

GRAPH STRUCTURE OF THE WEB

Module 4

BOWTIE STRUCTURE OF THE WEB

- Represented the Web as a directed graph where the webpages are treated as vertices and hyperlinks are the edges and included the properties in this graph: diameter, degree distribution, connected components and macroscopic structure.
- The dark-web (part of the Web that is composed of webpages that are not directly accessible (even by Web browsers)) were disregarded.



POWER LAW

- The power law is a functional relationship between two quantities where a relative change in one quantity results in a proportional relative change in the other quantity, independent of the initial size of those quantities, i.e, one quantity varies as the power of the other.
- Hence the name power law.
- Power law distributions as defined on positive integers is the probability of the value i being proportional to $1/i_k$ for a small positive integer k .



BREADTH FIRST SEARCH (BFS) ALGORITHM

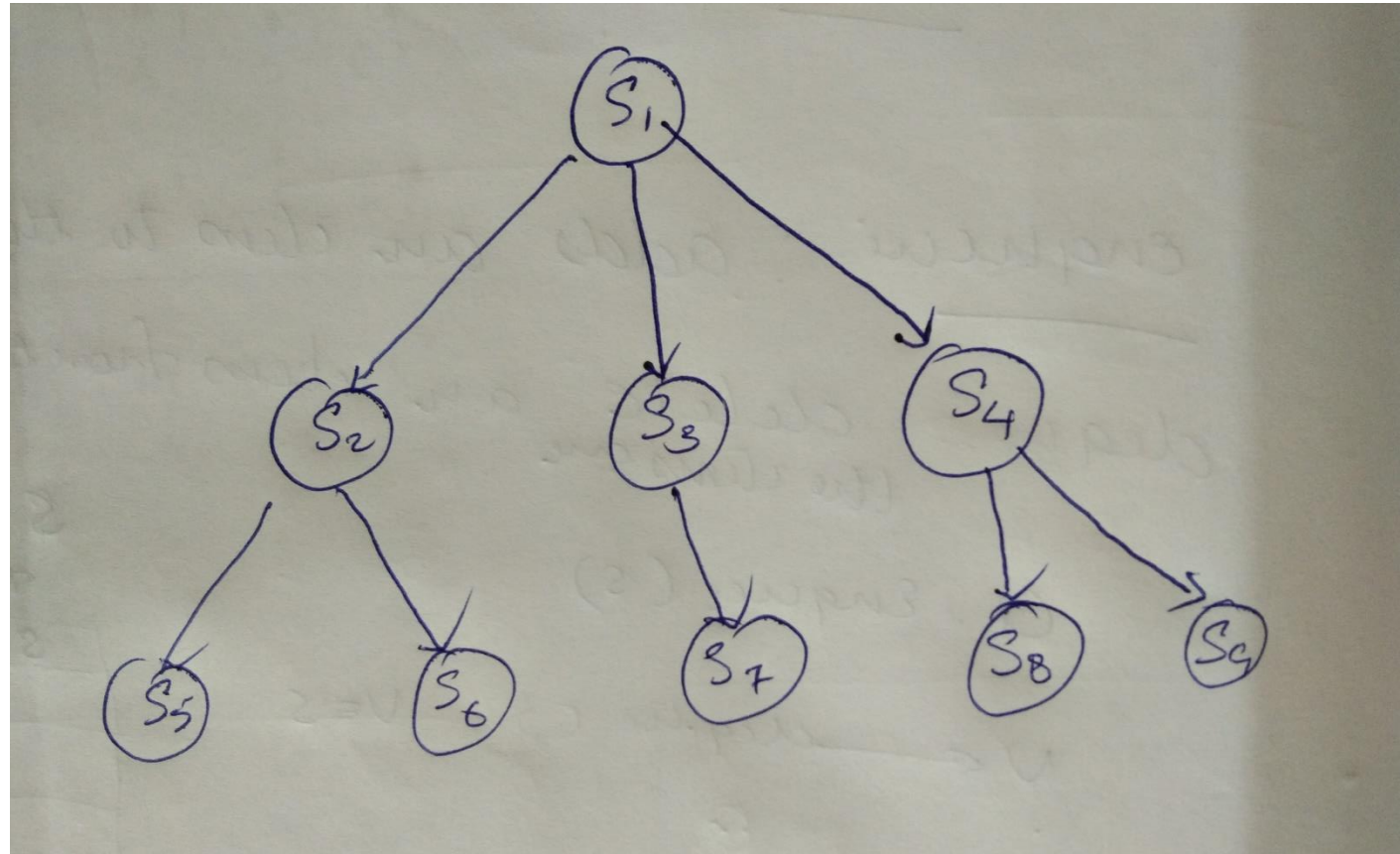
- The BFS algorithm takes as input a graph $G(V, E)$ and a source vertex s .
- The algorithm returns the set of vertices that the source vertex has a path to.
- The Web crawl proceeds in BFS, subject to various rules designed to avoid overloading, infinite paths, spam, time-outs etc.
- Each build is based on crawl data after further filtering and processing.
- Due to multiple starting points, it is possible for the resulting graph to have several connected components



