



DEEP  
LEARNING  
INSTITUTE



DLI Accelerated Data Science Teaching Kit

# Lecture 17.3 - Centralities: Degree, Betweenness, Clustering Coefficient



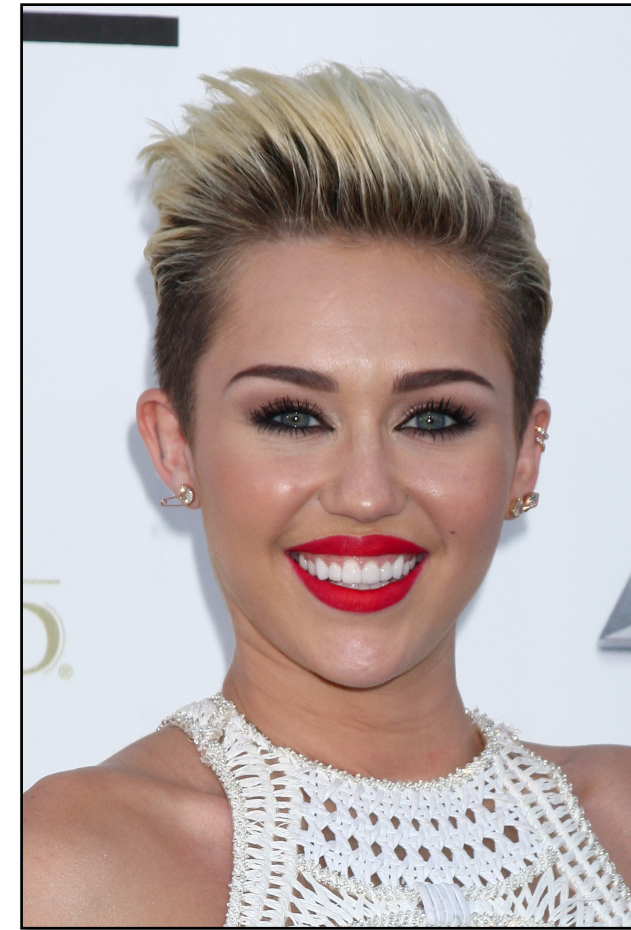
The Accelerated Data Science Teaching Kit is licensed by NVIDIA, Georgia Institute of Technology, and Prairie View A&M University under the [Creative Commons Attribution-NonCommercial 4.0 International License](https://creativecommons.org/licenses/by-nc/4.0/).

**Centrality**  
= “Importance”

# Why Node Centrality?

What can we do if we can rank all the nodes in a graph (e.g., Facebook, LinkedIn, Twitter)?

- Find **celebrities** or influential people in a social network (Twitter)
- Find “**gatekeepers**” who connect communities (headhunters love to find them on LinkedIn)



# Why Node Centrality?

Helps **graph analysis, visualization, understanding**, e.g.,

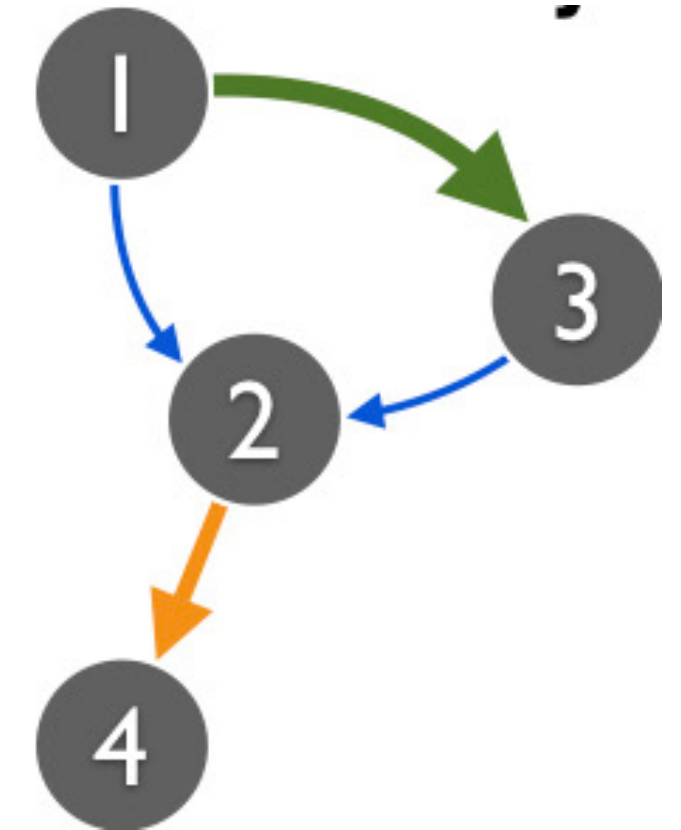
- Let us rank nodes, group or study them by centrality
- Only show subgraph formed by the **top 100 nodes**, out of the millions in the full graph
- Similar to google search results (ranked, and they only show you 10 per page)
- Most graph analysis packages already have centrality algorithms implemented. Use them!

