## Limited Path Percolation in Complex Networks Eduardo López

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- Outline
  - Motivation. Percolation and its effects.
  - Presentation of new limited path length percolation model
  - Scaling theory of new model and results
  - Targeted percolation, theory and results.
  - Conclusions

#### References

"Limited path length percolation in complex networks", López, Parshani, Cohen, Carmi and Havlin, Phys. Rev. Lett. (in press). cond-mat/0702691.

#### Collaborators

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### Motivation: How to go from Salem to Boston?



Question: How many roads need to be closed before most people cannot get to work?Answer: from Percolation theory



## Motivation: What's the problem with percolation?



- Salem-Boston connected with any path!
- Long or short paths OK
- Percolation finds critical percentage  $p_c$  of roads needed to keep cities connected.
- Percolation increases path lengths (and time), i.e., smaller
   p⇒longer path.
- There is practical limit to connectivity ⇔ longer paths not useful.

Answer: sometimes percolation accepts useless paths.

#### Social contact network



- New percolation model applied to complex networks
- •Definition of connection: *i* and *j* are connected if  $l'_{ij} \le a l_{ij}$ •Notation:

 $S_a(p)$ : Largest cluster size at occupation p, length condition a

- Is there a critical occupation p = p̃<sub>c</sub> above which S<sub>a</sub>~N?
  Results: New limited path percolation transition
  Scaling theory
- •Find new critical occupation  $\tilde{p}_c > p_c$
- •Critical point is now a critical range:  $S_a \sim N^{\delta}, \delta = \delta(a, p) \left( p_c$
- •Below and above range, behavior is  $P_{\infty}$  similar to regular percolation:

$$S_a \sim \log N \left( p < p_c \right)$$
$$S_a \sim N \left( p > \widetilde{p}_c \right)$$



## Theory of model networks: Erdős-Rényi



- Developed in the 1960's by Erdős and Rényi. (Publications of the Mathematical Institute of the Hungarian Academy of Sciences, 1960).
- N nodes and each pair connected with probability  $\phi$ .
- Define k as the degree (number of links of a node), and  $\langle k \rangle$ is average number not in the network.

Construction degree of j,  $k_i=3$ 





• Distribution of degree is Poisson-like (exponential)  $P(k) = e^{-\langle k \rangle} \frac{\langle k \rangle^k}{k}$ 



Outline of scaling theory for Limited Path Percolation Example: Erdős-Rényi



- Before percolation, typical path length  $l \sim \log N/\log \langle k \rangle$
- After percolation, local structure is tree-like, with branching factor  $\kappa = p \langle k \rangle + 1$
- Tree approx.  $\Rightarrow S_a \sim (\kappa 1)^l = (p < k >)^{a \log N/\log < k >} = N^{\delta}$
- Scaling exponent  $0 \le \delta \equiv a(1 + \log p/\log < k) \le 1$
- $\delta \leq 1$  be dist  $S_a$  cannot exceed N
- Solving  $\delta = 1 \implies \tilde{p}_{q} \log \langle k \rangle^{(1-a)/a}$
- Usual percolation recovered with  $a \rightarrow \infty$ :  $\widetilde{p}_c \xrightarrow{a \rightarrow \infty} p_c = \langle k \rangle^{-1}$



Limited path percolation predicts a larger communication threshold.