Algorithm 1 BFS algorithm

```
1: procedure BFS( $G(V, E), s$ )
2:   Let reachable be an array
3:   Let  $Q$  be a queue
4:    $Q.enqueue(s)$ 
5:   reachable.add(s)
6:   while  $Q$  is not empty do
7:      $v \leftarrow Q.dequeue()$ 
8:     for each neighbour  $w$  of  $v$  in  $V$  do
9:       if  $w$  is not in reachable then
10:         $Q.enqueue(w)$ 
11:        reachable.add(w)
12:       end if
13:     end for
14:   end while
15:   return reachable
16: end procedure
```





IN AND OUT OF A VERTEX

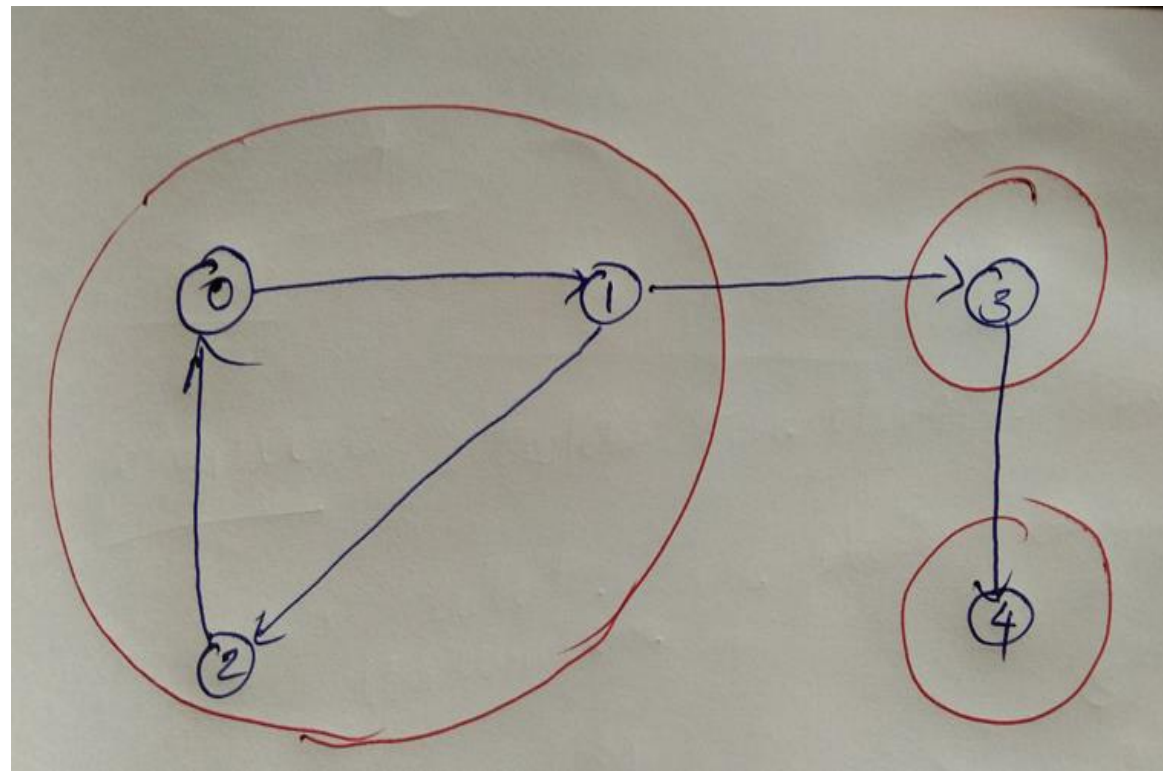
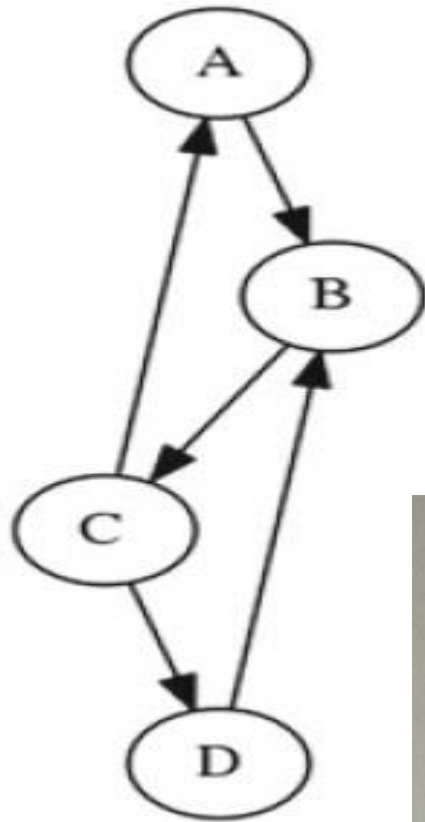
- For a graph $G(V, E)$, we define the *In* and *Out* of a vertex $v \in V$ as given
- $\text{In}(v) = \{w \in V \mid \text{there exists a path from } w \text{ to } v\}$
- $\text{Out}(v) = \{w \in V \mid \text{there exists a path from } v \text{ to } w\}$



