On the Densest k-Subgraph problem

Aravindan Vijayaraghavan

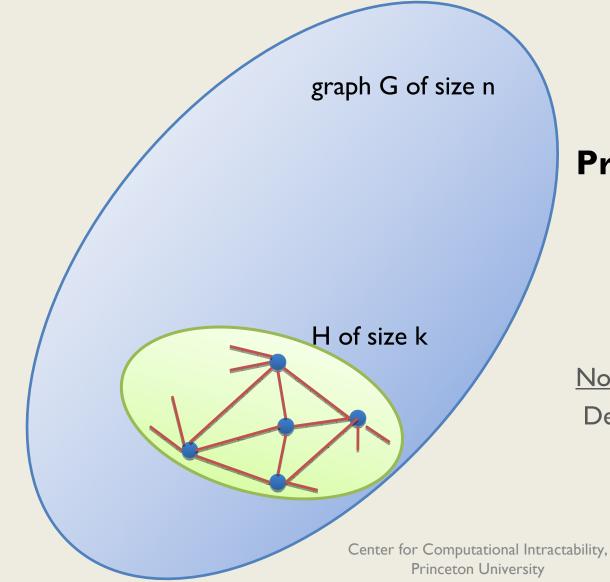
Princeton University & Center for Computational Intractability

Based on joint work

[Aditya Bhaskara, Moses Charikar, Eden Chlamtac, Uri Feige, V '10]

[Aditya Bhaskara, Moses Charikar, Venkat Guruswami, V, Yuan Zhou '11]

The Dense Subgraph Problem

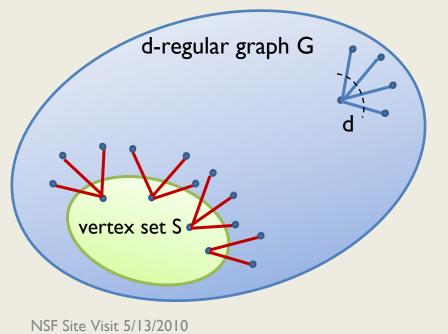


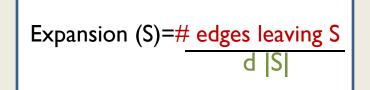
Problem. Given G, find a subgraph H of size k of max. density (think of k as n^p).

<u>Notation:</u> Density (H) = Avg. degree in induced subgraph H

Related problems

- Max–density subgraph (no size restriction): Polynomial time algorithm [GGT'87]
- Small set expansion





Center for Computational Intractability

Dense subgraphs are everywhere !

A useful subroutine for many applications

- <u>Social networks:</u> Trawling the Web for emerging cyber-communities [KRRT '99]
 - Web communities are characterized by dense bipartite subgraphs
- <u>Computational biology</u>: Mining dense subgraphs across massive biological networks for functional discovery [HYHHZ '05]
 - dense protein interaction subgraph corresponds to a protein complex [BD'03] [SM'03][SS '05]

Dense subgraphs are everywhere !

- A useful subroutine for many applications.
- A useful candidate hard problem with many consequences

Average case hardness assumption

- [ABW '10] Variant was used as the hardness assumption in Public Key Cryptography.
 Non-expanding small set – private key.
- [ABBG'10] Toxic assets can be hidden in complex financial derivatives to commit undetectable fraud
- [KZ'II,CMVZ'II] Evidence of inapproximability for many problems assuming hardness of planted variants.

How does DkS fit in?

Densest k-subgraph as a CSP with a strict budget:

DkS = (trivial) Max 2-AND at most k-variables set to 1

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Reeling in the years...

Problem. Given G, find a subgraph of size k with the maximum number of edges (think of k as n^{ρ})

Algorithms:

[FKP 93] give an $O(n^{1/3} - 1/90)$ approximation algorithm

Inapproximability:

[Feige 03] No PTAS under the Random 3-SAT assumption

[Khot 05] No PTAS unless NP \subseteq BPTIME(sub-exp)

[RS 10] No constant factor approx assuming Small Set Expansion Conjecture

[FS 97] Natural SDP has an $\Omega(n^{1/3})$ integrality gap

Algorithm

[Bhaskara, Charikar, Chlamtac, Feige, V'10]

Theorem. $O(n^{1/4 + \epsilon})$ approximation for DkS in time $O(n^{1/\epsilon})$

(Informal) Theorem. Can efficiently detect subgraphs of high log-density.