#### Some basic network properties

#### Erdős-Rényi networks

Scale-free networks

•Narrow range of typical degree

$$\langle k \rangle - \sqrt{\langle k \rangle} \le k \le \langle k \rangle + \sqrt{\langle k \rangle}$$

•Small diameter

 $D \sim \ln N$ 

•Wide range of typical degree

$$k_{\min} \leq k \leq k_{\min} N^{1/(\lambda-1)}$$

 $(k_{\min} \text{ is minimum degree})$ 

•Small or ultra-small diameter

 $D \sim \ln(\ln N) [2 < \lambda < 3]$  $D \sim \ln N [\lambda > 3]$ 

Scaling theory for limited path percolation on scale-free networks
For λ>3:

$$S_a \sim N^{a \left[1 + \log p / \log (\kappa_o - 1)\right]}$$
$$\widetilde{p}_c = (\kappa_o - 1)^{(1-a)/a}$$

• For  $2 < \lambda < 3$ :

Tree approximation invalid. Networks are ultra-small:  $l \sim \log \log N / |\log(\lambda - 2)|$   $l' \sim \log \log P_{\infty} N / |\log(\lambda - 2)|$ Therefore:

$$a = \frac{l'}{l} \sim \frac{\log \log P_{\infty} N}{\log \log N} \xrightarrow[N \to \infty]{} 1$$



## Results for $S_a \sim N^{\delta}$ Scale-free





## Targeted attacks on scale-free networks

- Scale-free networks have sensitive nodes (hubs) with large k.
- Examples: Airline hubs, central communication nodes, disease super-spreaders.

### Model for targeted percolation



*p*: fraction of lowest degree nodes present.
In targeted percolation (no length)

• In targeted percolation (no length restriction)  $p_c$  is large:  $p_c=1 \ (\lambda \rightarrow 2)$  $p_c$  close to 1  $(\lambda > 2)$ 

Network falls apart with few node removals

Question: What happens for limited path percolation?



Scaling theory for limited path targeted percolation on scale-free networks

• For 
$$\lambda > 3$$
:  
 $S_a \sim N^{a \log(\kappa - 1)/\log(\kappa_o - 1)}$   
 $\widetilde{p}_c = \widetilde{p}_c(a, \kappa, \kappa_o)$ 

• For  $2 < \lambda < 3$ :

Tree approximation valid again after percolation:

$$S_a \sim (\log N)^{2a\log(\kappa-1)/|\log(\lambda-2)|}$$

Any finite *a* fails to produce transition to linear phase:  $\widetilde{p}_c = 1$ 





## Results for $S_a \sim N^{\delta}$ Scale-free targeted removal



Differences in Limited Path Percolation due to network structure and removal method at  $p_c \leq p \leq \tilde{p}_c$ Random removal



### Scaling function for $S_a$

•For Erdös-Rényi, and scale-free  $\lambda > 3$  with random and targeted removal, there are two phases above and below  $\tilde{p}_c$ 

•Therefore:

$$S_a \sim c(p) N^{\delta} f\left(\frac{P_{\infty}N}{c(p)N^{\delta}}\right) f(x) \sim \begin{cases} x, & x <<1\\ cnst., & x >>1 \end{cases}$$
$$c(p) \equiv c_o \left[p (\kappa_o -1) + 1\right] / \left[p(\kappa_o -1) - 1\right]$$
Two limits:

i) 
$$S_a \sim c(p) N^{\delta}$$
  $(p_c$ 

ii)  $S_a \sim P_\infty N$   $(\widetilde{p}_c$ 

### Results for scaling of $S_a$



### Conclusions

• We define a new percolation model which takes into account the length restriction of useful paths.

• This model is important in real-world applications such as epidemics, data transfer, and transportation.

• We find a new percolation transition at  $\tilde{p}_c = (\kappa_0 - 1)^{(1-a)/a} > p_c$ which implies when lengths are constrained, more connections are necessary to percolate. Transition preserves path length scaling.

• We encounter two typical phases: i) power-law with  $S_a \sim N^{\delta}$ , and ii) a linear phase  $S_a \sim N$ .

### Conclusions

- We define a new percolation model which takes into account the length restriction of useful paths.
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• We encounter two typical phases: i) power-law with  $S_a \sim N^{\delta}$ , and ii) a linear phase  $S_a \sim N$ .

• Few models of percolation exist. Our model is an innovative new approach to percolation with great opportunities for research.



