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## Denser than the Densest Subgraph: Extracting Optimal Quasi-Cliques with Quality Guarantees

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## Denser than the densest

- Densest subgraph problem is very popular in practice. However, not what we want for many applications.
- $\delta=$ edge density, $D=$ diameter, $\tau=$ triangle density
densest subgraph optimal quasi-clique

|  | $\frac{\|S\|}{\|V\|}$ | $\delta$ | $D$ | $\tau$ | $\frac{\|S\|}{\|V\|}$ | $\delta$ | $D$ | $\tau$ |
| ---: | :--- | :--- | :--- | :--- | :---: | :---: | :---: | :---: |
|  | 0.32 | 0.33 | 3 | 0.04 | 0.12 | 0.68 | 2 | 0.32 |
| Dolphins | 0.09 | 4 | 0.03 | 0.10 | 0.73 | 2 | 0.34 |  |
| Football | 1 | 0.34 | 3 | 0.08 | 0.15 | 1 | 1 | 1 |
| Jazz | 0.50 | 0.34 |  |  |  |  |  |  |
| Celeg. N. | 0.46 | 0.13 | 3 | 0.05 | 0.07 | 0.61 | 2 | 0.26 |
|  |  |  |  |  |  |  |  |  |

## Graph mining applications

- Thematic communities and spam link farms [Gibson, Kumar, Tomkins '05]
- Graph visualization[Alvarez-Hamelin etal.'05]
- Real time story identification [Angel et al. '12]
- Motif detection [Batzoglou Lab ‘o6]
- Epilepsy prediction [lasemidis et al. 'o1]
- Finding correlated genes [Horvath et al.]
- Many more ..


## Measures

- Clique: each vertex in S connects to every other vertex in S . $\alpha$-Quasi-clique: the set $S$ has at least $\alpha|S|(|S|-1) / 2$ edges.
- $k$-core: every vertex connects to at least $k$ other vertices in S .


## Measures

$\delta(\mathrm{S})=\frac{e[S]}{\binom{|S|}{2}}$
Density
$\mathrm{d}(\mathrm{S})=\frac{2 e[S]}{|S|}$
Average degree
$\mathrm{t}(\mathrm{S})=\frac{t[S]}{\binom{|S|}{3}}$
Triangle Density

## Contributions

- General framework which subsumes popular density functions.
- Optimal quasi-cliques.
- An algorithm with additive error guarantees and a local-search heuristic.
- Variants
- Top-k optimal quasi-cliques
- Successful team formation


## Contributions

- Experimental evaluation
- Synthetic graphs
- Real graphs
- Applications
- Successful team formation of computer scientists
- Highly-correlated genes from microarray datasets

First, some related work.

## Cliques

- Maximum clique problem: find clique of maximum possible size. NP-complete problem
- Unless P=NP, there cannot be a
 polynomial time algorithm that approximates the maximum clique problem within a factor better than $O\left(n^{1-\varepsilon}\right)$ for any $\varepsilon>0$ [Håstad 'gg].


## (Some) Density Functions

$\delta(\mathrm{S})=\frac{e[S]}{(S \mid} \quad$ A single edge achieves $\binom{(S)}{2}$ always maximum possible $\delta(\mathrm{S})$
$\mathrm{d}(\mathrm{S})=\frac{e[S]}{|S|} \quad$ Densest subgraph problem
$\mathrm{d}(\mathrm{S})=\frac{e[S]}{|S|},|\mathrm{S}|=\mathrm{k} \quad$ k-Densest subgraph problem
$\mathrm{d}(\mathrm{S})=\frac{e[S]}{|S|},|\mathrm{S}| \geq \mathrm{k}(|\mathrm{S}| \leq \mathrm{k}) \quad$ DalkS (Damks)

## Densest Subgraph Problem

- Maximize average degree
- Solvable in polynomial time
- Max flows (Goldberg)
- LP relaxation (Charikar)
- Fast $1 / 2$-approximation algorithm (Charikar)


## k-Densest subgraph

- k-densest subgraph problem is NP-hard
- Feige, Kortsatz, Peleg

Bhaskara, Charikar, Chlamtac, Vijayraghavan Asahiro et al.
Andersen
Khuller, Saha [approximation algorithms], Khot [no PTAS].

## Quasicliques

- A set $S$ of vertices is $\alpha$-quasiclique if

$$
e[S] \geq \alpha\binom{|S|}{2}
$$

[Uno '10] introduces an algorithm to enumerate all $\alpha$-quasicliques.

## Edge-Surplus Framework

- For a set of vertices $S$ define

$$
f_{\alpha}(S)=g(e[S])-a h(|S|)
$$

where $\mathrm{g}, \mathrm{h}$ are both strictly increasing, $\alpha>0$.

- Optimal ( $\alpha, g, \mathrm{~h})$-edge-surplus problem Find S* such that $f_{\alpha}\left(S^{*}\right) \geq f_{\alpha}(S)$.