Strong Hierarchy Integrality gaps

[Bhaskara, Charikar, Guruswami, V, Zhou'II]

Theorem. $\Omega^{\sim}(n^{1/4})$ approximation for DkS for $\Omega(\log n/\log \log n)$ levels of SA+ (Sherali-Adams +SDP)

Theorem. $n^{\Omega(\epsilon)}$ gap for $n^{1-\epsilon}$ levels of Lasserre hierarchy

Outline

- Notion of log-density
- Algorithms for DkS:
 § Planted DkS: 'Local counting' based algorithms.
 § LP hierarchies to imitate arguments in worst case.
- Integrality gaps for strong hierarchies
- Open problems

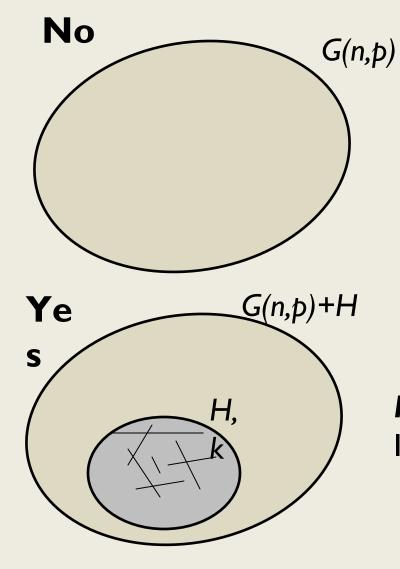
Log density

A graph on *n* vertices has **log-density** δ if the average degree is n^{δ}

$$\delta = \frac{\log d_{avg}}{\log |V|}$$

Question. Given G, can we detect the presence of a subgraph on k vertices, with higher log-density?

Planted versions of DkS



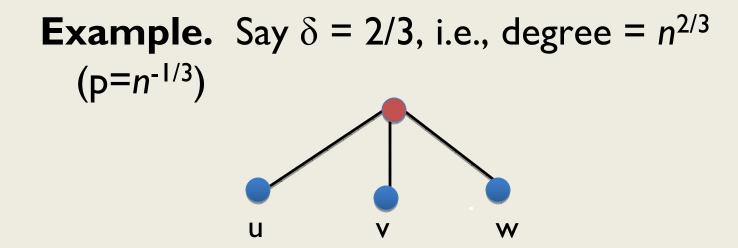
- Assume G does not have dense subgraphs
- Good algorithm for $DkS \Rightarrow$ we can distinguish

Problem. Distinguish between NO: G(n,p) of log-density δ YES: G(n,p) (same p) with ksubgraph of log-density $\delta + \epsilon$

Note:

In G(n,p), a k-subgraph H has density~ $kp = k (n^{\delta}/n) < k^{\delta}$

Main idea



random graph $G(n, n^{-1/3})$:

any three vertices have $O(\log n)$ common neighbors w.h.p. (n.p³ in expectation)

planted siste hiptezwith log conniton2/dighbors

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Main idea (contd.)

Example 2. $\delta = 1/3$, i.e., degree = $n^{1/3}$ ($p = n^{-2/3}$)

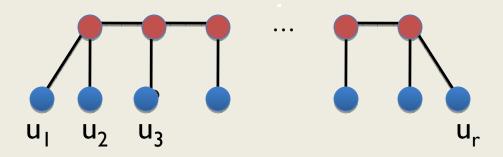


random graph $G(n, n^{-1/3})$:

any pair of vertices have $O(\log^2 n)$ paths of length 3, w.h.p. $(n^2p^3 \text{ in expectation})$ planted graph: size k, log-density 1/3+ ϵ : exists a pair of vertices with k^{ϵ} paths

Main idea (contd.)

