

Finding low-tension communities

A **community-search** problem with **opinion dynamics**:
find a subgraph that connects the seed nodes and has low social tension.

► SOCIAL NETWORKS WITH PROFILES

We consider a social network $G = (V, E)$ where nodes in V represent individuals, and edges in E represent their interactions.

Each individual has his own preferences, habits, opinions, etc. However, individuals may choose not to act in accordance with their true preferences as they try to minimize peer pressure by conforming their preferences to those of their peers.

Each node i is associated to a **latent profile**, x_i , representing the individual's true preferences, and a **conformed profile**, f_i , representing his expressed preferences, both take value in the interval $[0, 1]$.

► PROBLEM STATEMENT

Given a network $G = (V, E)$, latent profiles \mathbf{x} and a set of seed nodes $Q \subseteq V$, find $V' \subseteq V$ such that $Q \subseteq V'$, the graph G' induced by V' on G is connected and $T(G', \mathbf{x}, \mathbf{f})$ is minimized, where \mathbf{f} is computed by the repeated averaging model on G' .

► ALGORITHMS

SPANNING-TREE APPROACH

Build a spanning tree between the query nodes, using the 2-approximation to Steiner tree problem.

Different ways to score the tree yield variants:

- CTree(e) Number of edges involved
- CTree(s) Sum of weights of the edges along the path

Computing the conformed profiles is costly. Instead, use proxy for contribution of each pair of neighboring nodes. Assign weight $w_{ij} = |x_i - x_j|$ to each edge $(i, j) \in E$.

TOP-DOWN APPROACH

Iteratively remove nodes until it is no longer possible without disconnecting the query nodes.

Different way to pick next node yield variants:

- CPeel(r) Random
- CPeel(s) Sum of adjacent edges weight
- CPeel(m) Max of adjacent edges weight

► MEASURING THE SOCIAL TENSION

Each node i bears an **inner tension** caused by the difference between its own latent and conformed profiles, and a **cross tension** caused by the difference between its own conformed profile and those of its neighbors:

$$T_i(G, \mathbf{x}, \mathbf{f}) = (x_i - f_i)^2 + \sum_{j \in N_G(i)} (f_i - f_j)^2.$$

The **social tension** of the network is the sum of the individual tensions:

$$\begin{aligned} T(G, \mathbf{x}, \mathbf{f}) &= \sum_{i \in V} T_i(G, \mathbf{x}, \mathbf{f}) \\ &= \sum_{i \in V} ((x_i - f_i)^2 + \sum_{j \in N_G(i)} (f_i - f_j)^2) \\ &= \sum_{i \in V} (x_i - f_i)^2 + \sum_{(i,j) \in E} 2(f_i - f_j)^2. \end{aligned}$$

