

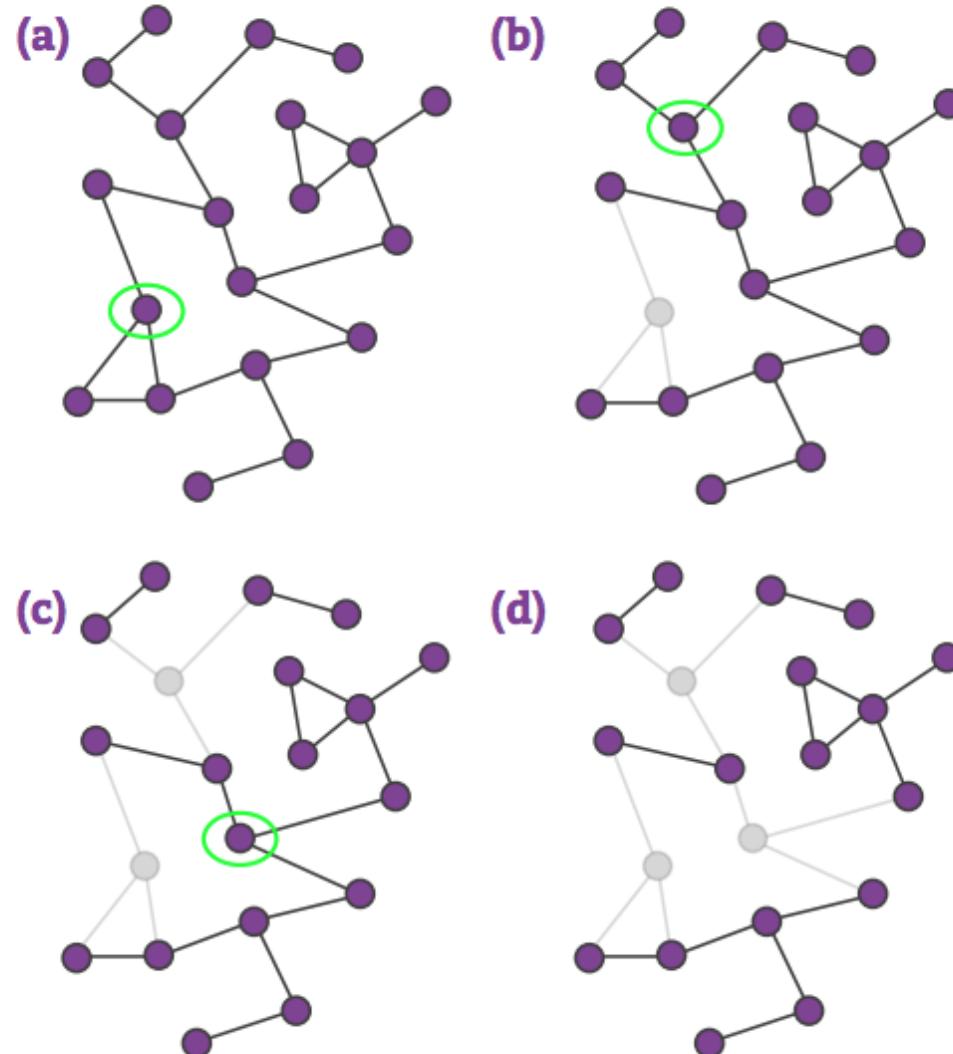
NETWORK ROBUSTNESS

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INTRODUCTION

Network Robustness – the network's structure plays a role in its ability to survive

- Random failures
- Deliberate attacks



MOTIVATIONS

Biology: some mutations lead to diseases while others do not

Ecology: failure of an ecosystem based on human activity

Engineering: communication systems, power grids, and component failures



VULNERABILITY METRICS

Centrality Metrics

Degree

Closeness

Centroid

Eccentricity

Betweenness

Eigenvector

Robustness Measure

Average path length

Efficiency (power grids)

