



## Dense Subgraph Discovery Meets Strong Triadic Closure

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## Dense subgraph discovery

- Finding dense subgraphs has many applications in diverse domains such as community detection, biological system analysis, and anomaly detection.
- Definition of **density** :

$$d(S) = \frac{|E(S)|}{|V(S)|}$$

- **Densest subgraph problem** : Find a subset of vertices which maximizes the density.
- Exact<sup>1</sup> and greedy<sup>2</sup> algorithms.

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<sup>1</sup> Andrew V Goldberg. Finding a maximum density subgraph. 1984.

<sup>2</sup> Moses Charikar. Greedy approximation algorithms for finding dense components in a graph. In International workshop on approximation algorithms for combinatorial optimization, pages 84–95. Springer, 2000.

## STC-constrained dense subgraph discovery problem

### Input:

- A graph.
- Weight parameter :  $\lambda$ .

### Output:

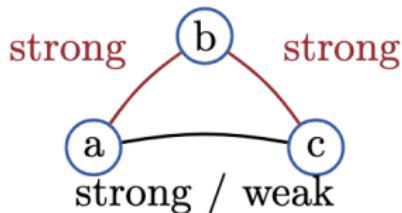
- A subgraph.
- A labeling of each edge as either **strong** or **weak** within the subgraph.

### Such that:

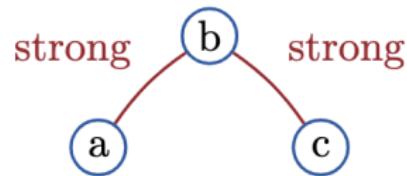
- Our density-like score  $\frac{|S| + \lambda |W|}{|U|}$  is maximized.
- Strong edges should obey strong triadic closure (which we define next).

## Strong triadic closure (STC)

- If  $(a, b)$  and  $(b, c)$  are **strongly** connected, there should be at least a **weak** connection between  $a$  and  $c$ .



✓



✗

## Our problem : STC-DEN

- Given a graph  $G = (V, E)$  and a weight parameter  $\lambda$ ,  
find a subset of vertices  $U \subseteq V$  and  
split  $E(U)$  into two sets: set of **strong** edges  $S$  and **weak** edges  $W$   
such that the STC property is satisfied in  $(U, E(U))$  and  
$$\frac{|S| + \lambda |W|}{|U|}$$
 is maximized.
- Our problem is **NP-hard**.
- The case of  $\lambda = 0$  reduces to the **MAX-CLIQUE** problem.
- The case of  $\lambda = 1$  reduces to the densest subgraph problem which is polynomial time solvable.

## Auxiliary problem : STC-DEN( $\alpha$ )

- Given a graph  $G = (V, E)$ , a weight parameter  $\lambda$ , and a number  $\alpha$ , find a subset of vertices  $U \subseteq V$  and split  $E(U)$  into two sets: set of **strong** edges  $S$  and **weak** edges  $W$  such that the STC property is satisfied in  $(U, E(U))$  and  $|S| + \lambda|W| - \alpha|U|$  is maximized.

## ILP formulation : STC-DEN( $\alpha$ )

$$\begin{array}{ll}
 \text{MAXIMIZE} & \sum x_{ij} + \lambda \sum z_{ij} - \alpha \sum y_i \\
 \text{SUBJECT TO} & \\
 & x_{ij} + z_{ij} \leq y_i \quad ij \in E \\
 & x_{ij} + z_{ij} \leq y_j \quad ij \in E \\
 & x_{ij} + x_{jk} \leq y_j \quad (i, j, k) \in Z \\
 & x_{ij}, z_{ij} \in \{0, 1\} \quad ij \in E \\
 & y_i \in \{0, 1\} \quad i \in V
 \end{array}$$

- $\alpha$  : Guess parameter.
- $y_i$  : Whether the node  $i \in U$ .
- $x_{ij}$  : Whether the edge  $(i, j)$  is strong or not.
- $z_{ij}$  : Whether the edge  $(i, j)$  is weak or not.
  
