Faculty Development Programme on Network Science: Foundation Of Social Network Analysis

Generative Models

Hands-on Session (Day 4)

Swagata Duari Kirti Jain

Real World Network: FACEBOOK

Download Facebook graph: https://snap.stanford.edu/data/ego-Facebook.html

Reading edge list data from file:

> el = read.csv ("facebook_combined.txt", sep = "", head=F)

Turning data frame(having edge list) into graph object:

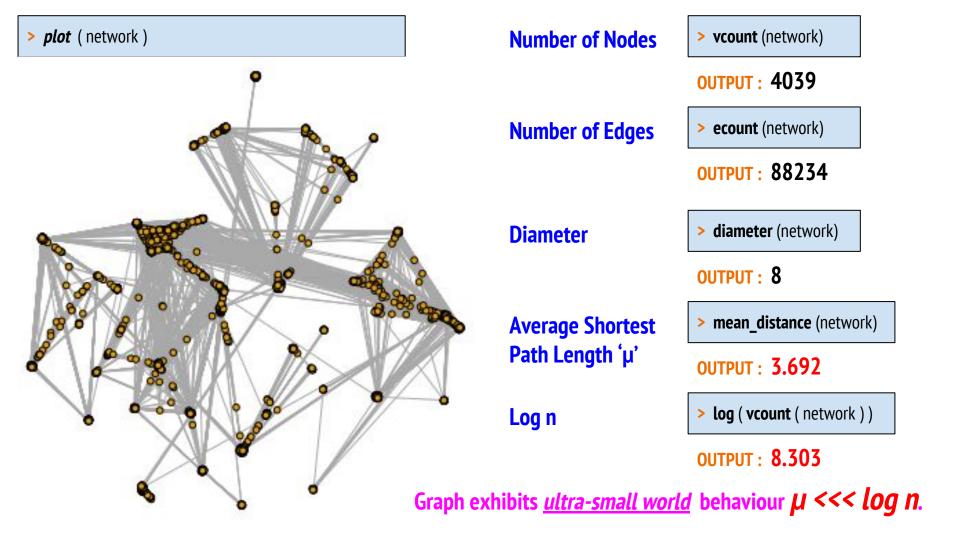
- network <- graph_from_data_frame (el , dir = FALSE)</p>
- > network

```
IGRAPH UN-- 4039 88234 --
```

- + attr: name (v/c)
- + edges (vertex names):

```
[1] 0--1 0--2 0--3 0--4 0--5 0--6 0--7 0--8 0--9 0--10 0--11 0--12 0--13 0--14 0--15 0--16 0--17 0--18 0--19 0--20 0--21 0--22 0--23 0--24 0--25 0--26 0--27 0--28 0--29 0--30 0--31 0--32 0--33 0--34 0--35 0--36 0--37 0--38 0--39 0--40 0--41 0--42 0--43 0--44 0--45 0--46 0--47 0--48 0--49 0--50 0--51 0--52 0--53 0--54 0--55 0--56 0--57 0--58 0--59 0--60 0--61 0--62 0--63 0--64 0--65 0--66 0--67 0--68 0--69 0--70 0--71 0--72 0--73 0--74 0--75 0--76 0--77 0--78 0--79 0--80 0--81 0--82 0--83 0--84 0--85 0--86 0--87 0--88 0--89 0--90 0--91 0--92 0--93 0--94 0--95 0--96 0--97 0--98 0--99
```

+ ... omitted several edges



Demonstration of Scale-Free Property of Graph

Get the degree of the graph:

> d <- degree (network , mode="all")

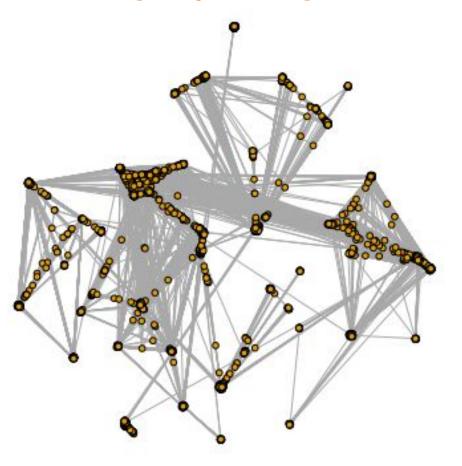
Fits power-law distribution to degree:

