#### Data Visualization (CSCI 627/490)

Networks

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## 3D to 2D: Projection

















### Projection Classification













# Don't Just Create Population Maps!





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PET PEEVE #208: GEOGRAPHIC PROFILE MAPS WHICH ARE BASICALLY JUST POPULATION MAPS









### Cartograms



- Data: geographic geometry data & two quantitative attributes (one part-of-whole)
  - Derived data: new geometry derived from the part-of-whole attribute
- Tasks: trends, comparisons, part-of-whole
- How: area marks from derived geometry,
  - color hue/saturation/luminance
- Scalability: thousands of regions
- Design choices:
  - Colormap
  - Geometric deformation











#### D3 Map Example







## Midterm

- Thursday, October 15, 9:30am-10:45am
- Covers material through last week
- Format:
  - Multiple Choice
  - Free Response (often multi-part)
  - discussed

#### - CS 627 students will have extra questions related to the research papers





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- Two Possibilities:
  - Create an interactive visualization
  - Work on a research project
- Dataset Choices:
  - Information Wanted: A dataset of immigrants looking for family & relatives - Vessel Tracking: A dataset of ship movements near the United States - Prescription Drug Costs: A dataset of medicine costs in the United States
- First step:
  - Examine data, determine format, attribute characteristics, etc. - Think of questions a visualization can help answer







### Networks

- Why not graphs?
  - Bar graph
  - Graphing functions in mathematics
- Network: nodes and edges connecting the nodes
- Formally, G = (V, E) is a set of nodes V and a set of edges E where each edge connects two nodes.
- Nodes == items, edges connect items
- Both nodes and edges may have attributes









### Arrange Networks and Trees







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### Molecule Graph







### Molecule Graph







### Molecule Graph







#### Web Sites as Graphs (amazon.com)



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#### Social Networks











### Networks as Data

#### Nodes

ID	Atom	Electrons	Protons
0	Ν	7	7
1	С	6	6
2	S	16	16
3	С	6	6
4	Ν	7	7

Edges

ID1	ID2	Bonds
0	1	┱
1	2	1
T	3	2
З	4	1







# Node-Link Diagrams

- Data: nodes and edges
- Task: understand connectivity, paths, structure (topology)
- Encoding: nodes as point marks, connections as line marks
- Scalability: hundreds
- ...but high **density** of links can be problematic!
- Issue with the encoding?









### Arc Diagram









## Network Layout

- Need to use spatial position when designing network visualizations
- Otherwise, nodes can **occlude** each other, links hard to distinguish
- How?
  - With bar charts, we could order using an attribute...
  - the data usually)
- Possible metrics:
  - Edge crossings
  - Node overlaps
  - Total area

- With networks, we want to be able to see connectivity and topology (not in





## Force-Directed Layout

- Nodes push away from each other but edges are springs that pull them together • Weakness: nondeterminism, algorithm may produce difference results each time it runs









# Constraint-Based Optimization (CoLa)

- Higher quality layout
- More **stable** in interactive applications (no "jitter")
- Allows user specified constraints such as alignments and grouping
- Can avoid overlapping nodes
- Provides flow layout for directed graphs
- May be less scalable to very large graphs
- Can route edges around nodes

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[T. Dwyer et al. (WebCoLa); M. Bostock (Example), 2018]



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## sfdp



JGD\_Homology@cis-n4c6-b14. 7220 nodes, 13800 edges.









#### "Hairball"



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JGD\_Homology@cis-n4c6-b4. 26028 nodes, 100290 edges.







## Hierarchical Edge Bundling













## Hierarchical Edge Bundling















## Hierarchical Edge Bundling

- Flexible and generic method
- - information
  - explicit adjacency edges between their respective child nodes

 Reduces visual clutter when dealing with large numbers of adjacency edges Provides an intuitive and continuous way to control the strength of bundling. - Low bundling strength mainly provides low-level, node-to-node connectivity

- High bundling strength provides high-level information as well by implicit visualization of adjacency edges between parent nodes that are the result of













## Bundling Strength

















## Adjacency Matrix

- Change network to tabular data and use a matrix representation
- Derived data: nodes are keys, edges are boolean values
- Task: lookup connections, find wellconnected clusters
- Scalability: millions of edges
- Can encode edge weight, too















### Cliques in Adjacency Matrices

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#### Structures from Adjacency Matrices











# Node-Link or Adjacency Matrix?

- adjacency better for large graphs
- Multi-link paths are hard with adjacency matrices
- Immediate connectivity or neighbors are ok, estimating size (nodes & edges also ok)
- People tend to be more familiar with node-link diagrams
- Link density is a problem with node-link but not with adjacency matrices

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• Empirical study: For most tasks, node-link is better for small graphs and