Can also compute edge centrality. Here we focus on node centrality.

# Degree Centrality (Easiest)

**Degree = number of neighbors**

- For directed graphs
  - In degree = No. of incoming edges
  - Out degree = No. of outgoing edges
- For undirected graphs, only degree is defined.
- Algorithms?
  - Sequential scan through **edge list**
  - What about for a **graph stored in SQLite**?



1, 2  
1, 3  
2, 4  
3, 2

# Computing Degrees using SQL

Recall simplest way to store a graph in SQLite:

```
edges(source_id, target_id)
```

1, 2

1, 3

1. If slow, first create index for each column

2, 4

3, 2

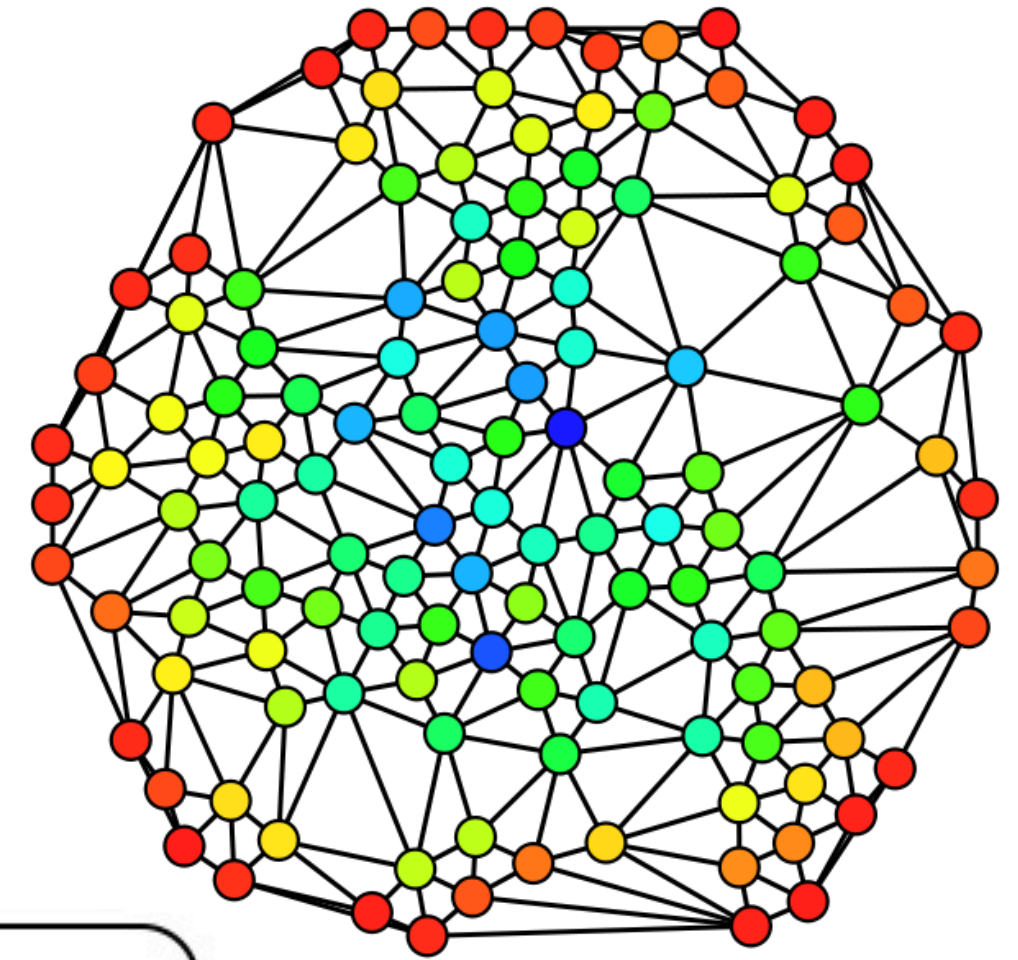
2. Use **group by** statement to find **out degrees**

```
select count(*) from edges group by source_id;
```



# Betweenness Centrality

High betweenness = “gatekeeper”



Betweenness of a node  $v$

$$= \sum_{s \neq v \neq t \in V} \frac{\sigma_{st}(v)}{\sigma_{st}}$$

Number of shortest paths between  $s$  and  $t$  that **goes through  $v$**

Number of shortest paths between  $s$  and  $t$

= how often a node serves as the “bridge” that connects two other nodes.