Largest connected component

PSEUDOCODE – BRANDES ALGORITHM

Algorithm 1: Betweenness centrality in unweighted graphs

```
 $C_B[v] \leftarrow 0, v \in V;$ 
for  $s \in V$  do
   $S \leftarrow$  empty stack;
   $P[w] \leftarrow$  empty list,  $w \in V$ ;
   $\sigma[t] \leftarrow 0, t \in V; \sigma[s] \leftarrow 1;$ 
   $d[t] \leftarrow -1, t \in V; d[s] \leftarrow 0;$ 
   $Q \leftarrow$  empty queue;
  enqueue  $s \rightarrow Q$ ;
  while  $Q$  not empty do
    dequeue  $v \leftarrow Q$ ;
    push  $v \rightarrow S$ ;
    foreach neighbor  $w$  of  $v$  do
      //  $w$  found for the first time?
      if  $d[w] < 0$  then
        enqueue  $w \rightarrow Q$ ;
         $d[w] \leftarrow d[v] + 1$ ;
      end
      // shortest path to  $w$  via  $v$ ?
      if  $d[w] = d[v] + 1$  then
         $\sigma[w] \leftarrow \sigma[w] + \sigma[v]$ ;
        append  $v \rightarrow P[w]$ ;
      end
    end
  end
 $\delta[v] \leftarrow 0, v \in V;$ 
//  $S$  returns vertices in order of non-increasing distance from  $s$ 
while  $S$  not empty do
  pop  $w \leftarrow S$ ;
  for  $v \in P[w]$  do  $\delta[v] \leftarrow \delta[v] + \frac{\sigma[v]}{\sigma[w]} \cdot (1 + \delta[w]);$ 
  if  $w \neq s$  then  $C_B[w] \leftarrow C_B[w] + \delta[w];$ 
end
end
```

Uses BFS for unweighted graphs

Runtime: $O(m \times n)$ on unweighted

Space: $O(m+n)$

Weighted networks:

Runtime: $O(m^*n + n^2 \log(n))$

LARGEST CONNECTED COMPONENT

Algorithm 2: Largest Connected Component

Data: Graph $G = (V, E)$

Result: Largest Connected Component of unweighted graphs

$Vis[v] \leftarrow 0, v \in V;$

for $v \in V$ **do**

if $Vis[v] == 0$ **then**
 | $DFSuntil(v);$

end

end

Data: Node v , $Vis[]$, Graph $G = (V, E)$

$v == 1$ **for** u adjacent to v **do**

if $Vis[u] == 0$ **then**
 | $DFSuntil(u, Vis[], G);$

end

end

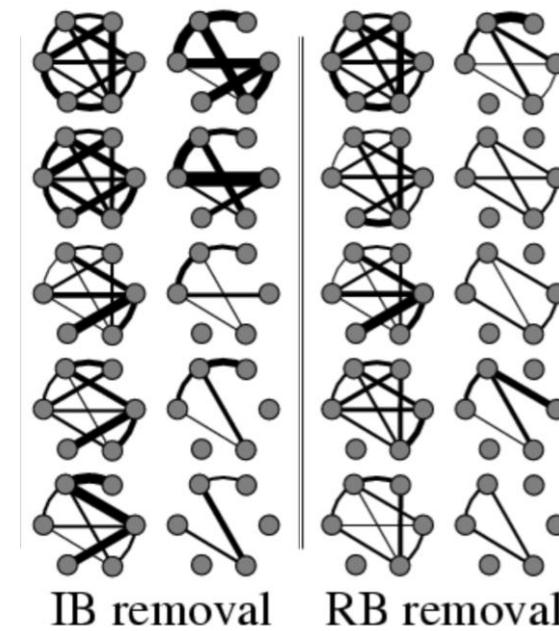
Can be performed with either BFS or
DFS

Runtime: $O(m+n)$

BETWEENNESS CENTRALITY – ROBUSTNESS

Pipeline

- Calculate Betweenness Centrality
- Remove nodes/edges
- Measure robustness
- Compare with initial measure



