

# **The Geometric Protean Model for On-Line Social Networks**

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- Introduce a new geometric, rank-based model for the link structure of on-line social networks (OSNs)
- With high probability, the geo-protean (GEO-P) model generates graph satisfying many observed properties of OSNs

# Large-scale

- Complex network with large number of nodes
- Some nodes have very high degree.

# Small world property and shrinking distance

- Demands a low diameter of  $O(\log n)$ , and a higher clustering coefficient than found in a binomial random graph with the same number of nodes and average degree

# Power law degree distribution

- In a graph  $G$  of order  $n$ , let  $N_k$  be the number of nodes of degree  $k$ . The degree distribution of  $G$  follows a power law if  $N_k$  is proportional to  $k^{-b}$ , for a fixed exponent  $b > 2$ .

# Bad spectral expansion

- Social network often organize into separate clusters in which the intra-cluster links are significantly higher than the number of inter-cluster links

# The GEO-P Model for OSNs

- Produces a sequence  $(G_t: t \geq 0)$  of undirected graph on  $n$  nodes, where  $t$  denotes time.  $G_t = (V_t, E_t)$ .
- 4 parameters: the attachment strength  $\alpha \in (0, 1)$ , the density parameter  $\beta \in (0, 1 - \alpha)$ , the dimension  $m \in \mathbb{N}$ , and the link probability  $p \in (0, 1]$
- Each nodes  $v \in V_t$  has rank  $r(v, t) \in [n]$
- The rank function  $r(\cdot, t): V_t \rightarrow [n]$  is a bijection for all  $t$ , so every node has a unique rank.
- Highest ranked node has rank equal to 1, the lowest ranked node has rank  $n$ .
- For each  $v \in V_{t-1} \cap V_t$ ,  $r(v, t) = r(v, t-1) + \delta - \gamma$   
where  $\delta = 1$  if  $r(v, t-1) > R_t$  and 0 otherwise, and  $\gamma = 1$  if the rank of the node deleted in step  $t$  is smaller than  $r(v, t-1)$ , and 0 otherwise.

-Let  $S$  be the unit hypercube in  $\mathbb{R}^m$ , with the torus metric  $d(\cdot, \cdot)$  derived from the  $L^\infty$  metric. In particular, for any two points  $x$  and  $y$  in  $\mathbb{R}^m$ ,  
$$d(x, y) = \min\{\|x-y+u\|^\infty : u \in \{-1,0,1\}^m\}.$$

-To initialize the model, let  $G_0 = (V_0, E_0)$  be any graph on  $n$  nodes that are chosen from  $S$ . We define the influence region of node  $v$  at time  $t \geq 0$ , written  $R(v, t)$ , to be the ball around  $v$  with volume  
$$|R(v, t)| = r(v, t) - \alpha n - \beta$$



For  $t \geq 1$ , we form  $G_t$  from  $G_{t-1}$  according to the following rules.

1. Add a new node  $v$  that is chosen uniformly at random from  $S$ . Next, independently, for each node  $u \in V_{t-1}$  such that  $v \in R(u, t-1)$ , an edge  $vu$  is created with probability  $p$ . Note that the probability that  $u$  receives an edge is equal to  $p r(u, t-1)^{-\alpha n - \beta}$ .

The negative exponent  $(-\alpha)$  guarantees that nodes with higher ranks ( $r(u, t-1)$  close to 1) are more likely to receive new edges than lower ranks

2. Choose uniformly at random a node  $u \in V_{t-1}$ , delete  $u$  and all edges incident to  $u$ .

3. Update the ranking function  $r(\cdot, t) : V_t \rightarrow [n]$ .

The following theorem demonstrates that the geo-protean model generates power law graphs with exponent

$$b = 1 + 1/\alpha$$

**Theorem 1.** *Let  $\alpha \in (0, 1)$ ,  $\beta \in (0, 1 - \alpha)$ ,  $m \in \mathbb{N}$ ,  $p \in (0, 1]$ , and*

$$n^{1-\alpha-\beta} \log^{1/2} n \leq k \leq n^{1-\alpha/2-\beta} \log^{-2\alpha-1} n.$$

*Then wep  $GEO-P(\alpha, \beta, m, p)$  satisfies*

$$N_{\geq k} = \left(1 + O(\log^{-1/3} n)\right) \frac{\alpha}{\alpha + 1} p^{1/\alpha} n^{(1-\beta)/\alpha} k^{-1/\alpha}.$$

For a graph  $G = (V, E)$  of order  $n$ , define the *average degree of  $G$*  by  $d = \frac{2|E|}{n}$ . Our next results shows that geo-protean graphs are dense.

