## On Breaking k-Trusses

## Dutch Seminar on Optimization

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## Data Sanitization



Analysis

## Data Sanitization

- Privacy Constraints
- Utility Properties


CWI

## Communities



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## CWI

## Community Breaking

- Maintaining communities in social networks


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- Maintaining communities in social networks
- Assessing resilience to attacks and errors in communication networks


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- Assessing resilience to attacks and errors in communication networks
- Hiding membership to communities in social networks
- Preventing detection of confidential communities


## k-Trusses

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Problem (MIN-k-TBS)
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## Problem (MIN-k-CBS)

Find a minimum set of edges incident to $U$ such that no node in $U$ is in a $k$-truss.

## NP-hardness

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For $\epsilon>0$, both MIN-k-TBS and MIN-k-CBS cannot be approximated within a multiplicative factor of $(k-2-\epsilon)$, assuming the unique games conjecture.

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- Let $G$ be an instance of MIN-3-TBS.
- Turn each triangle in $G$ into a $k$-clique by adding $k-3$ vertices to obtain $G^{\prime}$.



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- MIN- $k$-TBS in $G^{\prime}$ is equivalent to MIN-3-TBS in $G$.



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Therefore MIN- $k$-TBS is NP-hard.


## Algorithms

- Exact Algorithm


## CWI

## Algorithms

- Exact Algorithm
- Heuristics


## Truss Decomposition

## Definition ( $k$-Truss)

A $k$-truss is a subgraph in which each edge is contained in at least $k-2$ triangles of the subgraph.

Algorithm
$k=2$ (every graph $G$ is a 2-truss)
while $G$ non-empty:

- $k=k+1$
- while there is an edge in less than $k-2$ triangles:
- remove that edge
- $G$ is currently the maximum $k$-truss


## Dynamic Truss Update for Edge Deletion

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Update Algorithm

- Similar Algorithm
- Store the triangles each edge is in.
- Store the number of triangles of trussness $k$ each edge is in.
- Time is proportional to the updated number of triangles.
- Propagation can cause $O\left(\mathcal{T}_{G}\right)$ time per update.
- Very good amortized time complexity $O\left(t(G) \cdot \mathcal{T}_{G}\right)$


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Idea

- List all minimal $k$-trusses
- Find a minimum hypergraph transversal


## Max-Truss Breaking Heuristic (MBH)

- Let $k^{\prime}$ be the highest number such that $G$ contains a $k^{\prime}$-truss.
- Let $M$ be the maximal $k^{\prime}$-truss.
- Let $\left|\mathrm{TRI}_{\geq k^{\prime}}(M, e)\right|$ be the number of triangles in $M$ containing edge $e$.


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- Let $\left|\mathrm{TRI}_{\geq k^{\prime}}(M, e)\right|$ be the number of triangles in $M$ containing edge $e$.

While $k^{\prime} \geq k$ :
$\mathrm{MBH}_{S}$ Delete the edge with the highest $\left|\mathrm{TRI}_{\geq k}(M, e)\right|$
$\mathrm{MBH}_{C}$ Delete the edge in the highest $\left|\mathrm{TRI}_{\geq k}(M, e)\right| /\left|\mathrm{TRI}_{<k}(M, e)\right|$

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- Give more weight to edges which are in many triangles
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Delete edge with minimal
$\sum_{\{e, f, g\} \in \operatorname{TRI}_{\geq k}} \frac{\left|\mathrm{TRI}_{\geq k}(M, e)\right|}{\max \left(\left|\mathrm{TRI}_{\geq k}(M, f)\right|-k+2,1\right)}+\frac{\left|\mathrm{TRI}_{\geq k}(M, e)\right|}{\max \left(\left|\mathrm{TRI}_{\geq k}(M, g)\right|-k+2,1\right)}$

## Small Networks



## Benchmarking Large Networks

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$A$ graph on $n$ vertices and more than $\frac{(k-2)}{(k-1)} \cdot \frac{n^{2}}{2}$ edges contains a $k$-clique.

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Partition the graph into cliques.
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Theorem (Conte et al.)
A graph with $m$ edges and $\mathcal{T}_{G}$ triangles has trussness at least $\mathcal{T}_{G} / m+1$.

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