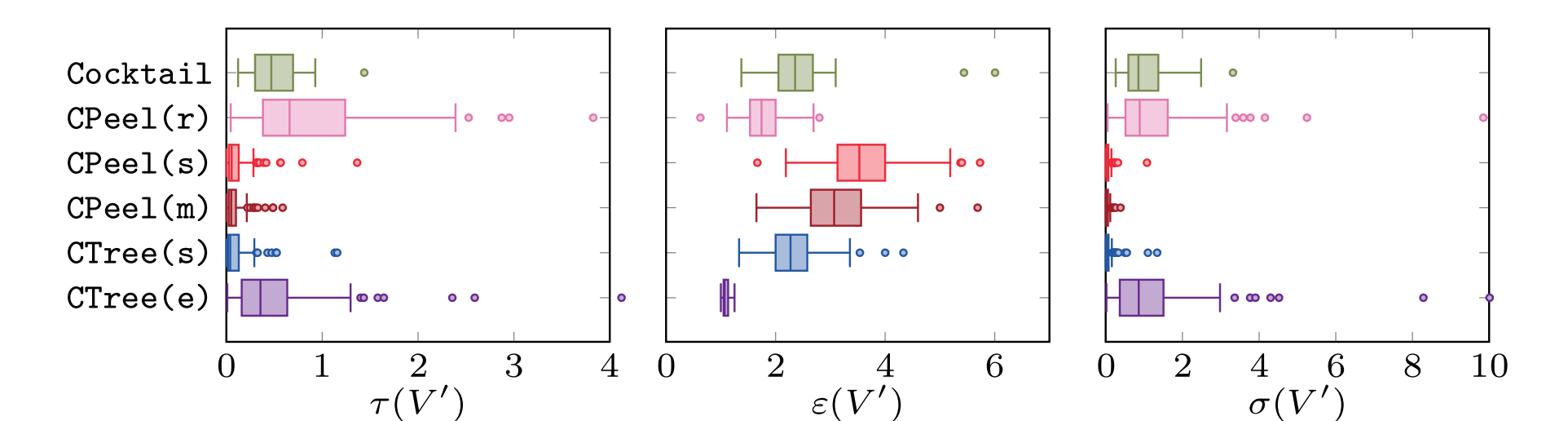
► CONFORMATION PROCESS

Consider a **repeated averaging process** where at each step each node adjusts its conformed profile by setting it to the average of its latent profile and the conformed profile of its neighbors:

$$f_i(t+1) = \frac{x_i + \sum_{j \in N_G(i)} f_j(t)}{1 + |N_G(i)|}.$$

The repeated averaging model is equivalent to choosing f_i to minimize $T_i(G, \mathbf{x}, \mathbf{f})$. It yields a Nash equilibrium for the tension, not a social optimum.

Results for the 2-hop ego-network of C.Papadimitriou with single-attribute latent profiles derived from conferences.



► EXPERIMENTS

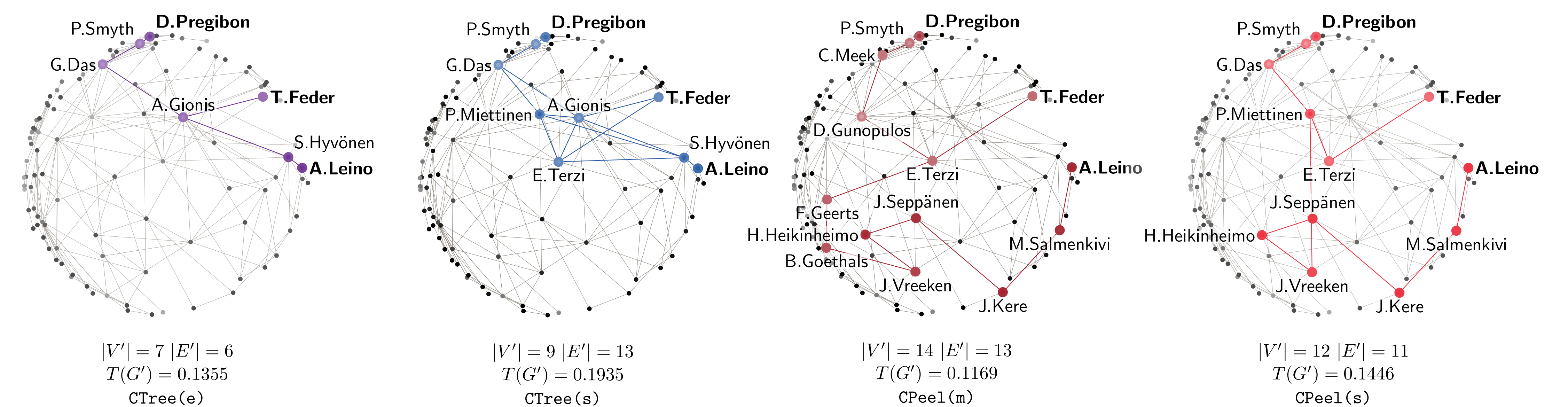
Comparing the algorithm variants and the Cocktail algorithm [3].

Evaluation measures:

- standardized social tension** (main measure) $\tau(V') = T(V') / (2e_b \cdot \overline{w^2}(V))$,
- standardized solution size** (auxiliary measure) $\epsilon(V') = |E(V')| / e_b$, and
- standardized average edge weight** (auxiliary measure) $\sigma(V') = \overline{w^2}(V') / \overline{w^2}(V)$.

where e_b is the size of the minimum spanning tree connecting the query nodes, and $\overline{w^2}(V)$ is the average squared edge weight in V .

Example solutions for connecting three seed nodes in the 1-hop ego-network of H.Mannila with single-attribute latent profiles derived from keywords.



► REFERENCES

- [1] Bindel, Kleinberg and Oren (2011) *How Bad is Forming Your Own Opinion?* FOCS
- [2] Gionis, Terzi, and Tsaparas (2013) *Opinion Maximization in Social Networks.* SDM
- [3] M. Sozio and A. Gionis (2010) *The community-search problem and how to plan a successful cocktail party.* KDD