STRONGLY CONNECTED COMPONENT

- A strongly connected component (SCC) is a set of vertices S such that it satisfies the following conditions:
 - Every pair of vertices in S has a path to one another.
 - There is no larger set containing S that satisfies this property
 - $\text{In}(v) = \text{Out}(v) \forall v \in V$





STRONGLY CONNECTED COMPONENTS (SCC) ALGORITHM

- The SCC algorithm takes a graph $G(V, E)$ as input and returns a list of all the SCCs in this graph as output.
- The SCC of G is computed using the equation
- $In(v) = Out(v) \forall v \in V$
- The function unique returns a list of all the distinct elements in the input list.
 - For instance, $unique([1,2,1,3,2,2])$ will return $[1,2,3]$.



Algorithm 2 In algorithm

```
1: procedure IN( $G(V, E), s$ )
2:   Let  $In$  be an array
3:   for each vertex  $v$  in  $V$  do
4:     if  $s$  in  $BFS(G(V, E), v)$  then
5:        $In.add(v)$ 
6:     end if
7:   end for
8:   return  $In$ 
9: end procedure
```

Algorithm 3 Out algorithm

```
1: procedure OUT( $G(V, E), s$ )
2:   Let  $Out$  be an array
3:    $Out \leftarrow BFS(G(V, E), s)$ 
4:   return  $Out$ 
5: end procedure
```



Algorithm 4 SCC algorithm

```
1: procedure SCC( $G(V, E)$ )
2:   Let  $SCC$  be an array
3:   for each vertex  $v$  in  $V$  do
4:      $SCC.add(In(G(V, E), v) \cap Out(G(V, E), v))$ 
5:   end for
6:   return  $unique(SCC)$ 
7: end procedure
```



WEAKLY CONNECTED COMPONENTS (WCC) ALGORITHM

- The WCC algorithm computes the list of all WCCs given a graph $G(V, E)$ as input.