# (Local) Clustering Coefficient

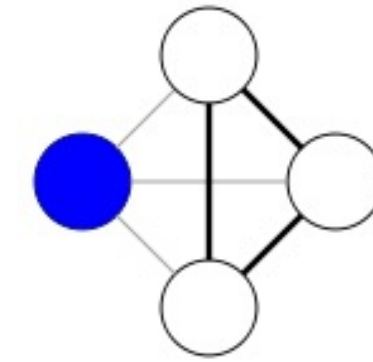
A node's clustering coefficient is a measure of **how close the node's neighbors are from forming a clique**.

1 = Neighbors form a clique

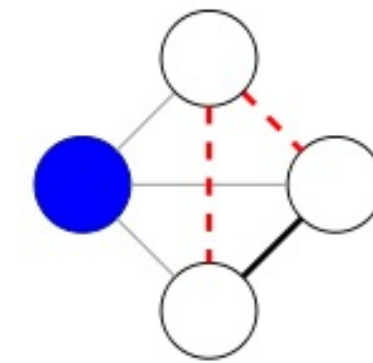
0 = No edges among neighbors

(Assuming undirected graph)

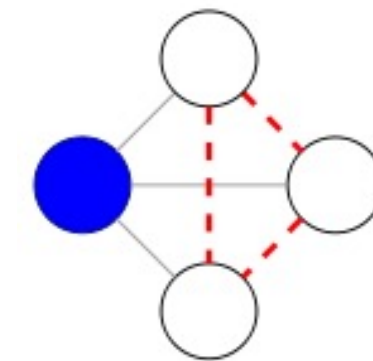
“Local” means it's for a node; can also compute a graph's “global” coefficient



$$c = 1$$



$$c = 1/3$$



$$c = 0$$

# Computing Clustering Coefficients...

Requires **triangle counting**

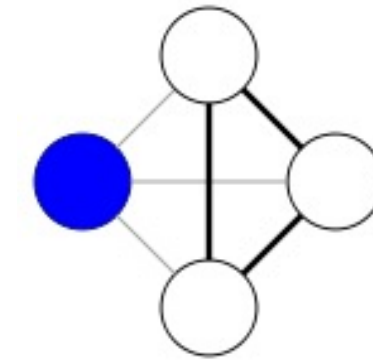
Real social networks have a lot of triangles

Friends of friends are friends

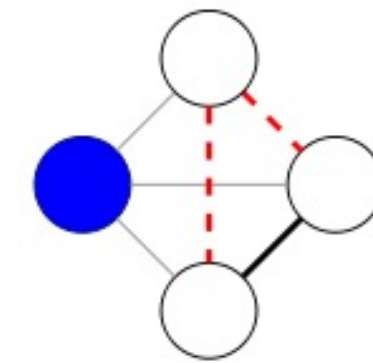
Triangles are **expensive** to compute

(neighborhood intersections; several approx. algos)

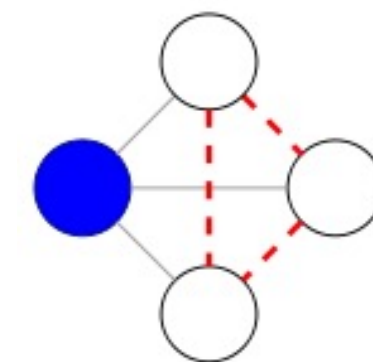
Can we do that quickly?



$$c = 1$$



$$c = 1/3$$



$$c = 0$$

Algorithm details:

Faster Clustering Coefficient Using Vertex Covers

[http://www.cc.gatech.edu/~ogreen3/\\_docs/2013VertexCoverClusteringCoefficients.pdf](http://www.cc.gatech.edu/~ogreen3/_docs/2013VertexCoverClusteringCoefficients.pdf)

