

Efficient Social-aware Content Placement in Opportunistic Networks

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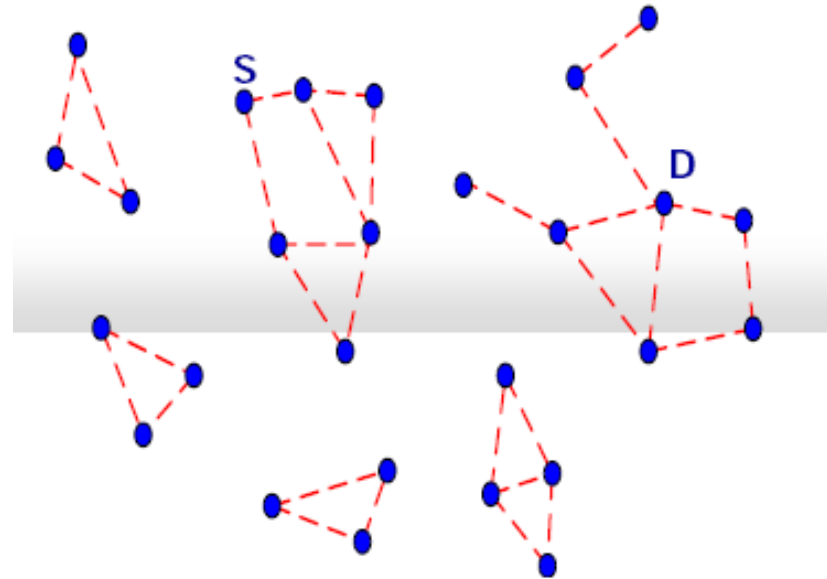
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a delay - tolerant environment

A wireless network that is very sparse and partitioned

- ✓ disconnected clusters of nodes appear

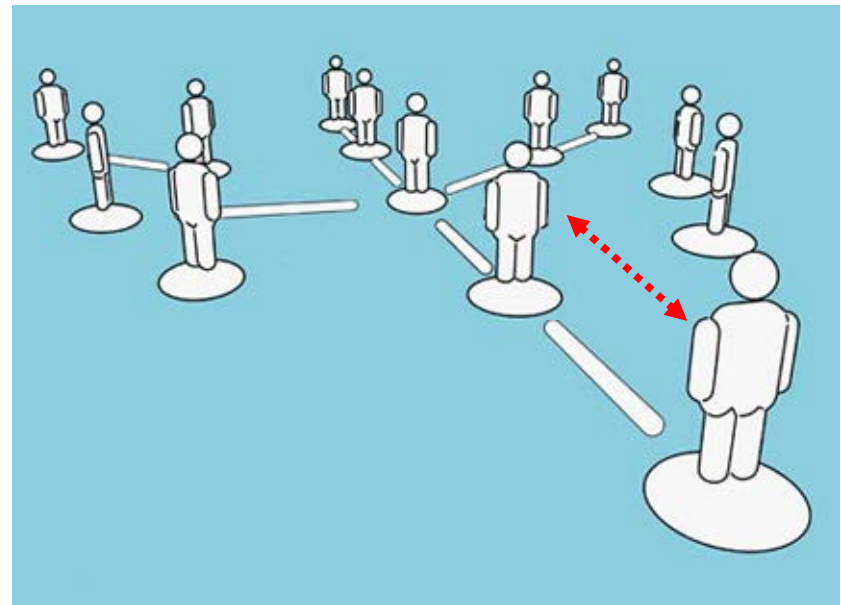


- Nodes are highly mobile making the clusters change often over time
- No contemporaneous end-to-end path

graph representation of a DTN

A graph that stands for a DTN:
add an edge between two nodes if

- ✓ the frequency of encounters
or
- ✓ the cumulative time of direct
contact
exceeds a pre-defined threshold



approximate solution to the classic facility location problem*

- ***K-median problem*** : given a fixed number of facilities, minimize the total service cost