SEQUENTIAL IMPLEMENTATION

```
for times in range(0,stop):
    # add a column with NaN
    node_betweenness[col] = np.NaN

    # calculate the betweenness of all the nodes
    time_start = time.time()
    all_betweenness = nx.betweenness_centrality(G, normalized=True)
    time_end = time.time()
    # print("Betweenness Time:", time_end-time_start)

    # insert the betweenness into the panda
    for node, bet_val in all_betweenness.items():
        n = int(node)
        node_betweenness.at[n,col] = bet_val

    # remove the largest betweenness node
    max_betweenness = node_betweenness[col].idxmax()
    max_betweenness_val = node_betweenness[col].max()
    G.remove_node((max_betweenness))
    node_remove_list.append(max_betweenness)
    bet_remove_list.append(max_betweenness_val)
    # print("removed:", max_betweenness, "| val:", max_betweenness_val)

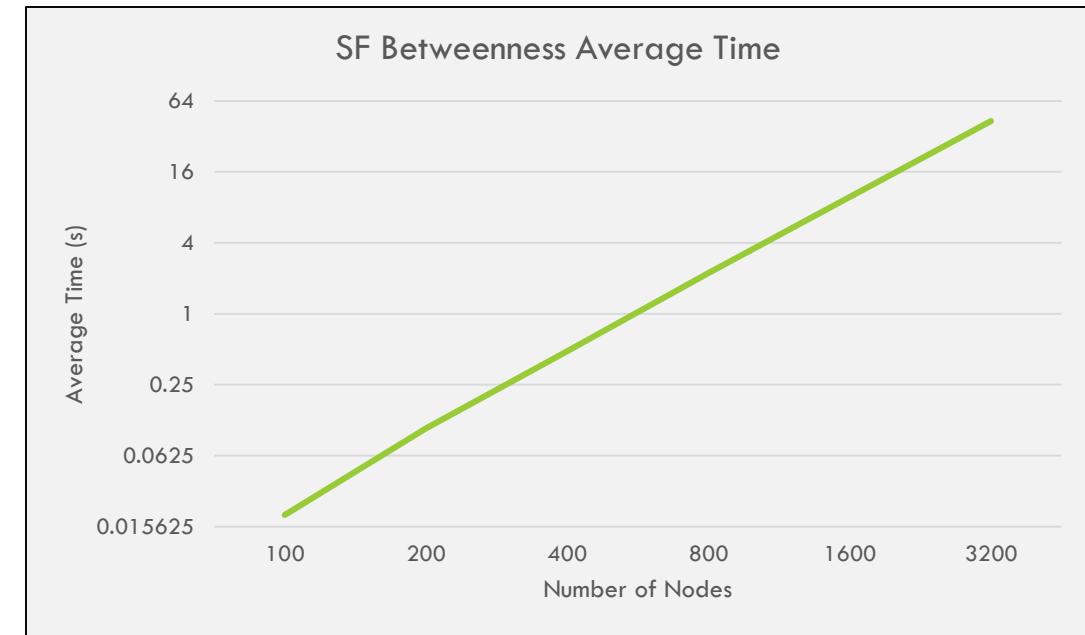
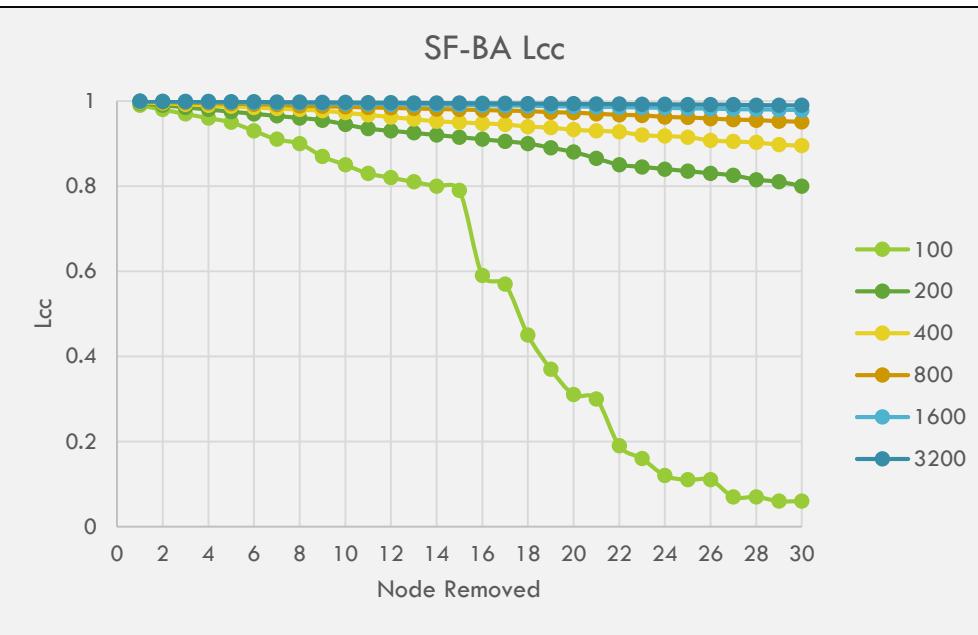
    # find the num of nodes/edges in each subgraph
    all_sungraphs = list(nx.connected_component_subgraphs(G))
    num_nodes_edges_subgraphs = list()
    for k in range(0,len(all_sungraphs)):
        subgraph_nodes = all_sungraphs[k].number_of_nodes()
        subgraph_edges = all_sungraphs[k].number_of_edges()
        num_nodes_edges_subgraphs.append([subgraph_nodes,subgraph_edges])

    # find the largest connected component
    Gcc = max(nx.connected_component_subgraphs(G), key=len)
    Gcc_list.append(Gcc.number_of_nodes())
    col = col + 1
    temp = pd.DataFrame([[time_end-time_start, max_betweenness, max_betweenness_val, \
        Gcc.number_of_nodes()/num_nodes, G.number_of_nodes(), \
        G.number_of_edges(), num_nodes_edges_subgraphs]], \
        columns=['Time', 'Node Rem', 'Bet Value', 'LCC', 'G_nodes', 'G_edges', 'subgraphs (N,E)'])
    df_SF = df_SF.append(temp, ignore_index=True)
```