Algorithm 5 WCC algorithm

```
1: procedure WCC( $G(V, E)$ )
2:   Let  $WCC$  be an array
3:   Let  $G'(V', E')$  be  $G(V, E)$  as an undirected graph
4:   for each vertex  $v$  in  $V'$  do
5:      $WCC.add(BFS(G'(V', E'), v))$ 
6:   end for
7:   return  $unique(WCC)$ 
8: end procedure
```



DEGREE DISTRIBUTION

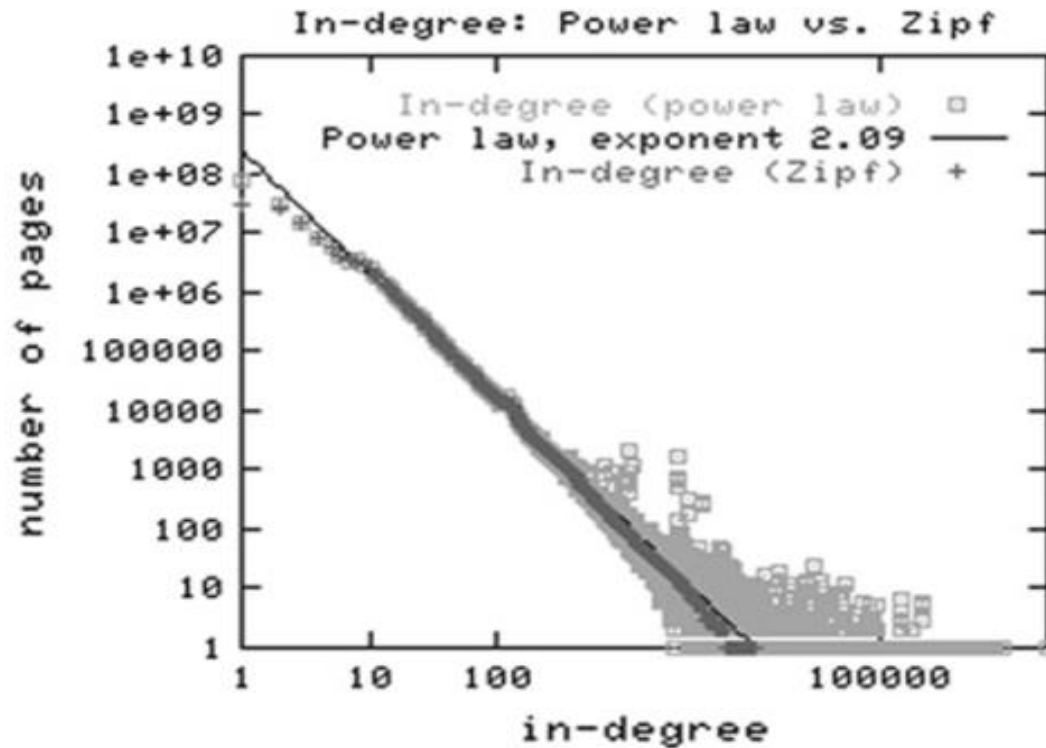
- The degree distribution of a graph $G(V, E)$, denoted by $P(k)$, is defined as the probability that a random chosen vertex has degree k .
- If $|V|_k$ denotes the number of vertices with degree k ,
- $$P(k) = |V|_k / |V|$$
- The degree distribution is plotted either as a histogram of k vs $P(k)$ or as a scree plot of k vs $|V|_k$



ZIPF'S LAW

- Zipf's law states that the frequency of occurrence of a certain value is inversely proportional to its rank in the frequency table.





In-degree distributions plotted as Power Law and Zipf's Law



- From the giant undirected component, the *DI SCONNECTED COMPONENT S* and the *SCC*.
- The 100 million vertices whose forward BFS traversals exploded correspond to either the *SCC* component or a component called *IN*.
- Since, *SCC* corresponds to 56 million vertices, this leaves with 44 million vertices ($\approx 22\%$) for *IN*.
- The 100 million vertices whose backward BFS traversals exploded correspond to either *SCC* or a component called *OUT*.
- The remaining vertices which are not accounted were called *TENDRILS*.