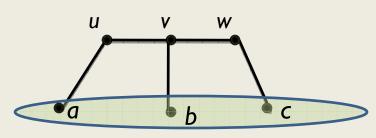
General strategy: For each rational δ , consider appropriate `caterpillar' structures, count how many `supported' on fixed set of leaves

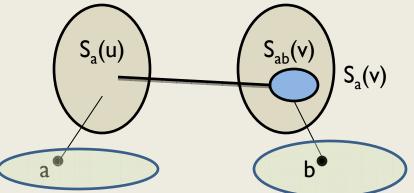


S Random graph G(n,p), log-density δ:
 every leaf tuple supports polylog(n) caterpillars
 S Planted graph, size k, log-density δ+ε :
 some leaf tuple supports at least k^ε caterpillars

Analysis for NO case ($\delta = 2/5$ i.e. $p=n^{-3/5}$)

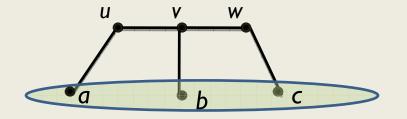
TO SHOW: Every leaf tuple supports polylog(n) caterpillars





Idea: Upper bound #candidates for each internal node by polylog(n). Fix tuple (a,b,c). Eg: $S_{ab}(v) \rightarrow candidates$ for v after fixing a,b. $\mathbf{E}[|S_a(u)|] \sim D = np = n^{2/5}$, and it is concentrated. Similarly, $\mathbf{E}[|S_a(v)|] \sim n^{4/5}$ and concentrated. $\mathbf{E}[|S_{ab}(v)|] \sim n^{4/5}p \sim n^{1/5}$ and it's concentrated. Similarly, $\mathbf{E}[|S_{abc}(w)|] \sim n^{1/5}$. np. p = O(1)

Proof for δ = 2/5



- # of "candidate w's" given leaves a,b,c is < log n w.h.p.
- The same is true for "candidate v's and u's" too by similar arguments.

Thus the number of structures is $< (\log^4 n)$ w.h.p.

Dense vs. Random – conclusion

Theorem. For every $\epsilon > 0$, and $0 < \delta < 1$, we can distinguish between G(n,p) of log-density δ , and a graph with a *k*-subgraph of log-density $\delta + \epsilon$, in time $n^{O(1/\epsilon)}$.

(Pick a rational no. in $[\delta, \delta + \epsilon)$ and use the appropriate caterpillar)

- k-subgraphs in G(n,p) have density max{1,kn^δ/n}
- Can detect planted k-subgraphs of density $k^{\delta+}$
- Distinguishing ratio ~ max $\frac{k^{\delta}}{\max\{1, kn^{\delta}/n\}} = O(n^{1/4})$

DkS in general graphs

Moving from average case to worst case

DkS in general graphs

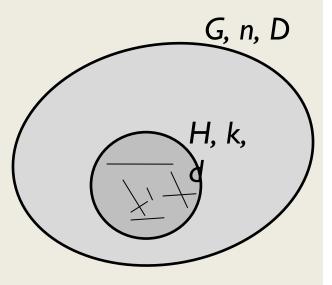
Input. G on *n* vertices, degree $\leq D$ **Promise.** There is a subgraph H on k vertices with average degree d **Question.** How dense a k-subgraph can we find?

An algorithm in worst case by mimicking our distinguishing algorithm for random graphs.

Some simplifications

Given: A regular graph G with degree
D= n^δ such that k.D=n
(k-subgraph in G has ~ O(1) density.)

H is k-subgraph of G with min-degree $d=k^{\delta+\epsilon}$ (higher log-density)



Aim: Enough to output a k-subgraph of density ρ (ρ is a large constant)

Observation: Can return a ρ -dense subgraph with \leq k vertices (remove, repeat)

An outline of the algorithm

- G, n, D=n^{δ} H, k, d
- Inspired by algorithm for Planted problem.
- Algorithm for each δ uses the structure Cat $_{\delta}$ (size s $_{\delta}$)

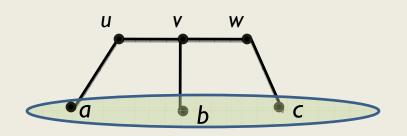
Algorithm proceeds for s_{δ} steps.

