An Empirical Investigation of Statistical Backbone Filtering Techniques
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What is a Network Backbone?

Filtering Noise?
Preserving Topological Properties?  

No Universal Definition!

Types of Backbone Filtering Techniques

1. Structural Backbone Filtering Techniques.
2. Statistical Backbone Filtering Techniques.

Statistical Backbone Filtering Techniques

- **Noise Corrected Filter**
  
  edge weights are drawn from a binomial distribution. However, it estimates the probability of observing a weight connecting two nodes using a Bayesian framework.

- **Marginal Likelihood Filter**
  
  each unit edge randomly chooses two nodes, which results in a binomial distribution.

- **ECM Filter**
  
  Employs a null model based on the canonical maximum-entropy ensemble of weighted networks having the same degree and strength distribution as the original network. Employs the empirical cumulative density function to judge statistical significance.

- **LANS Filter**
  
  a node distributes its strength (summation of weights) uniformly on the edges.

- **Disparity Filter**
  
  a node distributes its strength at random through a Polya process, which results in a Beta-binomial distribution.

- **Polya Urn Filter**
  
  edge weights are randomly drawn from the empirical weight distribution.

- **GloSS Filter**
  
  The edge weights are drawn from a binomial distribution. However, it estimates the probability of observing a weight connecting two nodes using a Bayesian framework.

Evaluation Using Four Indicators

<table>
<thead>
<tr>
<th>Filter</th>
<th>Technique Behavior</th>
<th>Giant Component</th>
<th>Isolated Nodes</th>
<th>Weights Retained</th>
</tr>
</thead>
<tbody>
<tr>
<td>Marginal Likelihood</td>
<td>Conservative</td>
<td>Always</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Noise Corrected</td>
<td>Conservative</td>
<td>Always</td>
<td>Low</td>
<td>Small &amp; High</td>
</tr>
<tr>
<td>ECM</td>
<td>Flexible</td>
<td>$\alpha &gt; 10^{-2}$</td>
<td>$\alpha &lt; 10^{-2}$</td>
<td>Small &amp; High</td>
</tr>
<tr>
<td>LANS</td>
<td>Aggressive</td>
<td>Always</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Disparity</td>
<td>Aggressive</td>
<td>$\alpha &gt; 10^{-2}$</td>
<td>High</td>
<td>High</td>
</tr>
<tr>
<td>Polya Urn</td>
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</tr>
</tbody>
</table>

What about the distributions?