**Theorem 2.** *Wep the average degree of  $GEO-P(\alpha, \beta, m, p)$  is*

$$d = (1 + o(1)) \frac{p}{1 - \alpha} n^{1-\alpha-\beta}. \quad (2)$$

Note that the average degree tends to infinity with  $n$ ; that is, the model generates graphs satisfying a *densification power law*. In [17], densification power laws were reported in several real-world networks such as the physics citation graph and the internet graph at the level of autonomous systems.

Our next result describes the diameter of graphs sampled from the GEO-P model. While the diameter is not shrinking, it can be made constant by allowing the dimension to grow as a logarithmic function of  $n$ .

**Theorem 3.** *Let  $\alpha \in (0, 1)$ ,  $\beta \in (0, 1 - \alpha)$ ,  $m \in \mathbb{N}$ , and  $p \in (0, 1]$ . Then *wep* the diameter of  $\text{GEO-P}(\alpha, \beta, m, p)$  is*

$$O(n^{\frac{\beta}{(1-\alpha)m}} \log^{\frac{2\alpha}{(1-\alpha)m}} n). \quad (3)$$

We note that in a geometric model where regions of influence have constant volume and possessing the same average degree as the geo-protean model, the diameter is  $\Theta(n^{\frac{\alpha+\beta}{m}})$ . This is a larger diameter than in the GEO-P model. If  $m = C \log n$ , for some constant  $C > 0$ , then *wep* we obtain a diameter bounded by a constant. We conjecture that *wep* the diameter is of order  $n^{\frac{\beta}{(1-\alpha)m} + o(1)}$ . In the full version of the paper, we prove that *wep* the GEO-P model generates graph with constant clustering coefficient.

The normalized Laplacian of a graph relates to important graph properties; see [7]. Let  $A$  denote the adjacency matrix and  $D$  denote the diagonal degree matrix of a graph  $G$ . Then the normalized Laplacian of  $G$  is  $\mathcal{L} = I - D^{-1/2}AD^{-1/2}$ . Let  $0 = \lambda_0 \leq \lambda_1 \leq \dots \leq \lambda_{n-1} \leq 2$  denote the eigenvalues of  $\mathcal{L}$ . The *spectral gap* of the normalized Laplacian is

$$\lambda = \max\{|\lambda_1 - 1|, |\lambda_{n-1} - 1|\}.$$

A spectral gap bounded away from 0 is an indication of bad expansion properties, which are characteristic of OSNs (see property (iv) in the introduction). The next theorem represents a drastic departure from the good expansion found in binomial random graphs, where  $\lambda = o(1)$  [7,8].

**Theorem 4.** *Let  $\alpha \in (0, 1)$ ,  $\beta \in (0, 1 - \alpha)$ ,  $m \in \mathbb{N}$ , and  $p \in (0, 1]$ . Let  $\lambda(n)$  be the spectral gap of the normalized Laplacian of  $GEO-P(\alpha, \beta, m, p)$ . Then wep*

1. *If  $m = m(n) = o(\log n)$ , then  $\lambda(n) = 1 + o(1)$ .*
2. *If  $m = m(n) = C \log n$  for some  $C > 0$ , then*

$$\lambda(n) \geq 1 - \exp\left(-\frac{\alpha + \beta}{C}\right).$$

# Dimension of OSNs

-Given an OSN, we may estimate the corresponding dimension parameter  $m$  if we assume the GEO-P model.

-In particular, if we know the order  $n$ , power law exponent  $b$ , average degree  $d$ , and diameter  $D$  of an OSN, then we can calculate  $m$  using our theoretical results.

-An estimate for  $m$  is given by:

$$m = \frac{1}{D^*} \left( 1 - \left( \frac{b-1}{b-2} \right) d^* \right).$$

The following chart gives the predicted dimension  $m$  for each network. We round  $m$  up to the nearest integer

$n$  is total number of users for Cyworld, Flickr, Twitter, and YouTube

$b$  is the power law exponent for the in-degree distribution

<b>Parameter</b>	<b>OSN</b>			
	Cyworld	Flickr	Twitter	YouTube
$n$	$2.4 \times 10^7$	$3.2 \times 10^7$	$7.5 \times 10^7$	$3 \times 10^8$
$b$	5	2.78	2.4	2.99
$d^*$	0.22	0.17	0.17	0.1
$D^*$	0.11	0.19	0.1	0.16
$m$	7	4	5	6