Idea. Look at the 'set of candidates' for a non-leaf after *fixing* a prefix of the leaves

 S_t -- candidate vertices at step t of the caterpillar. LP(S) -- the number of vertices from H in S. Algorithm either finds dense-subgraph from S_t or lt 'behaves' as in random case and lower bound LP(S_{t+1})/ $|S_{t+1}|$

Finally, $LP(S_t)/|S_t| > 1$ (contradiction)

Algorithm using Cat_{δ} (plot outline)



<u>Procedure LocalSearch(S)</u> Tries to find a dense subgraph greedily between **S** and $\Gamma(S)$

- I. $S_0 = V$. Perform LocalSearch(S_0)
- 2. If we don't get a dense subgraph, then $\exists a$ s.t. $|S_a(u)| \le U_1$ (as in random graph) and $|LP(S_a(u))| \ge L_1$.
- 3. Do LocalSearch($S_a(u)$). If it fails then $|S_a(v)| \le U_2$ and $|LP(S_a(v))| \ge L_2$
- 4. Do LocalSearch(S_a(v)). If fail, ∃ b s.t bounds like random Keep doing this ... At the last step, the parameters give a contradiction!

LP relaxation (a hierarchy) for Cat_{δ}

Intended solution: k-subgraph H with minimum degree d Simple LP:

$$\sum_{i \in V} y_i \leq k \text{ and}$$
(1) (size at most k)

$$\exists y_{ij} : i, j \in V \text{ s.t.}$$
(2) (min degree d in H)

$$\forall i, j \in V \quad \sum_{j \in \Gamma(i)} y_{ij} \geq dy_i$$
(3)

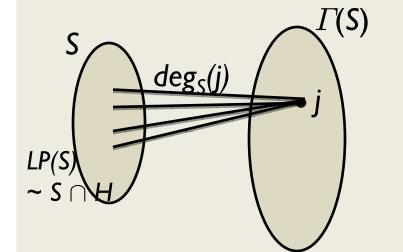
$$\forall i, j \in V \quad 0 \leq y_{ij} \leq y_i \leq 1$$
(4)

LP : Simple LP + LS hierarchy for s_{δ} levels.

• Captures fixing leaves since $\{y_{ij} / y_j\}$ satisfy LP too.

• LP is feasible for any constant number of conditionings (i.e. fixing leaves).

Main Component – LocalSearch(S)



Consider $k'=LP(\Gamma(S))$ (<= k)

$$Edges(S, S_{k'}) \ge \sum y_{j} deg_{S}(j)$$

$$j\epsilon\Gamma(S)$$

$$\ge \sum \sum y_{ij} \ge dLP(S)$$

$$i\epsilon \sum j\epsilon\Gamma(i) \qquad (due \text{ to eq } 2)$$

Greedy algorithm:

For each $k' = 1 \dots k$, do:

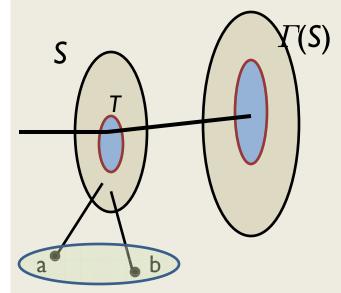
- $S_{k'} = k'$ vertices in $\Gamma(S)$ with the most edges to S.
- Let S* be k vertices from S with most edges into S_k.

If S_k, U S* has density $\geq \rho$, return it. If no ρ dense subgraph is found, return Fail

Lem. LocalSearch finds a graph of density at least $= \frac{d LP(S)}{LP(\Gamma(S))+|S|}$

Round or Bound -1 (backbone edge)