> power <- power.law.fit (d)

The power-law exponent:

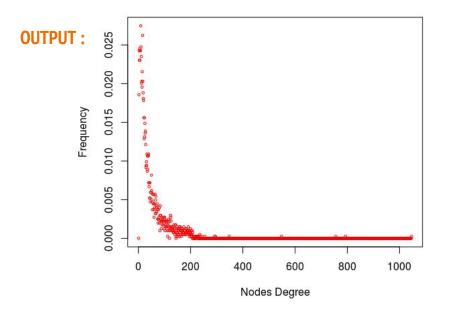
> power \$ alpha

OUTPUT: 2.510175



Degree Distribution of Facebook Network

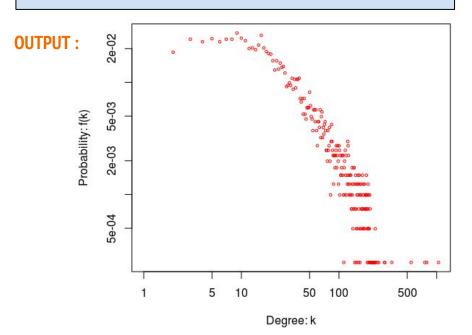
- > dd = **degree.distribution** (g, mode = "all", cumulative = FALSE)
- plot(dd, ylim=c(0,max(dd)),xlab="Nodes Degree", ylab="Frequency")



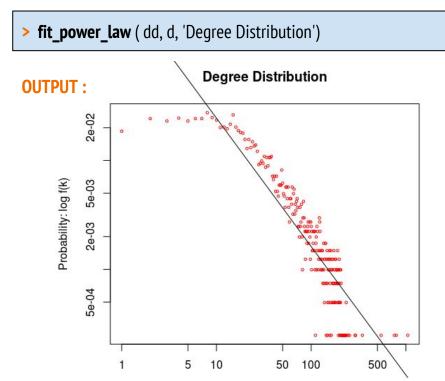
Plot shows that vast majority of nodes have very small degrees, whereas there are a few "hub" nodes that have high degrees.

Log-log plot of degree (k) versus degree distribution f(k):

plot (dd,log='xy', xlab="Degree: k", ylab="Probability: f(k)")



Use the newly defined function to plot the log-log plot:



Degree: log k

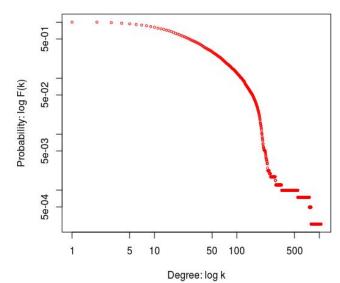
Note that there is **high level of noise for the higher degrees**, where frequency counts are the lowest. Solution: use the **Cumulative degree distribution F(k)**.

Cumulative degree distribution F(k):

- > cdd = degree.distribution(g, mode = "all", cumulative = TRUE)
- plot(cdd,log='xy', xlab="Degree: log k", ylab="Probability: logF(k)", main="Cumulative Degree Distribution")

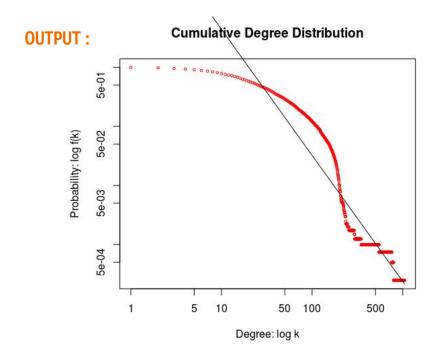
OUTPUT:

