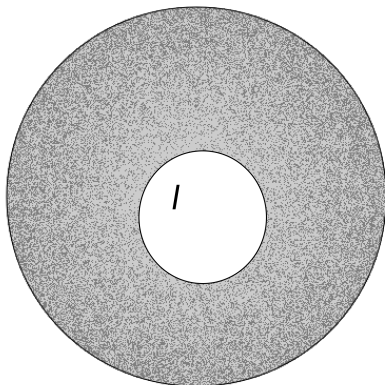
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Answer: Send the vertex $v \in I$ with highest degree.

Container Generation Algorithm

Input: Graph G and an independent set I

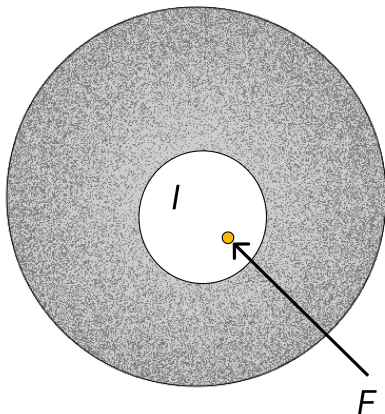
- ▶ Initialize fingerprint $F = \emptyset$ and container $C = V$
- ▶ Repeatedly select a vertex $v \in I$ with highest degree in $G[C]$ and add it to F . Remove $N(v)$ and all vertices with higher degree than v from C .



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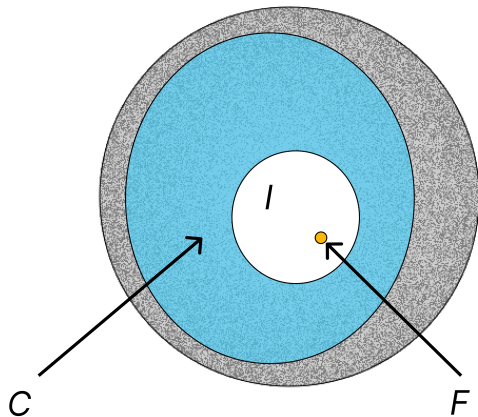


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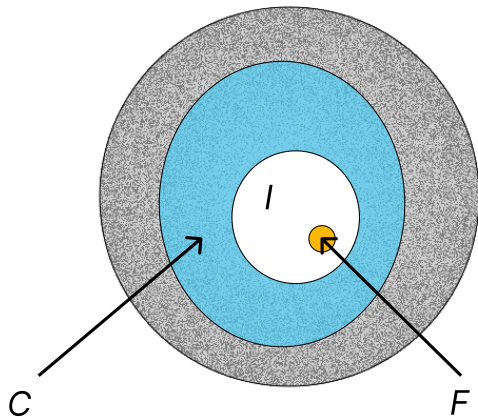


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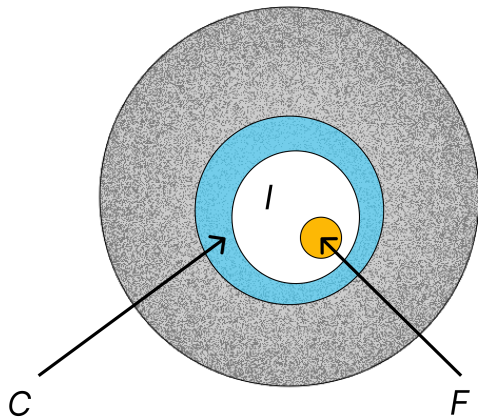


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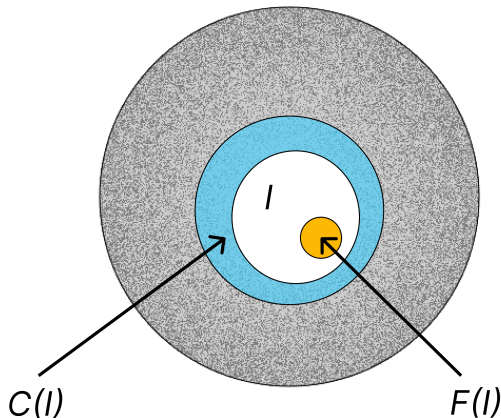


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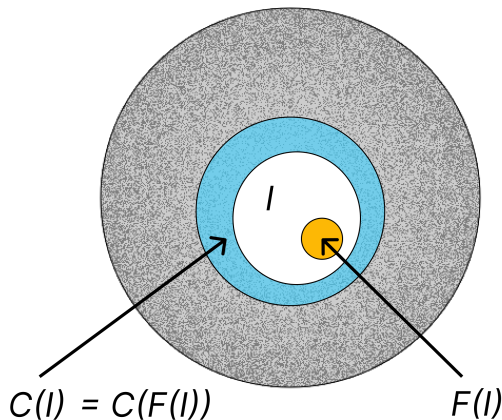


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Graph Container Lemma - Restated

Lemma. Let $G = (V, E)$ be ϵ -far from ρ -INDEPSET. Then, there exists a collection $\mathcal{C} \subseteq P(V)$ of containers that satisfies:

1. for every independent set I , there exists $C \in \mathcal{C}$ with $I \subseteq C$,
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Testing Independent Sets

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For example, at most $\binom{n}{O(1)}$ containers of size roughly $\left(\rho - \frac{\epsilon}{\rho}\right) n$.