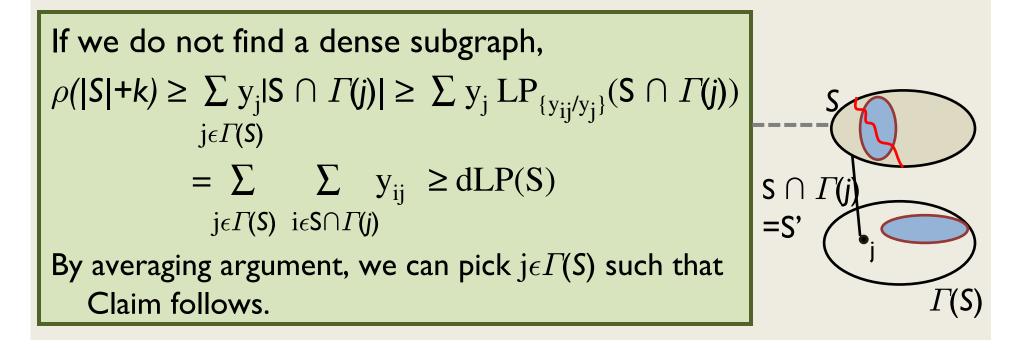
Claim1: Let S be candidates, $\{y_i\}$ be LP solution, we either a) Output a k-subgraph of density ρ using LocalSearch b) else $LP(\Gamma(S)) \ge d LP(S)/\rho$ (we can set $S_{new} = \Gamma(S)$)



If we do not find ρ dense subgraph, $S_{new} = \Gamma(S)$ LP($\Gamma(S)$) increases by at least d/ ρ and $|\Gamma(S)|$ increases by at most D (like in the random case)

Round or Bound – 2 (leaf/hair)

Claim 2: If S is candidate set, $\{y_i\}$ is LP solution, we either a) Find a k-subgraph of density ρ between S and $\Gamma(S)$ b) or find leaf j if $S_{new} = S \cap \Gamma(j)$ $LP(S_{new}) \ge d LP(S)/2k$ and $|S_{new}| \le \rho(|S|+k)/k$



To summarize...

Roughly speaking, if we don't find a dense subgraph in a step,

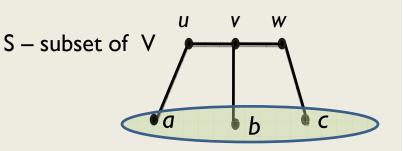
- every backbone step, LP(S)/|S| decreases by O(d/D)
- every hair step, LP(S)/|S| increases by at least $\Omega(d)$

Because of choice of structure, LP(S)/|S| becomes >1 at final step (a contradiction).

Completing the algorithm for $\delta = 2/5$

I.
$$S_0 = V$$
. $LP(S_0) / |S_0| = k/n$.

2. If LocalSearch(S₀) doesn't give a 100-dense subgraph, $\exists a$ to condition on so that, LP(S_a(u))/|S_a(u)| \geq dk/n



- 3. If LocalSearch($S_a(u)$) fails, LP($S_a(v)$)/ $|S_a(v)| \ge d^2k$ /Dn
- 4. If LocalSearch($S_a(v)$) fails, $\exists b LP(S_{ab}(v))/|S_{ab}(v)| \ge d^3k/Dn$.
- 5. If LocalSearch($S_{ab}(v)$) fails, LP($S_{ab}(w)$)/ $|S_{ab}(w)| \ge d^4k/D^2n$
- 6. If LocalSearch($S_{ab}(w)$) fails, LP($S_{abc}(w)$)/ $|S_{abc}(w)| \ge d^{5}k^{3}/n^{3} > I$

(a contradiction)

Beating the log-density barrier?