- These components altogether form the bowtie structure of the Web.
- The intuition behind these components is that IN corresponds to those webpages that can reach the SCC, but cannot be reached back from the IN component.
- This possibly contains new webpages that have not yet been discovered and linked to.
- The webpages that are accessible from the SCC, but do have a link back to this



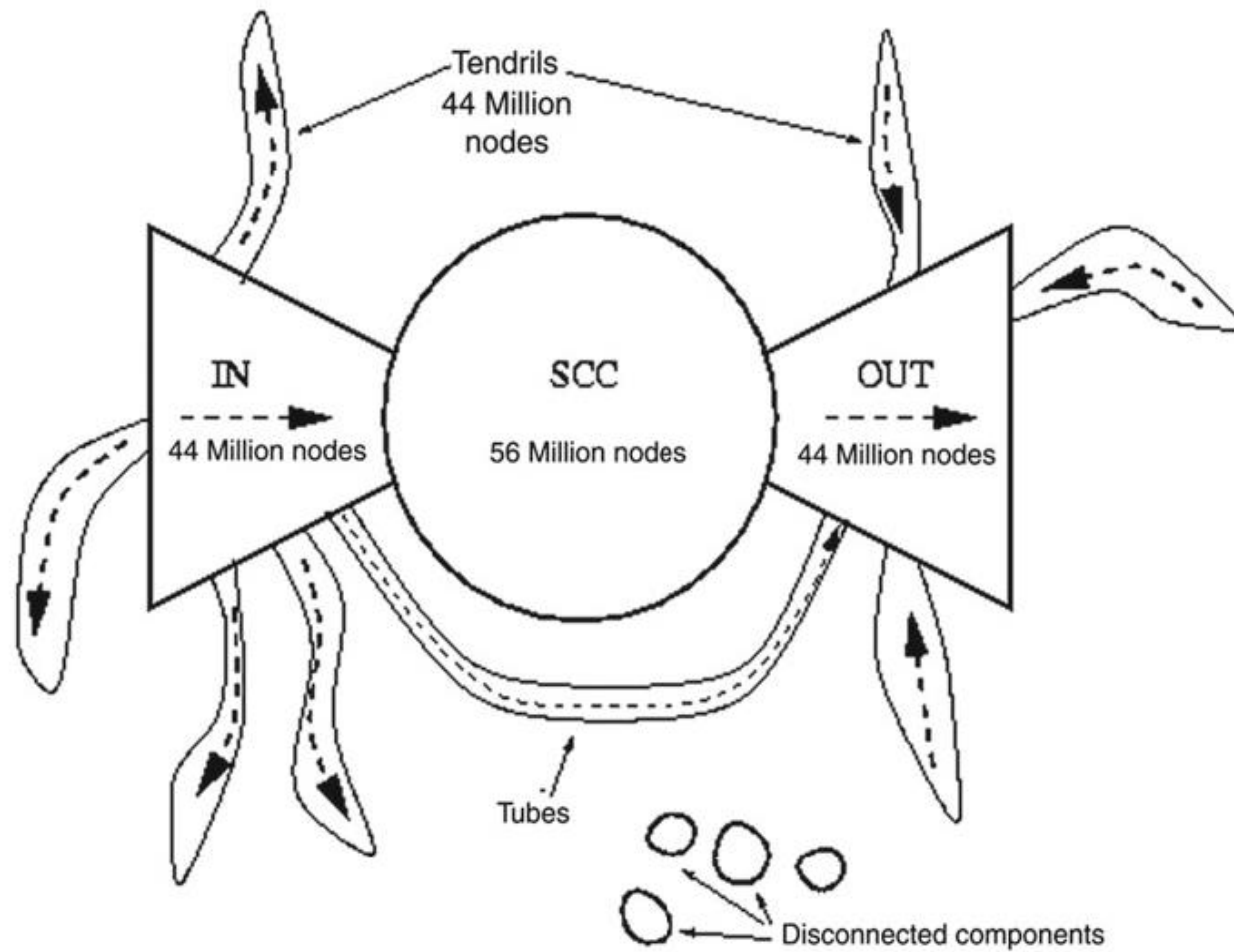


Fig. 2.11 Bowtie structure of the graph of the Web

THE POWER LAWS

- Power-laws are expressions of the form $y \propto x^a$, where a is a constant, x and y are the measures of interest, and \propto stands for “proportional to”.
- The exponents of the power-laws can be used to characterize graphs.