Proof of the Main Theorem

Theorem. $\mathcal{S}_{\rho\text{-INDEPSET}}(n, \epsilon) = \tilde{O}(\rho^3/\epsilon^2)$.

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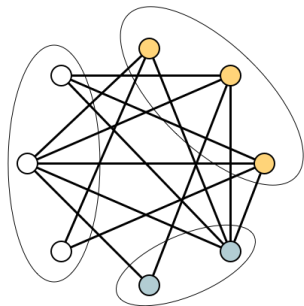
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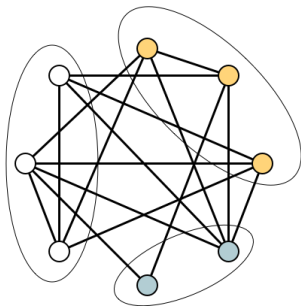
$$\begin{aligned} & \Pr[S \text{ contains } \rho s \text{ independent set}] \\ & \leq \sum_{F \in \mathcal{F}} \Pr[S \text{ contains a } \rho s \text{ independent set with fingerprint } F] \\ & \leq \sum_{F \in \mathcal{F}} \Pr[S \text{ contains } F \text{ and at least } \rho s \text{ vertices from } C(F)]. \end{aligned}$$

Testing Colorability

Testing k -Colorability



vs.



Goal: Distinguish between

- (i) G is k -colorable, and
- (ii) G is ϵ -far from being k -colorable.

Results: Testing k -Colorability

Prior Results. The sample complexity $\mathcal{S}_{k\text{-COLOR}}(n, \epsilon)$ for testing k -colorability is bounded above by:

- ▶ $O(k^2/\epsilon^3)$ [Goldreich, Goldwasser, Ron '98]
- ▶ $O(k/\epsilon^2)$ [Alon, Krivelevich '02]
- ▶ $O(k^4/\epsilon)$ [Sohler '12]

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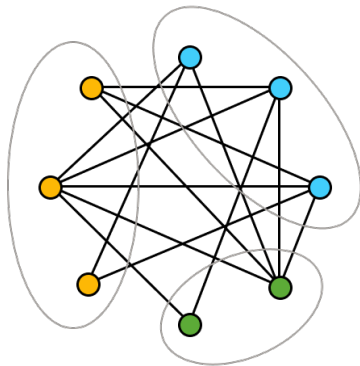
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Theorem. $\mathcal{S}_{k\text{-COLOR}}(n, \epsilon) = \tilde{O}(k/\epsilon)$.

Container Method for k -Colorability

Any k -colorable subgraph can be partitioned into k disjoint independent sets.



Run container procedure on each independent set, and take the union of the containers.

Summary: Property Testing and the Container Method

New graph container lemmas can be used to show:

▶ $\mathcal{S}_{\rho\text{-INDEPSET}}(n, \epsilon) = \tilde{\Theta}(\rho^3/\epsilon^2)$

▶ $\mathcal{S}_{k\text{-COLOR}}(n, \epsilon) = \tilde{O}(k/\epsilon)$

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- ▶ $\mathcal{S}_{k\text{-COLOR}}(n, \epsilon) = \tilde{O}(k/\epsilon)$
- ▶ New sample complexity bound of $\tilde{O}(kq^3/\epsilon)$ for satisfiability testing and new query complexity bound of $o((\rho^3/\epsilon^2)^2)$ for independent set testing.

[Blais, Seth, to appear at STOC '24]

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- ▶ $\mathcal{S}_{k\text{-COLOR}}(n, \epsilon) = \tilde{O}(k/\epsilon)$
- ▶ New sample complexity bound of $\tilde{O}(kq^3/\epsilon)$ for satisfiability testing and new query complexity bound of $o((\rho^3/\epsilon^2)^2)$ for independent set testing.

[Blais, Seth, to appear at STOC '24]

What else can the graph container method give for property testing/sublinear algorithms?

Summary: Property Testing and the Container Method

New graph container lemmas can be used to show:

- ▶ $\mathcal{S}_{\rho\text{-INDEPSET}}(n, \epsilon) = \tilde{\Theta}(\rho^3/\epsilon^2)$
- ▶ $\mathcal{S}_{k\text{-COLOR}}(n, \epsilon) = \tilde{O}(k/\epsilon)$
- ▶ New sample complexity bound of $\tilde{O}(kq^3/\epsilon)$ for satisfiability testing and new query complexity bound of $o((\rho^3/\epsilon^2)^2)$ for independent set testing.

[Blais, Seth, to appear at STOC '24]

What else can the graph container method give for property testing/sublinear algorithms?

Thank you!