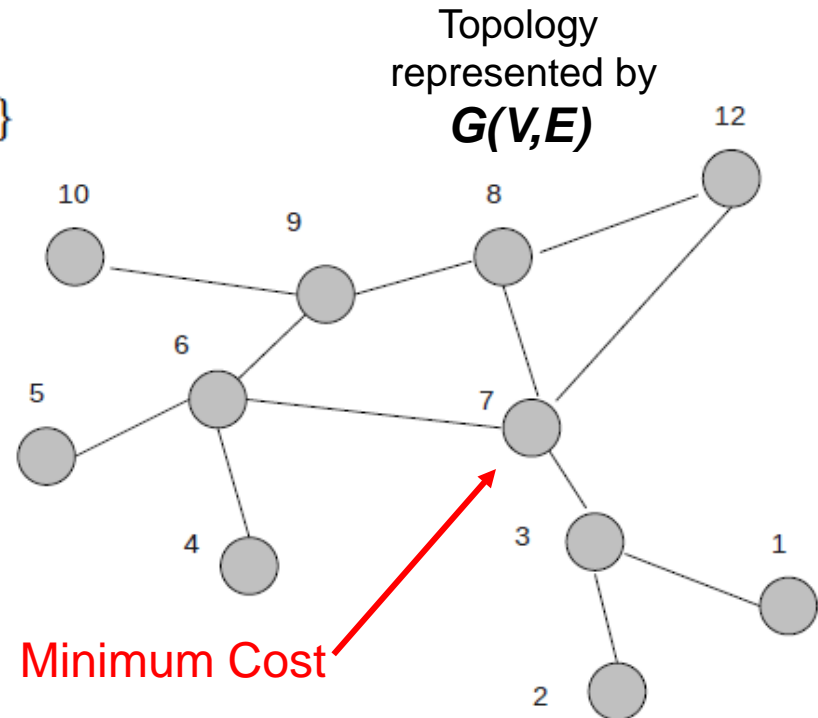
$$Cost(\mathcal{F}) = \sum_{n \in \mathcal{V}} w_n \cdot \min_{x_j \in \mathcal{F}} \{d(x_j, n)\}$$

$d(k, n)$: cost path between nodes k, n

w_n : demand generated by node n

- To cope with constraints, use:

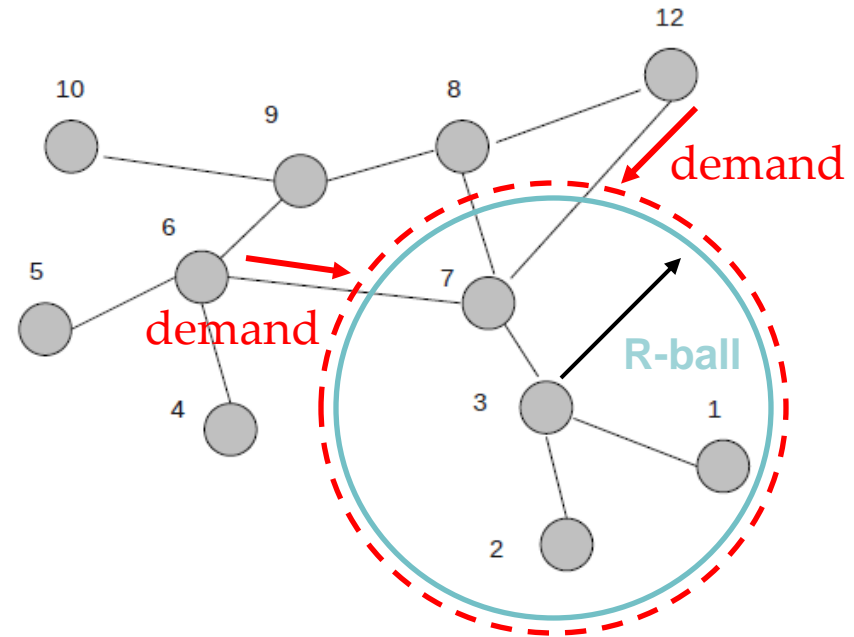
1. Local topology info
2. Local demand info



* P. Mirchandani and R. Francis. *Discrete Location Theory*. John Wiley and Sons, New York, NY, 1990

the R-balls heuristic*

- Solve the optimization problem within a limited neighborhood of selected facilities
- Implicit computation of the demand generated by outer nodes