RANK EXPONENT R

- From the log-log plots of the out-degree d_v as a function of the rank r_v in the sequence of decreasing out-degree
 - The out-degree, d_v , of a node v is proportional to the rank of this node, r_v , to the power of a constant, R .

$$d_v \propto r_v^R$$



- If the nodes of the graph are sorted in decreasing order of out-degree, then the rank exponent, R is defined to be *the slope of the node versus the rank of the nodes* in the log-log scale.
- The out-degree, d_v , of a node v is a function of the rank of this node, r_v , and the rank exponent, R ,

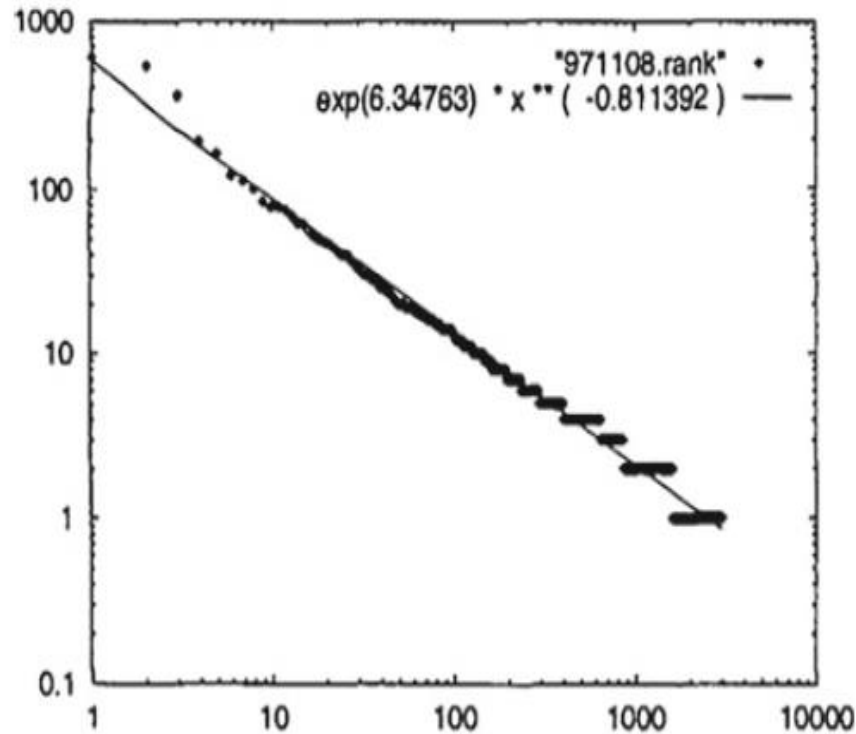
$$d_v = \frac{1}{N^R} r_v^R$$



- The number of edges, $|E|$, of a graph can be estimated as a function of the number of nodes, $|V|$, and the rank exponent, R ,

$$|E| = \frac{1}{2(R+1)} \left(1 - \frac{1}{N^{R+1}} \right)^{|V|}$$





Log-log plot of the out-degree d_v as a function of the rank r_v in the sequence of decreasing out-degree for Int-11-97