# Super Fast Triangle Counting

[Tsourakakis ICDM 2008]



details

But: triangles are expensive to compute  
(3-way join; several approx. algos)

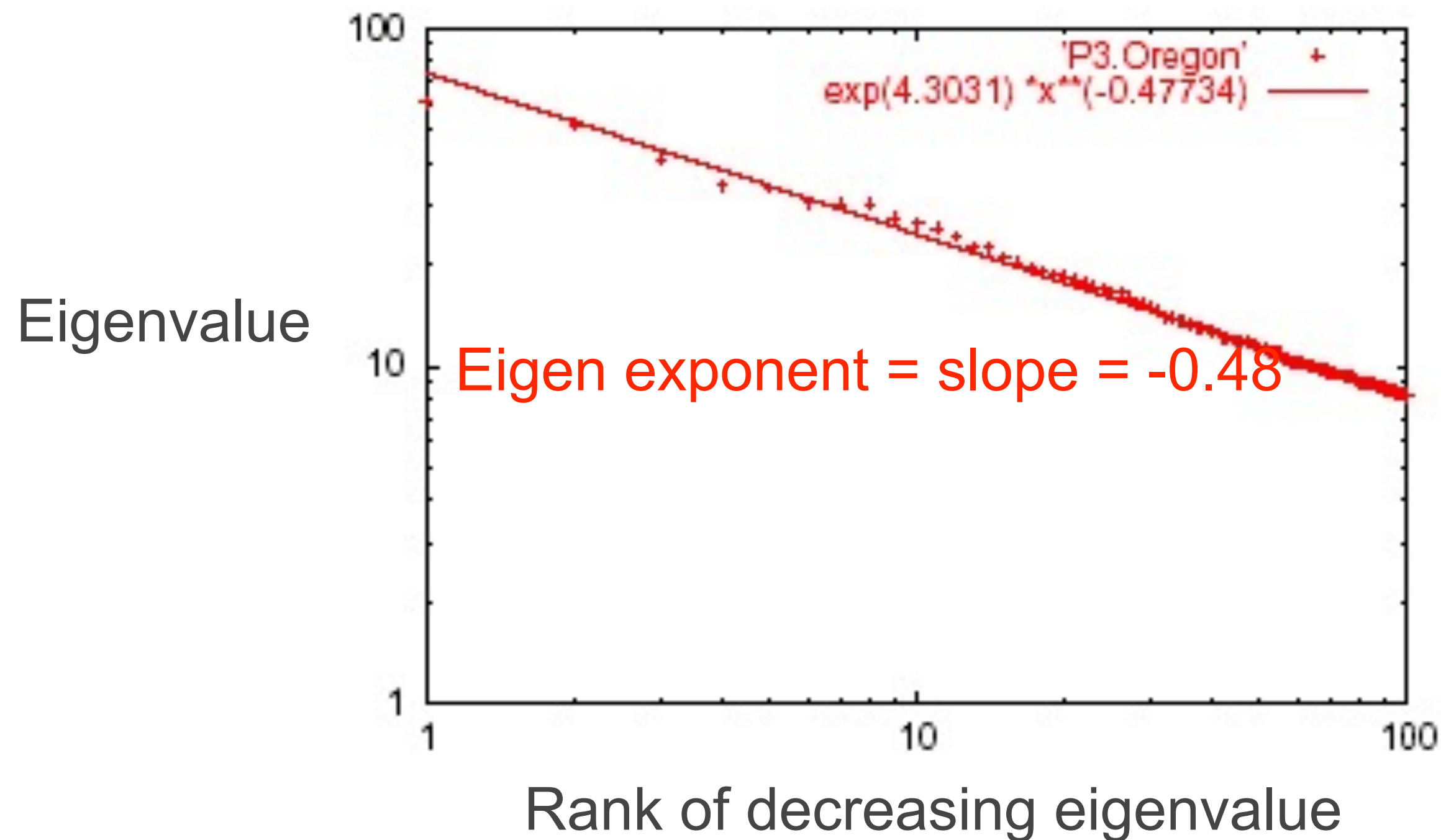
Q: Can we do that quickly?

A: Yes!