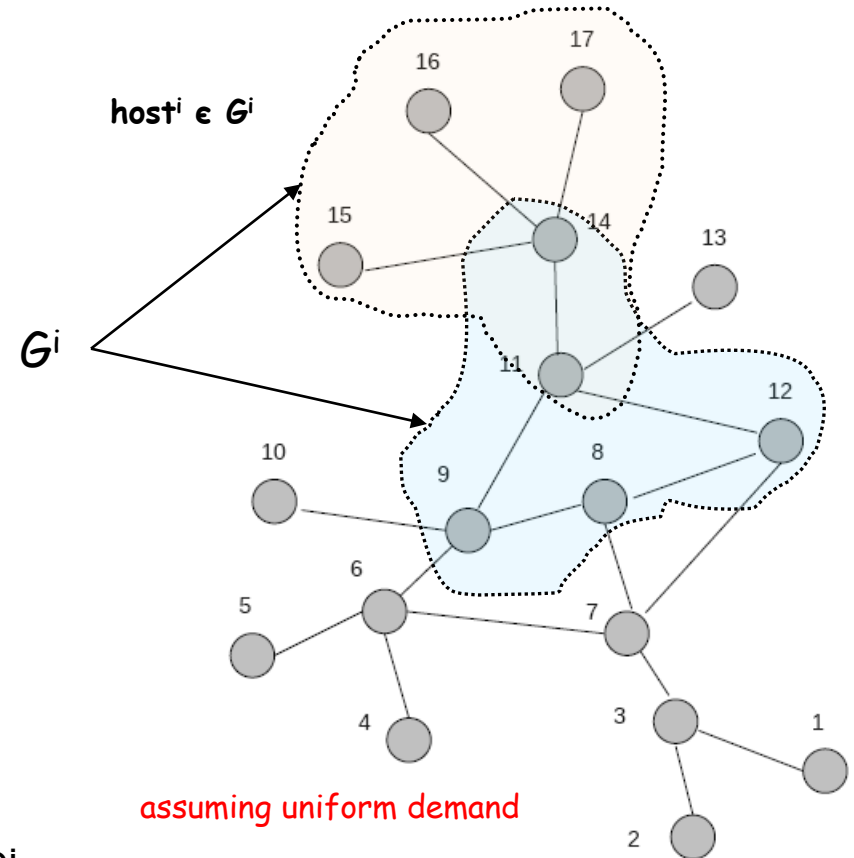


* G. Smaragdakis, N. Laoutaris, K. Oikonomou, I. Stavrakakis, A. Bestavros, “Distributed Server Migration for Scalable Internet Service Deployment”, to appear in IEEE/ACM T-Net. (2010)

content placement: a social-aware approach

1-Median Problem : Determine the physical location of content's single host, in a way that a cost metric is minimized

- Instead of a **large** optimization problem, solve *iteratively* **small-scale** ones, on subgraphs $G^i \in G$, trying to reach the optimal
- Pick G^i nodes around $host^i$ according to a social-inspired criterion
- Criterion should achieve *quantification* of demand from nodes **not** included in G^i



content placement: a social-aware approach

a measure of the importance of node's u social position : lies on paths linking others

Betweenness Centrality (u): portion of all pairs shortest paths of G that pass through node u

$$BC(u) = \sum_{s=1}^{|V|} \sum_{t=1}^{s-1} \frac{\sigma_{st}(u)}{\sigma_{st}}$$

Conditional Betweenness Centrality (u, t): portion of all shortest paths of G from node u to *target* t , that pass through node u

a measure of the importance of node's u social position : ability to control information flow towards *target* node

$$CBC(u;t) = \sum_{s \in V, u \neq t} \frac{\sigma_{st}(u)}{\sigma_{st}}$$



content placement: a social-aware approach

CBC criterion

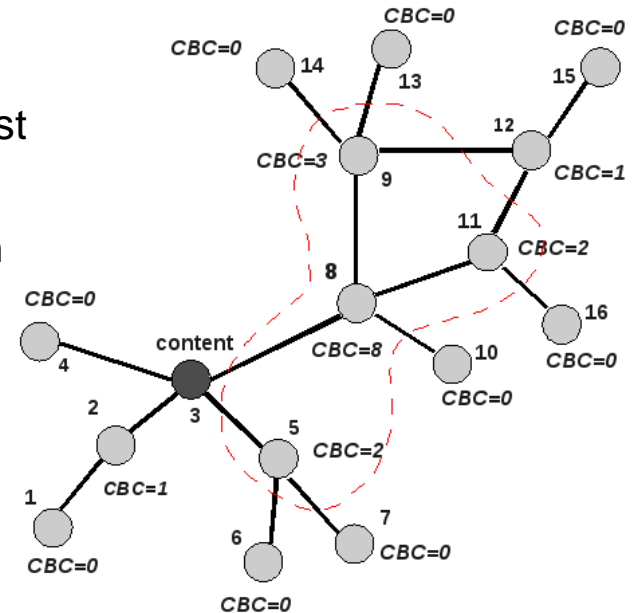
- small-scale *1-median* solution derived from any cost-effective algorithm (i.e. approx., heuristic, enumeration)