- STC-DEN( $\alpha$ ) yields the same solution as STC-DEN for the largest  $\alpha$  with a non-empty solution.

## LP relaxation

- Relax the integer program formulation.
- Construct a collection of sets  $\mathcal{S} = \{S_1, S_2, \dots, S_n\}$  where  $S_j = \{i : y_i^* \geq y_j^*\}$ .
- Order the edges based on  $z_e^*$  and label edges as strong if the STC property is not violated.
- Out of all subgraphs, pick the subgraph and labeling that maximizes our score.

## Algorithms based on label, find the densest subgraph, and relabel

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**Algorithm 1:** STC-CUT( $G, \lambda$ ) and STC-PEEL( $G, \lambda$ ).

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- 1  $L \leftarrow$  find an approximate labeling [3];
  - 2  $H \leftarrow$  the weighted graph by setting 1 to strong edges  
and  $\lambda$  to weak edges of  $G$ ;
  - 3  $U \leftarrow$  the densest subgraph (exact [1] or approximate [2]);
  - 4  $L' \leftarrow$  find an approximate labeling [3];
  - 5 **return** subgraph  $U$  and its labeling  $L'$ ;
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<sup>3</sup> Sintos, S., Tsaparas, P. (2014, August). Using strong triadic closure to characterize ties in social networks. In Proceedings of the 20th ACM SIGKDD international conference on Knowledge discovery and data mining (pp. 1466-1475).

## Greedy algorithm

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**Algorithm 2:** Greedy algorithm.

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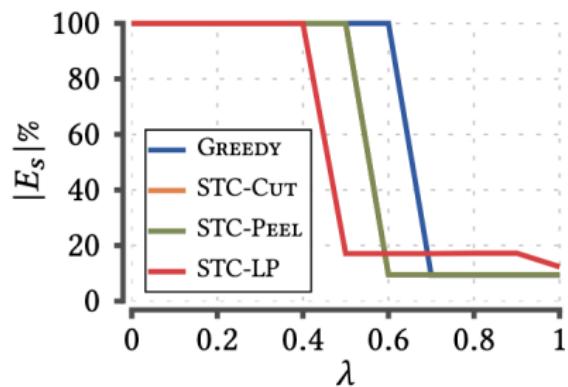
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1  $U \leftarrow V;$ 
2 while there are nodes do
3    $L \leftarrow$  find an approximate labeling [1] ;
4    $u \leftarrow \arg \min_{v \in U} \deg_\lambda(v, U, L, \lambda);$ 
5    $U \leftarrow U \setminus \{u\};$ 
6 return best tested  $U$  and its labeling  $L;$ 
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## Experiments with Synthetic Datasets

- | $E_s$ |% : The percentages of *strong* edges in discovered subgraph.



## Experiments with Real-world Datasets

- STC-ILP always give the highest score while all other algorithms produce approximately equal scores.
- When STC-ILP is not usable, STC-LP produce the maximum score in general.
- For high  $\lambda$ s all of them produce less deviated scores when compared to lower  $\lambda$ s.
- In general,  $|E_s|%$  monotonically decreases as  $\lambda$  increases.
- Greedy, STC-LP, and STC-CUT run significantly slower than STC-PEEL.

## Conclusion

- Introduced a novel dense subgraph discovery problem that takes into account the strength of ties within the subgraph.
- Our goal was to maximize a density-like measure defined as the sum of the number of strong edges and the number of weak edges weighted by a weight parameter, divided by the number of nodes within the subgraph.
- Proved that our problem is **NP-hard**.
- Designed an exact algorithm based on integer linear programming.
- Designed linear programming, a greedy heuristic algorithm, and two other straightforward algorithms based on the algorithms for the densest subgraph discovery which run in polynomial time.
- Showed experimentally that the algorithms could find the ground truth using synthetic dataset.
- Showed experimentally that our proposed algorithms discovered the subgraphs reasonably fast in practice.

**Thank you for your attention!!!**