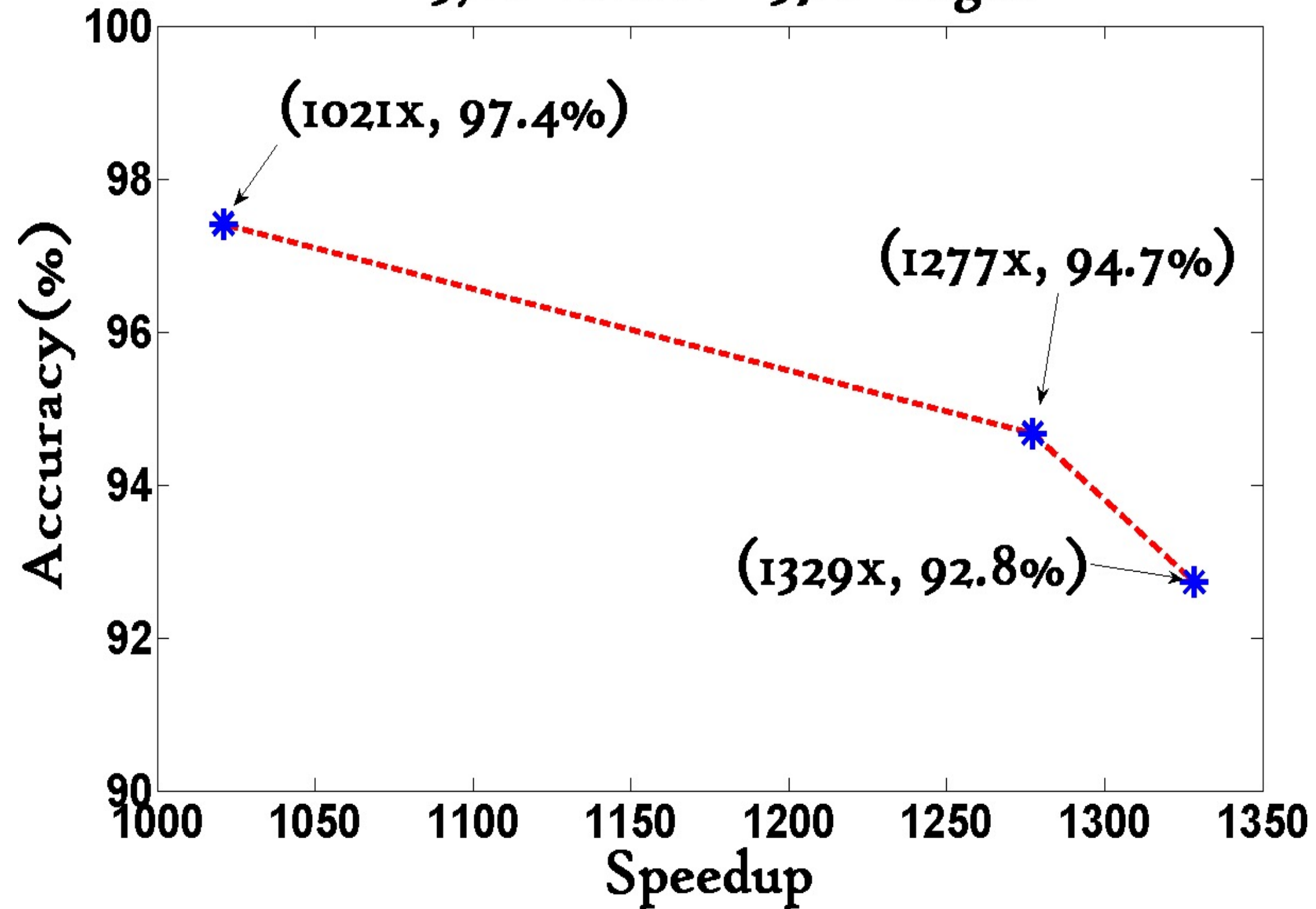
$$\text{\#triangles} = 1/6 \text{ Sum } ( \lambda_i^3 )$$

Because of skewness, we only need top few eigenvalues!

# Power Law in Eigenvalues



Wikipedia graph 2006-Nov-04  
 $\approx 3.1\text{M}$  nodes  $\approx 37\text{M}$  edges



1000x+ speed-up, >90% accuracy



# More Centrality Measures...

- Degree
- Betweenness
- Closeness, by computing
  - Shortest paths
  - “Proximity” (e.g., via random walks)
- Eigenvector
- ...



DEEP  
LEARNING  
INSTITUTE



DLI Accelerated Data Science Teaching Kit

# Thank You