Algorithm 1 Social-aware *1-median* in $G(V,E)$

1. choose randomly node s
 2. place *CONTENT* in s
 3. $C_{current} \leftarrow \infty$
 4. $\forall u \in G$ compute $CBC(u; s)$
 5. let G_s^o be $\alpha\%$ of G nodes with top CBC values
 6. *1-median* solution in $G_s^o \rightarrow$ node $Host$
 7. $C_{next} \leftarrow C(Host)$
 8. while $C_{next} < C_{current}$ do
 9. move *CONTENT* to $Host$
 10. $C_{current} \leftarrow C_{next}$
 11. $\forall u \in G$ compute $CBC(u; Host)$
 12. let G_{Host}^i be $\alpha\%$ of G nodes with top CBC values
 13. *1-median* solution in $G_{Host}^i \rightarrow$ node $NewHost$
 14. $Host \leftarrow NewHost$
 15. $C_{next} \leftarrow C(NewHost)$
 16. end while
-

content placement: a social-aware approach

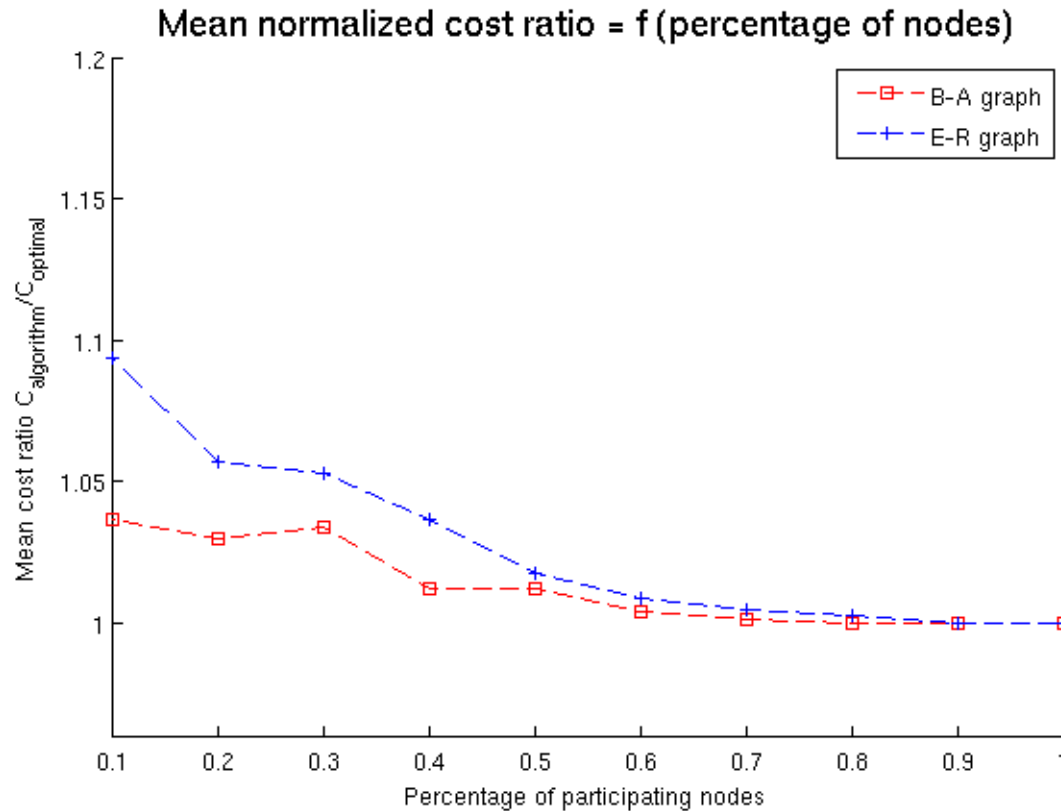
- CBC criterion picks neighboring nodes to the host
- Selected nodes “stretching” in a certain direction
- Solution space moved by selection towards directions populated by capable nodes of transporting information efficiently
- “Socially significant” direction is valid under any demand model but exploited only under uniform demand hypothesis



simulation settings

- Generate both E-R and B-A random graphs of $N=100$ nodes
- 10 simulation runs for any chosen configuration
- Results presented are the averages over the runs
- Regenerate the graph for each simulation run
- Probability of a link to exist approximates the theoretical value of the target graph model

simulation results (1/3)