## Edge-Surplus Framework

- When $g(x)=h(x)=\log (x), \alpha=1$, then

Optimal ( $\alpha, g, \mathrm{~h}$ )-edge-surplus problem
becomes max $\log \frac{e[S]}{|S|}$, which is the densest subgraph problem.
$g(x)=x, h(x)=0$ if $x=k$, o/w $+\infty$ we get the $k$ densest subgraph problem.

## Edge-Surplus Framework

- When $g(x)=x, h(x)=x(x-1) / 2$ then we obtain $\max _{S \subseteq V,|S| \geq 2} e[S]-\alpha\binom{|S|}{2}$, which we define as the optimal quasiclique (OOC) problem.
- Theorem 1: Let $g(x)=x, h(x)$ concave. Then the optimal ( $\alpha, g, h$ )-edge-surplus problem is polytime solvable.
- However, this family is not well suited for applications as it returns most of the graph.


## Hardness of OOC

- Conjecture: finding a planted clique $C$ of size $n^{\frac{1}{2}-\delta}, \delta>0$ in a random binomial graph $G\left(n, \frac{1}{2}\right)$ is hard.
- Let $\mathrm{f}(\mathrm{S})=\mathrm{e}[\mathrm{S}]-\frac{2}{3}\binom{|S|}{2}$. Then,

$$
\begin{aligned}
& \mathrm{f}(\mathrm{C})=\frac{1}{3}\binom{n^{\frac{1}{2}-\delta}}{2}>0 \\
& E[f(S)]=-\frac{1}{6}\binom{|S|}{2}<0 .
\end{aligned}
$$

## Multiplicative approximation algorithms

- Notice that in general the optimal value can be negative.
- We can obtain guarantees for a shifted objective but introduces large additive error making the algorithm almost useless, i.e., except for very special graphs.
- Other type of guarantees more suitable.


## Optimal Quasicliques

- Additive error approximation algorithm
${ }^{-} G_{n} \leftarrow G$
- For $k \leftarrow n \quad$ downto 1
- Let v be the smallest degree vertex in $G_{k}$.
- $G_{k-1} \leftarrow G_{k}-\{v\}$
- Output $\bar{S} \leftarrow \operatorname{argmax} x_{1 \leq k \leq n} f_{a}\left(G_{k}\right)$

Theorem: $\mathrm{f}_{\alpha}(\bar{S}) \geq f_{\alpha}\left(S^{*}\right)-\frac{\alpha}{2} "$ small" $\times|\bar{S}|$
Running time: $\mathrm{O}(\mathrm{n}+\mathrm{m})$. However it would be nice to have running time $\mathrm{O}(\mid$ output $\mid)$.

## Optimal Quasicliques Local Search Heuristic

1. Initialize $S$ with a random vertex.
2. For $t=1$ to $T_{\max }$
3. Keep expanding $S$ by adding at each time a vertex $v \notin S$ such that $f_{\alpha}(S \cup v) \geq f_{\alpha}(S)$.
4. If not possible see whether there exist $v \in S$ such that $f_{\alpha}(S-\{v\}) \geq f_{\alpha}(S)$.
5. If yes, remove it. Go back to previous step.
6. If not, stop and output $S$.

## Experiments

|  | Vertices | Edges | Description |
| :---: | :---: | :---: | :---: |
| Dolphins | 62 | 159 | Biological Network |
| Polbooks | 105 | 441 | Books Network |
| Adjnoun | 112 | 425 | Adj. and Nouns in 'David Copperfield' |
| Football | 115 | 613 | Games Network |
| Jazz | 198 | 2742 | Musicians Network |
| Celegans N . | 297 | 2148 | Biological Network |
| Celegans M. | 453 | 2025 | Biological Network |
| Email | 133 | 5451 | Email Network |
| AS-22july06 | 22963 | 48436 | Auton. Systems |
| Web-Google | 875713 | 3852985 | Web Graph |
| Youtube | 1157822 | 2990442 | Social Network |
| AS-Skitter | 1696415 | 11095298 | Auton. Systems |
| Wikipedia 2005 | 1634989 | 18540589 | Web Graph |
| Wikipedia 2006/9 | 2983494 | 35048115 | Web Graph |
| Wikipedia 2006/11 | 3148440 | 37043456 | Web Graph |

