Increasing accuracy in map visualizations of tagged networks

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Outline

- Introduction
 - Motivation
 - GMAP
 - Limitations of GMAP
 - Formalizing the problem
- Similarity
 - Notions of similarity
 - Calculating similarity
 - Modifications
- Miscellaneous
 - Maintaining aesthetics
 - Algorithmic complexity
- Conclusion

Motivation

- Suppose that you have a dataset *D* where each individual record *i* has a set of tags.
- How can we best visualize this network?

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"id": 30002,
"title": {
  "userPreferred": "Berserk"
"popularity": 93568,
"genres": [
 "Action",
 "Adventure",
 "Drama",
 "Fantasy",
 "Horror",
 "Psychological"
"averageScore": 92
"id": 30013.
"title": {
  "userPreferred": "ONE PIECE"
"popularity": 89948,
"genres": [
 "Action",
 "Adventure",
  "Comedy",
  "Fantasy"
"averageScore": 91
```

Fig 1. Anilist manga sorted by score

GMAP

- Gansner, Hu & Kuborov (2010) published GMAP an efficient algorithm for visualizing networks as maps.
- GMAP works by:
 - Applying a force directed algorithm to embed the graph
 - Performing cluster analysis
 - Generating Voronoi diagrams and adding noise to the boundaries
 - And finally, coloring the map

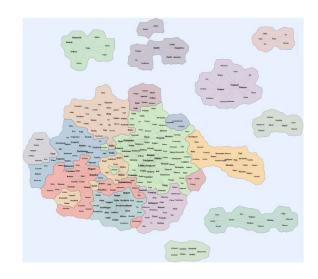


Fig 2. Application of GMAP

Limitations of GMAP

- False positives
 - Connected components end up next to each other without a strong connection between them.
 - Implies association between elements of the components
- False negatives:
 - Regions are not next to each other but they should be.
 - A harder problem to solve on a 2D-plane.

Formalizing the problem

- Assume that we are given a graph G(V,E) and a mapping $\Gamma: V \to T$ from the vertices to a set of possible tags. An edge exists between vertices i and j if and only if $|\Gamma(i) \cap \Gamma(j)| \ge 1$.
- We need to output a visualization *V* of the graph *G* such that the number of dissimilar regions are minimized.

Notions of Similarity

- How do we define the concept of "similarity" between (sub)graphs or the elements within?
 - o Isomorphism, edit distance, common subgraphs, etc
- Intuitive notion of similarity between regions
 - Vertex similarity as a local property
 - Jaccard similarity, cosine similarity, etc.
 - Holmes et al (2005)'s definition of similarity
 - *i* is similar to *j* if *i* has a neighbor *k* that is also similar to *j*
 - A recursive definition



Fig 3. A map of movies

Calculating similarity

- A mathematically complicated process.
- $\bullet \qquad S_{ij} = \phi \sum_{v} A_{iv} S_{vj} + \psi \delta_{ij},$
- A vertex is similar to itself.
- For large graphs, eigenvalues provide approximate number of expected paths.
- Finally similarity matrix can be calculated as:

$$\mathbf{S} = 2m\lambda_1 \mathbf{D}^{-1} \left(\mathbf{I} - \frac{\alpha}{\lambda_1} \mathbf{A} \right)^{-1} \mathbf{D}^{-1},$$

$$\mathbf{DSD} = \frac{\alpha}{\lambda_1} \mathbf{A}(\mathbf{DSD}) + \mathbf{I}.$$

Modifications

- Holmes et al's algorithm was designed for adjacency matrices of only 1s and 0s.
- Since we are using tagged networks, we will utilize Γ to modify the adjacency matrix to operate on values in [0,1].
- We do not need to work on the entire graph only on connected components.

Maintaining Aesthetics

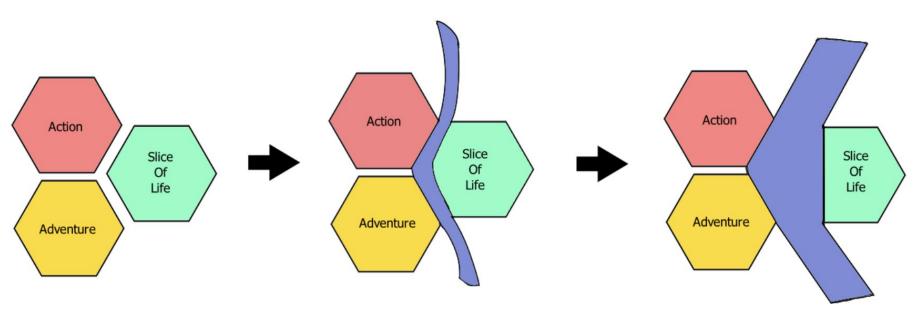


Fig 4. Adding rivers/lakes.

Algorithmic Complexity

- Complexity of GMap is O(|V|log|V|).
- Addition of extra components does not affect asymptotic bound.
- Calculating similarity uses only linear algebra operations
 - Matrix multiplication is typically $O(|V|^{(2.37)})$
 - Generally much faster and hardware optimized
 - Lots of libraries exist even for high level languages like Python

Conclusion

- We have shown a possible way to improve the aesthetics and accuracy of GMap by embedding a bounded number of points and adding repulsive force from those points.
- The location of these points can be found by calculating the similarity matrix for a subset of the points in the graph.
- Calculating similarity is expensive asymptotically but generally fast.
- Further work can be done to find a way to reduce the number of false negatives.

References

- 1. Gansner, Emden & Hu, Yifan & Kobourov, Stephen. (2010). GMap: Visualizing graphs and clusters as maps. Computer Graphics and Applications, IEEE. 30. 201 208. 10.1109/PACIFICVIS.2010.5429590.
- 2. Leicht, EA & Holme, Petter & Newman, M. (2006). Vertex similarity in networks. Physical review. E, Statistical, nonlinear, and soft matter physics. 73. 026120. 10.1103/PhysRevE.73.026120.
- 3. Zager, Laura & Verghese, George. (2008). Graph similarity scoring and matching. Applied Mathematics Letters. 21. 86-94. 10.1016/j.aml.2007.01.006.