Standard Error associated with each mean: $0.07\% \leq SE \leq 2.88\%$

simulation results (2/3)

- *Random policy* : place the content on a randomly selected node and serve demands
- ✓ Test the most costly scenario : $\max \beta_{algorithm}(\alpha) = \beta_{algorithm}(0.1)$

$$\beta_{algorithm}(\alpha) = E\left[\frac{C_{algorithm}}{C_{optimal}} \right]$$

$$\beta_{random} = E\left[\frac{C_{random}}{C_{optimal}} \right]$$

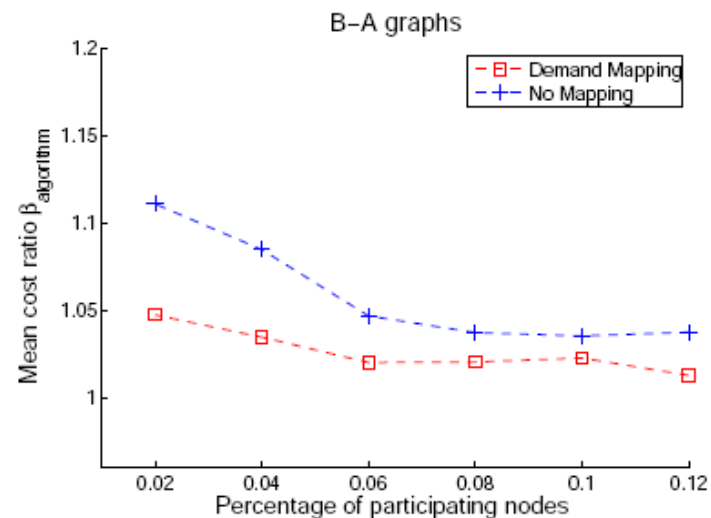
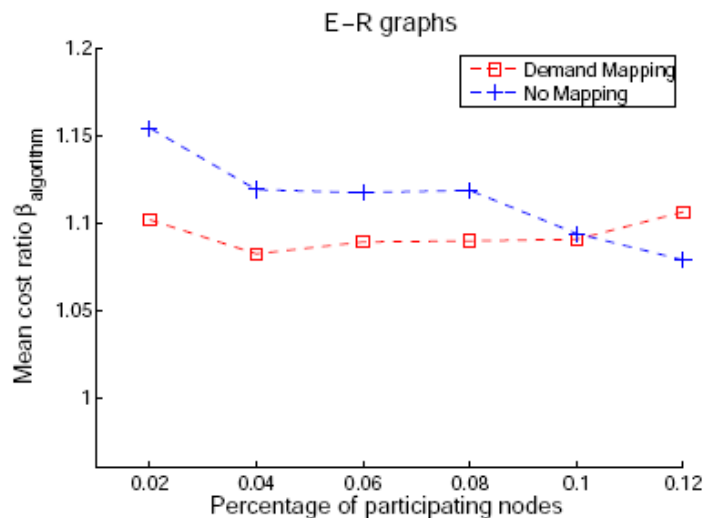
SOCIO-AWARE *1-median* VS SIMPLE RANDOM PLACEMENT

<i>E-R graphs</i>		<i>B-A graphs</i>	
max. $\beta_{algorithm}$	β_{random}	max. $\beta_{algorithm}$	β_{random}
1.0938	1.2736	1.0366	1.4953

simulation results (³/₃)

Demand mapping

- New demand model for the small-scale optimization to improve performance
- Mapping between CBC values of subgraph G^i and incoming demand load
- When the percentage of participating nodes $\rightarrow 1$, we solve a different problem



future work

- ✓ study whether such mechanisms can be effectively applied
- ✓ consider the threshold-based mapping of contacts to static graphs
- ✓ expand this work to include non-uniform demand patterns

Questions ?