## Experiments

|  | $\|S\|$ |  |  | $\delta$ |  |  | D |  |  | $\tau$ |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | densest | opt. quas | lique | densest | opt. qua | -clique | $\begin{gathered} \text { densest } \\ \text { subgraph } \end{gathered}$ | opt. quasi-clique |  | densest subgraph | opt. quasi-clique |  |
|  | subgraph | GREEDY | LS | subgraph | GREEDY | LS |  | GREEDY | LS |  | GREEDY | LS |
| Dolphins | 19 | 13 | 8 | 0.27 | 0.47 | 0.68 | 3 | 3 | 2 | 0.05 | 0.12 | 0.32 |
| Polbooks | 53 | 13 | 16 | 0.18 | 0.67 | 0.61 | 6 | 2 | 2 | 0.02 | 0.28 | 0.24 |
| Adjnoun | 45 | 16 | 15 | 0.20 | 0.48 | 0.60 | 3 | 3 | 2 | 0.01 | 0.10 | 0.12 |
| Football | 115 | 10 | 12 | 0.09 | 0.89 | 0.73 | 4 | 2 | 2 | 0.03 | 0.67 | 0.34 |
| Jazz | 99 | 59 | 30 | 0.35 | 0.54 | 1 | 3 | 2 | 1 | 0.08 | 0.23 | 1 |
| Celeg. N . | 126 | 27 | 21 | 0.14 | 0.55 | 0.61 | 3 | 2 | 2 | 0.07 | 0.20 | 0.26 |
| Celeg. M. | 44 | 22 | 17 | 0.35 | 0.61 | 0.67 | 3 | 2 | 2 | 0.07 | 0.26 | 0.33 |
| Email | 289 | 12 | 8 | 0.05 | 1 | 0.71 | 4 | 1 | 2 | 0.01 | 1 | 0.30 |
| AS-22july06 | 204 | 73 | 12 | 0.40 | 0.53 | 0.58 | 3 | 2 | 2 | 0.09 | 0.19 | 0.20 |
| Web-Google | 230 | 46 | 20 | 0.22 | 1 | 0.98 | 3 | 2 | 2 | 0.03 | 0.99 | 0.95 |
| Youtube | 1874 | 124 | 119 | 0.05 | 0.46 | 0.49 | 4 | 2 | 2 | 0.02 | 0.12 | 0.14 |
| AS-Skitter | 433 | 319 | 96 | 0.41 | 0.53 | 0.49 | 2 | 2 | 2 | 0.10 | 0.19 | 0.13 |
| Wiki '05 | 24555 | 451 | 321 | 0.26 | 0.43 | 0.48 | 3 | 3 | 2 | 0.02 | 0.06 | 0.10 |
| Wiki '06/9 | 1594 | 526 | 376 | 0.17 | 0.43 | 0.49 | 3 | 3 | 2 | 0.10 | 0.06 | 0.11 |
| Wiki '06/11 | 1638 | 527 | 46 | 0.17 | 0.43 | 0.56 | 3 | 3 | 2 | 0.31 | 0.06 | 0.35 |

\begin{tabular}{|l|l|l|l|l|l|l|l|l|l|l|l|l|}
\hline \& DS \& $M_{1}$ \& $M_{2}$ \& DS \& $M_{1}$ \& $M_{2}$ \& DS \& $M_{1}$ \& $M_{2}$ \& DS \& $M_{1}$ \& $M_{2}$ <br>

\hline Wiki `05 \& | 24.5 |
| :--- |
| K | \& 451 \& 321 \& .26 \& .43 \& .48 \& 3 \& 3 \& 2 \& .02 \& .06 \& .11 <br>

\hline
\end{tabular}

| Youtube | 1.9 K | 124 | 119 | 0.05 | 0.46 | 0.49 | 4 | 2 | 2 | .02 | .12 | .14 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |

## Top-k densest subgraphs

Wikipedia 2006/11


Wikipedia 2006/11


## Constrained Optimal Quasicliques

- Given a set of vertices O
$\max _{\mathrm{Q} \subseteq S \subseteq V} f_{\alpha}(\mathrm{S})=\max _{\mathrm{Q} \subseteq S \subseteq V} e[S]-\alpha\binom{|S|}{2}$
- Lemma: NP-hard problem.
- Observation: Easy to adapt our efficient algorithms to this setting.
- Local Search: Initialize S with O and never remove a vertex if it belongs to O
- Greedy: Never peel off a vertex from Q


## Application 1

- Suppose that a set Q of scientists wants to organize a workshop. How do they invite other scientists to participate in the workshop so that the set of all participants, including Q , have similar interests ?


## Query 1, <br> Papadimitriou and Abiteboul



## Query 2,

## Papadimitriou and Blum



## Application 2

- Given a microarray dataset and a set of genes Q , find a set of genes S that includes Q and they are all highly correlated.
- Co-expression network
- Measure gene expression across multiple samples
- Create correlation matrix
" Edges between genes if their correlation is > $\rho$.
- A dense subgraph in a co-expression network corresponds to a set of highly correlated genes.


## Query, p53



ACTIVATION OF P53


## Future Work

- Hardness
- Analysis of local search algorithm
- Other algorithms with additive approximation guarantees
- Study the natural family of objectives

$$
\max _{\subset \backslash|S|>2} e[S]-\alpha|S|^{\gamma}, \gamma>1
$$

## Thank you!