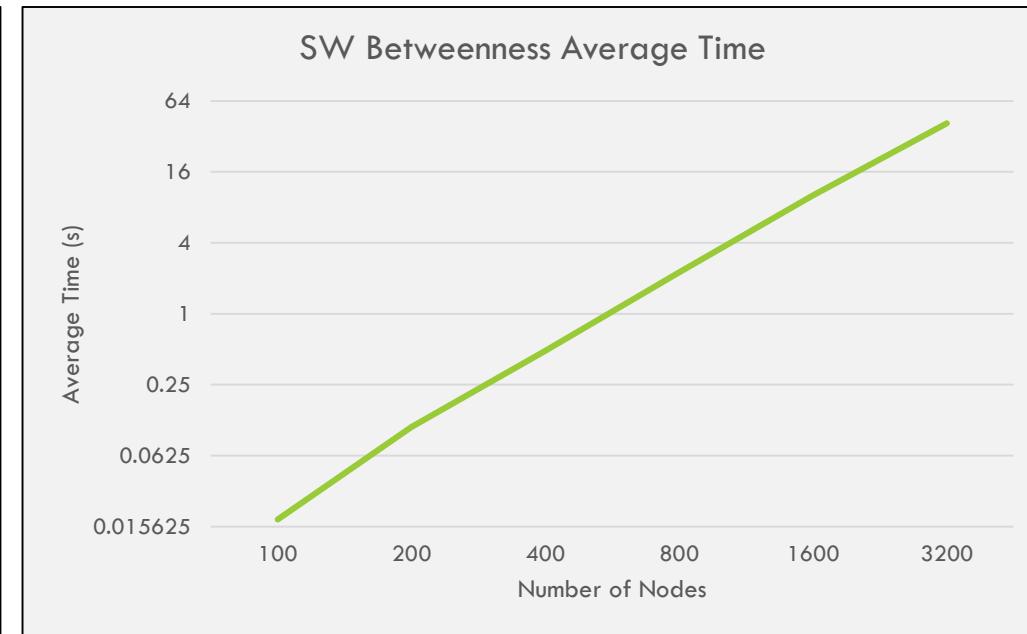
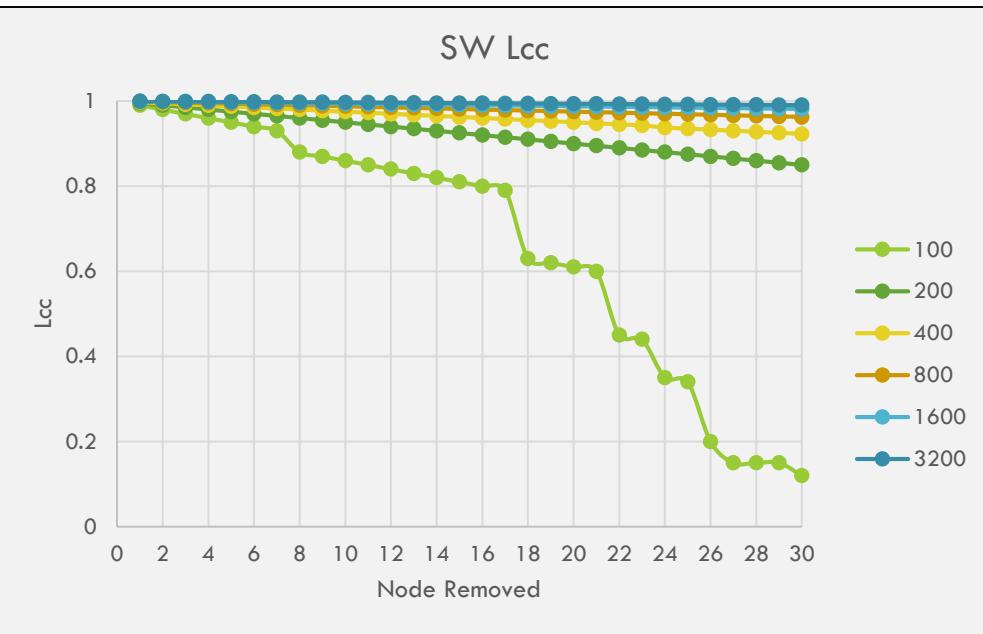
Python