Cumulative Degree Distribution



Use the newly defined function to plot the log-log plot:

fit_power_law (cdd, d, 'Cumulative Degree Distribution')

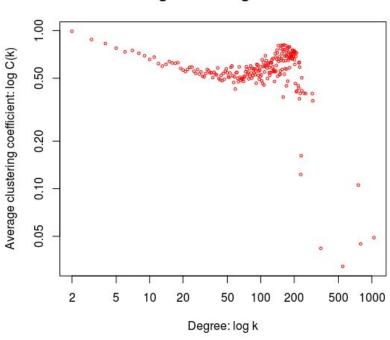


Clustering Coefficient of Facebook Network

> transitivity (network)

OUTPUT: 0.5191743

Average Clustering Coefficient

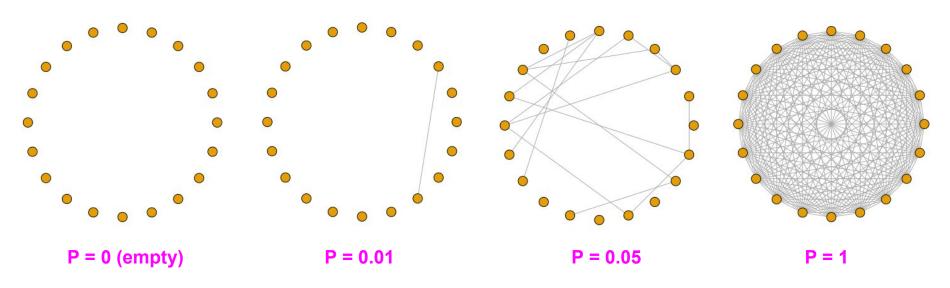


Erdos-Renyi (ER) Model

Generate random graphs according to the Erdos-Renyi model.

```
> erg <- erdos.renyi.game ( 20 , 0 , directed = FALSE ) #no edges
```

> plot (erg , layout = layout_in_circle, vertex.size=10)



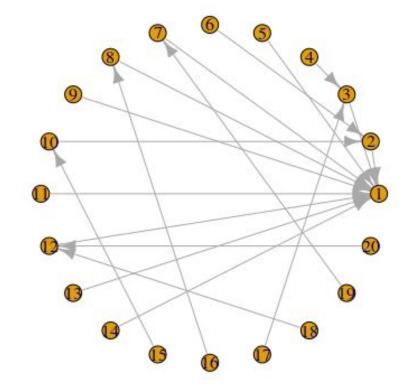
Increase the probability of connecting edges.

Barabasi-Albert preferential attachment Model (BA) Model

Generate scale-free graphs.

```
> bag <- sample_pa ( 20 )
```

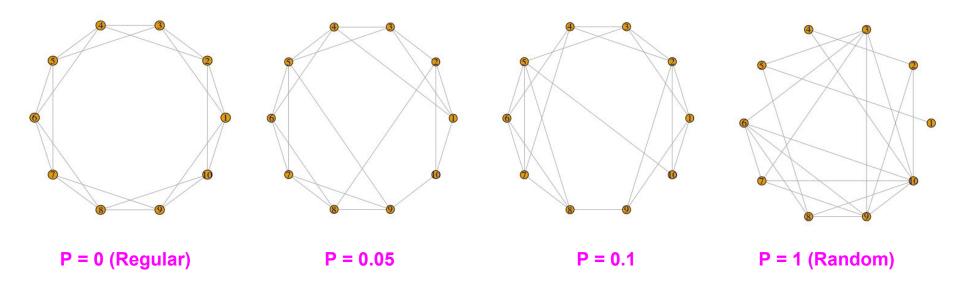
> plot (bag , layout = layout_in_circle, vertex.size=9)



Watts-Strogatz (WS) Model

Generate a graph according to the Watts-Strogatz network model.

- > wsg <- watts.strogatz.game (dim=1, size=10, nei=2, p=0)
- > plot (wsg , layout = layout_in_circle)



Increase the probability of rewiring edges. (more randomness)

Thankyou!