•
$$n^{(1-\varepsilon)/4}$$
 approximation in time $2^{n^{6\varepsilon}}$

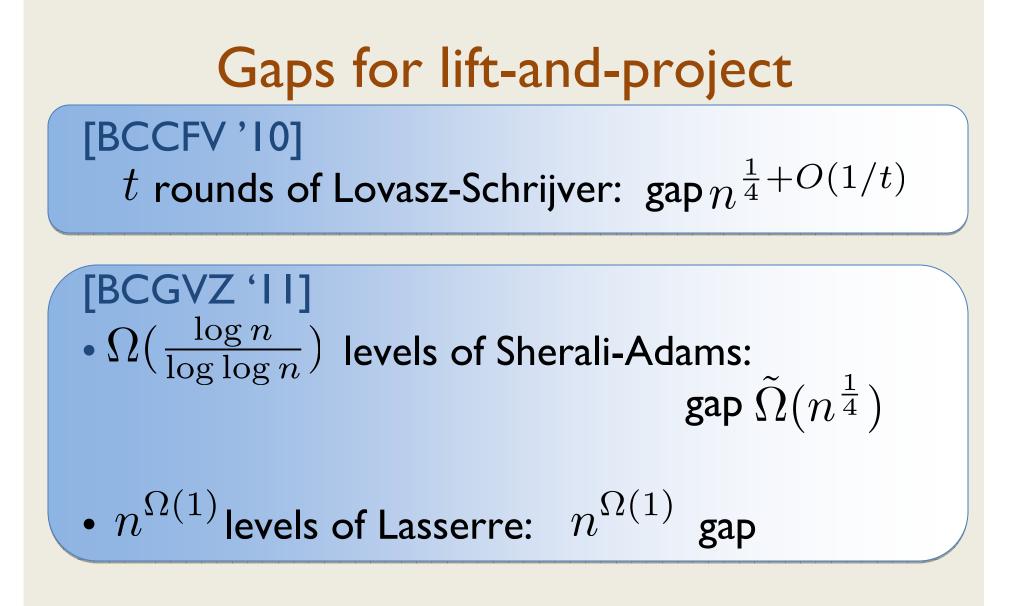
- Guess subsets of size n^{ε} for every leaf in caterpillar structure.
- Integrality gaps suggest polytime algorithms from Sherali-Adams (SA+) relaxations can not beat the barrier.

Stronger relaxations



Sherali-Adams

Lovasz-Schrijver



Lasserre gaps

- First constructs gaps for Max r-CSP(q) instances over large alphabet size r,q= n^{Ω(1)}.
- Simple reduction from Max r-CSP(q) to DkS
- Uses Tulsiani's framework to transform the Lasserre gaps for DkS.

Small Set Expansion (SSE) problem [RS '10]

Given $\varepsilon, \delta > 0$, D-regular graph G, distinguish between (think of D as constant)

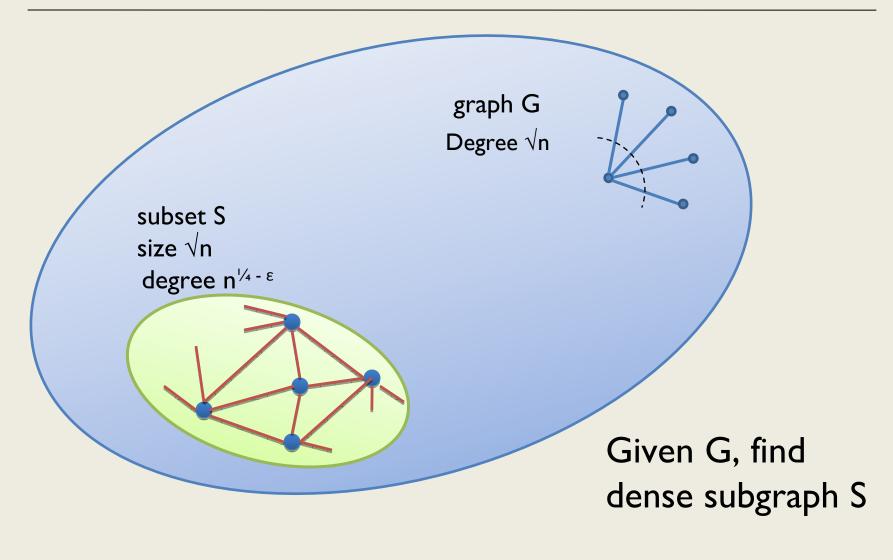
- [NO] Small subgraphs expand very well Every subgraph of size ≤δn has density ≤ εD
- [YES] Some small subgraph does not expand
 Some subgraph of size ≤δn has density ≥ (1-ε)D

[ABS'10] exp(n^{O(poly(ε))}) time algorithm for SSE. [BRS'11,GS'11] distinguish using n^{O(poly(ε))} levels of Lasserre.

 $n^{\Omega(1)}$ levels Lasserre gap for DkS seems to suggest that DkS much harder than SSE

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Open Problem



Open Problems

- Better algorithms using SDPs in certain ranges of parameters? (like [Steurer'11])
- Evidence of large inapproximability of DkS?
- Stronger integrality gaps?
 Maybe n^{1/4-ε} gap for n^ε levels of the hierarchy?

Thank you!