### Timeline of percolation theory Tree percolation



Gelation or how the egg hardens: Flory(1941) and Stockmayer(1943). Percolation Flow through a random medium: Broadbent and Hammersley(1957). Directed percolation Steady state chemical reactions: Schlögl (1972). Bootstrap percolation Ferromagnets: Pollak and Reiss (1975). Invasion percolation Displacement of fluid by another: Wilkinson and Willemsen (1983). Limited path percolation Communications and epidemics: López et al. (2007).

Molloy-Reed Algorithm for scale-free Networks Create network with pre-specified degree distribution P(k)

Example:

1) Generate set of nodes with pre-specified degree distribution form  $P(k) \sim k^{-\lambda}$ 

Degree: 2 3 5 2 3 3

2) Make  $k_i$  copies of node *i*:

3) Randomly pair copies excluding self-loops and double connections:

4) Connect network:

### Theory: Properties of scale-free networks

•Network size with branching factor  $\kappa_o$ :

~ ( $\kappa_o$ -1)<sup>l</sup> ( $\lambda$ >3); variable (2< $\lambda$ <3)

•Branching factor:

$$\kappa_o = \langle k^2 \rangle / \langle k \rangle = \text{cons.} (\lambda > 3); \text{ incres.} (2 < \lambda < 3)$$

•Typical distance *l* between nodes:  $l \sim \frac{\log N}{\log (\kappa_o - 1)} (\lambda > 3); \frac{\log \log N}{|\log (\lambda - 2)|} (2 < \lambda < 3)$ 

•Percolation thresholds:

$$p_c = (\kappa_o - 1)^{-1} (\lambda > 3); 0 (2 < \lambda < 3)$$

•Nodes connected at  $p=p_c$ :  $S \sim N^{(\lambda-3)/(\lambda-1)} (\lambda>3); N(2<\lambda<3)$ 

•Branching factor at occupation *p*:

$$\kappa - 1 = p(\kappa_o - 1) \text{ for } (\lambda > 3)$$

#### Summary of theoretical results

# Erdös-Rényi $\widetilde{p}_{c} = \langle k \rangle^{(1-a)/a}, \ S_{a} \sim N^{\delta}, \ \delta = a \left( 1 + \frac{\log p}{\log \langle k \rangle} \right)$

Scale-free ( $\lambda > 3$ )

$$\widetilde{p}_{c} = (\kappa_{o} - 1)^{(1-a)/a}, \ S_{a} \sim N^{\delta}, \ \delta = a \left(1 + \frac{\log p}{\log(\kappa_{o} - 1)}\right)$$

Scale-free (2 $<\lambda<3$ )

$$\widetilde{p}_c = 0, S_a \sim N$$

#### Summary of theoretical results

Targeted removal on scale-free networks

$$\widetilde{p}_{c} = \widetilde{p}_{c}(a, \kappa, \kappa_{o}), S_{a} \sim N^{\delta}, \delta = a \frac{\log(\kappa - 1)}{\log(\kappa_{o} - 1)}$$

2<*λ*<3

 $\lambda > 3$ 

$$\widetilde{p}_{c} = 1, S_{a} \sim (\log N)^{\delta}, \delta = 2a \frac{\log(\kappa - 1)}{\left|\log(\lambda - 2)\right|}$$

## Motivation: Where else does percolation fail?

• Communications such as data packet routing:

• Infectious diseases: Flu decays over time/season. Increase of immunity in population. Message route

- Transportation: Long commute times prohibitive. Communication problems
- Other **Reporting extresisensslike path path isoshold**erations. Long paths compound error + reduce performance + security

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• Communications such as data packet routing:

Infectious diseases:
 Flu decays over time/season.
 Increase of immunity in population.
 Message route
 Transportation:

Long commute times prohibitive. require data rerouting

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Erdős-Rényi Network





**Scale-free Network** 













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- •Analytical scaling theory
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