- Networkx
- Uses Brandes Algorithm

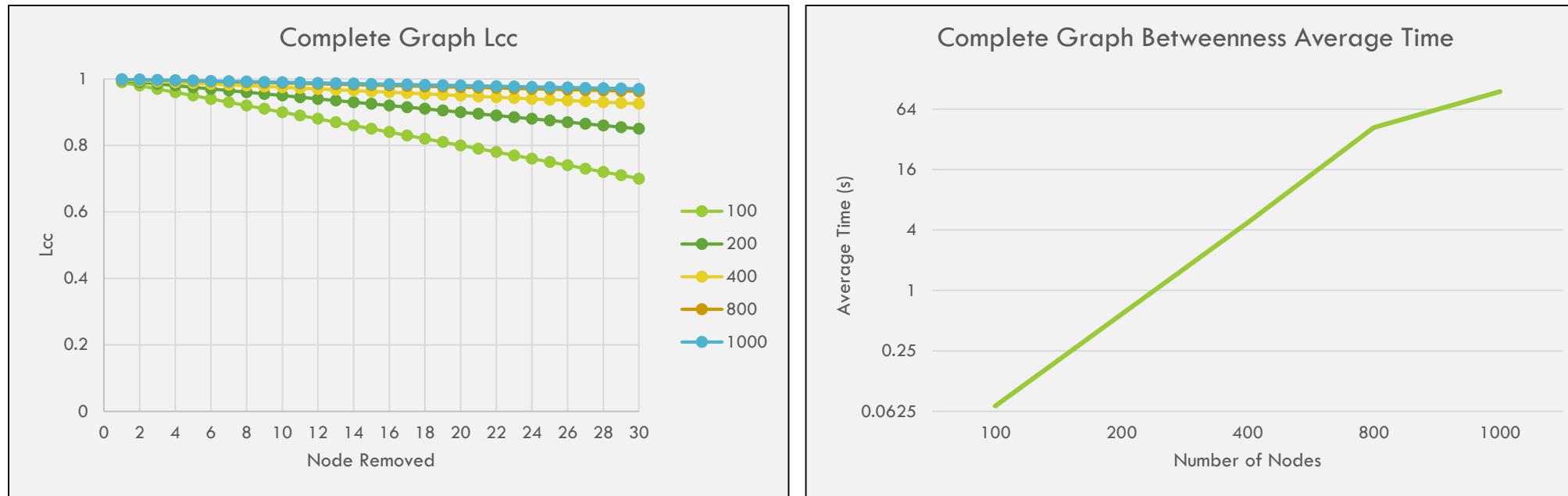
SCALE-FREE MODEL



SMALL WORLD MODEL



COMPLETE GRAPH MODEL



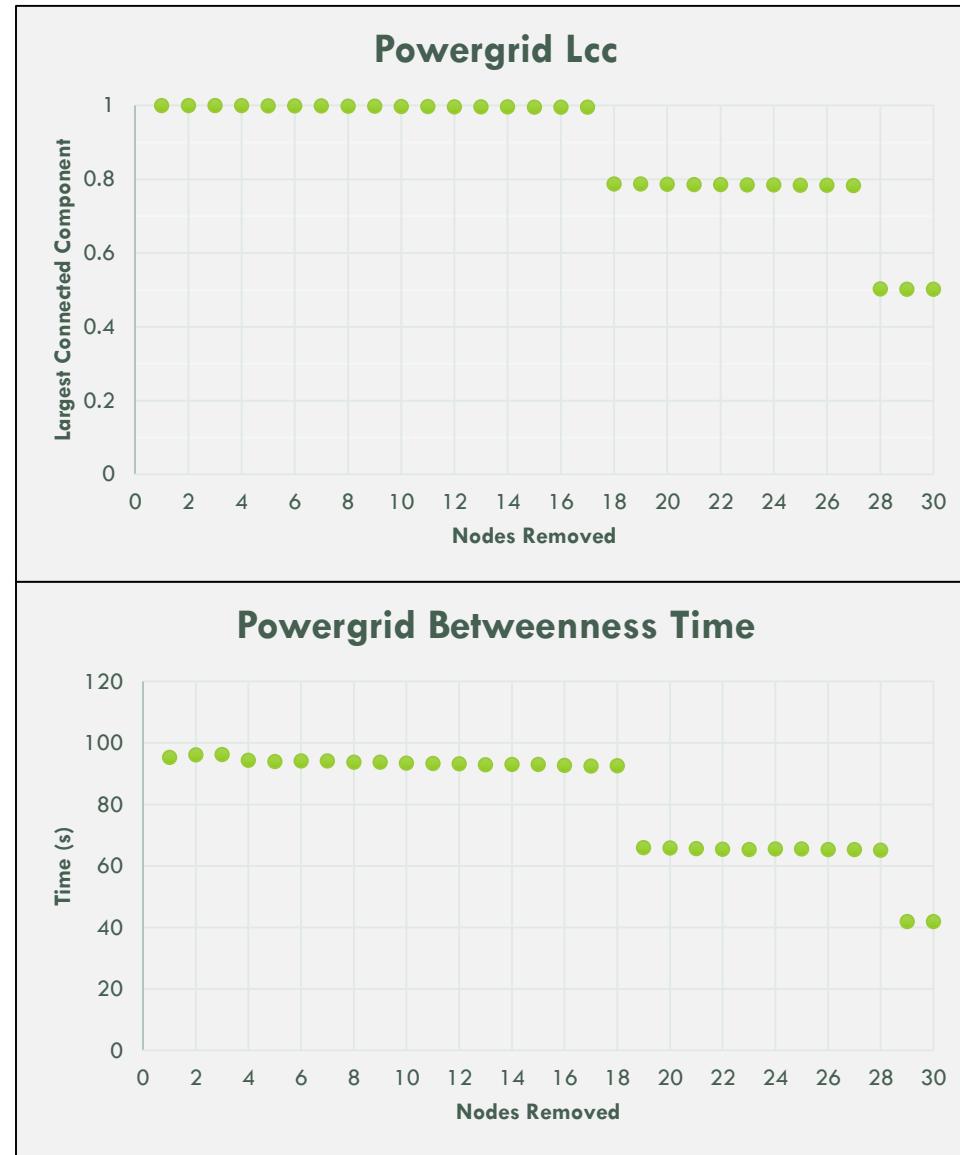
RESULTS

5000 nodes

6500 edges

90-Percentile Diameter: 28.17

Average Shortest Path: 20.09



FUTURE WORK

- Parallelize
- Streaming calculation
- Comparison of IB to RB
- Introduce cascading failure



QUESTIONS/COMMENTS

BACKUP SLIDES

