COMMUNITY DETECTION & RANDOM WALK

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Outline

- Random Walk
- Walktrap Algorithm
- Markov Clustering Algorithm

Random Walk on Graphs

 At each time step, a walker is on a vertex. Then he chooses the next vertex randomly and uniformly among its neighbors.

Random Walk on Graphs

 Hitting time: Expected number of steps in a random walk to go from u, and ends at v for the first time

Commute time: Sum of Hitting time from u to v, and from v to u

Why Random Walk ?

• Use random walk to define how community is formed.

• Particularly, use random walk in defining measures of vertex similarity

Why Random Walk ?

- Capture well and intuitively the community structure.
- Can be computed efficiently

Walktrap

- By Pascal Pons and Matthieu Latapy
- Implemented in iGraph library.
- Claim:
 - In real-world cases: time O(V^2 log V)
 - Worst case : time O(E V^2)
 - Space O(V^2)

Walktrap

 Intuition: random walkers tends to get 'trapped' into densely connected parts (communities)

Walktrap

- Consider the probability P^t to go from vertex i to j in t steps.
- Establish a distance measure between vertices (and between clusters) based on P^t
- Use clustering algorithm (Ward's method) to find communities

- By van Dongen
- Simple to implement.
- Time complexity: O(V^3)

• Intuition: More or less same as Walktrap

- Iterative clustering algorithm.
- Two main operations: expansion and inflation

- Start with the initial transfer matrix
- Expansion: Calculate the probability of a random walker going between vertices in t steps (by matrix multiplication)
- Inflation: Highlight probabilities of intracluster walks and demote inter-cluster walks

- Converges to a matrix 0-1 matrix M (like an adjacency matrix).
- M represents a disconnected graph. Each connected component is a community

Work to Do

• Do testing and benchmarking for the two algorithms

References

Pascal Pons, Matthieu Latapy, Computing communities in large networks using random walks

http://micans.org/mcl/