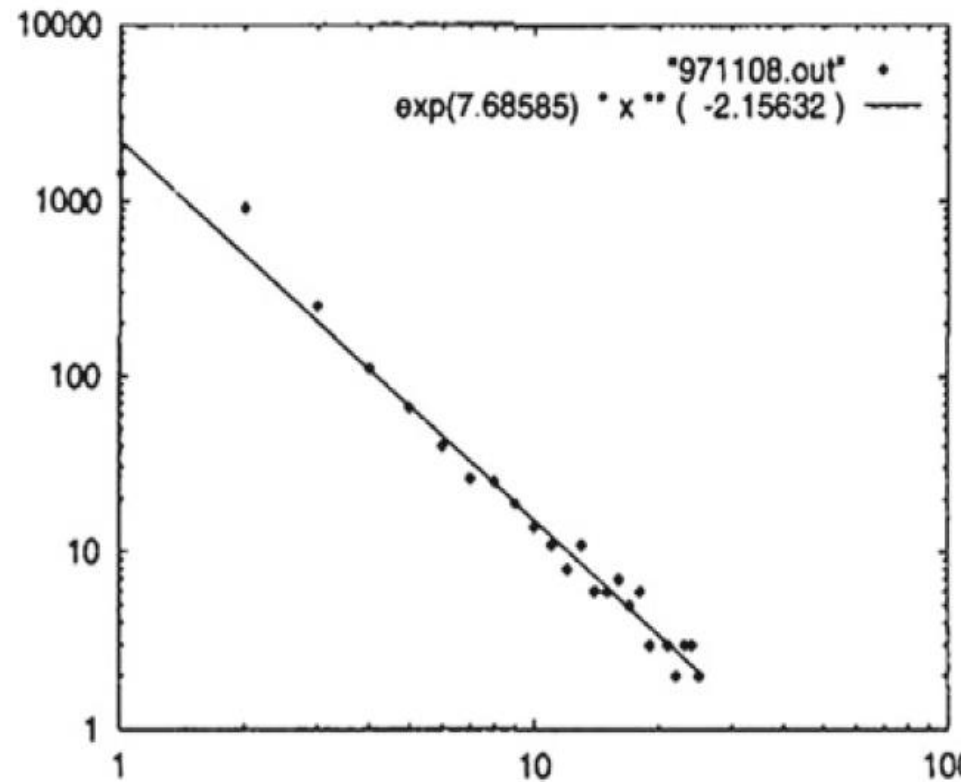
OUT-DEGREE EXPONENT O

- The log-log plots of the frequency f_d as a function of the out-degree d
 - The frequency, f_d , of an out-degree, d , is proportional to the out-degree to the power of a constant, O .

$$f_d \propto d^O$$

- The out-degree exponent, O , is defined to be *the slope of the plot of the frequency of the out-degrees versus the out-degrees in log-log scale.*





Log-log plot of frequency f_d versus the out degree for Int-11-97



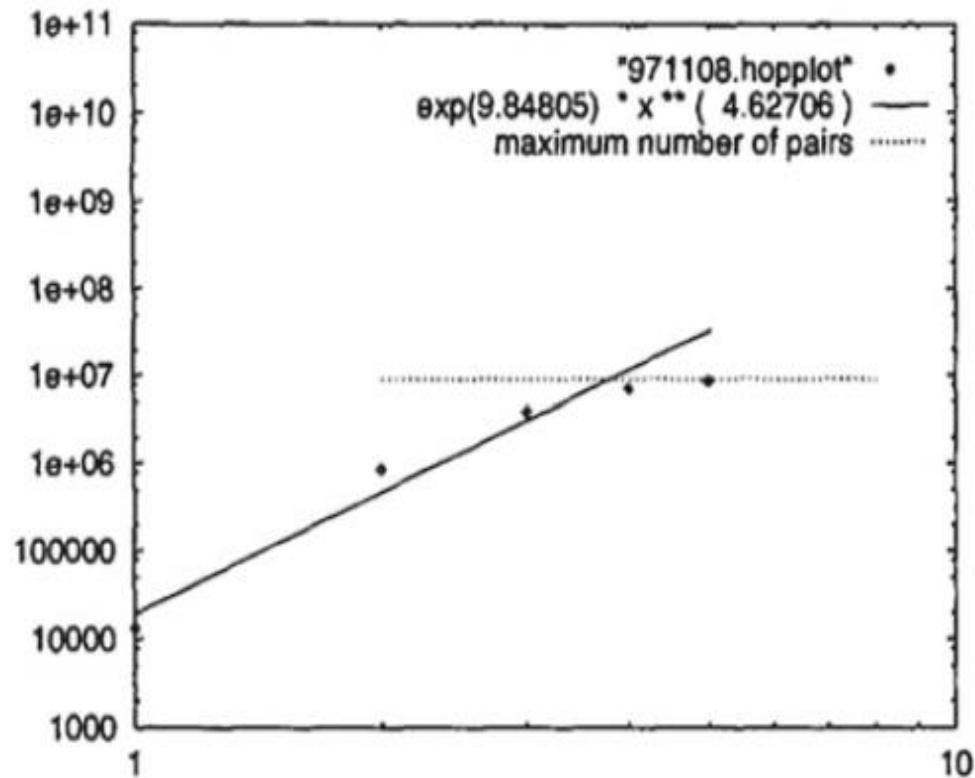
HOP PLOT EXPONENT H

- The total number of pairs of nodes $P(h)$ within h hops, defined as the *total number of pairs of nodes within less or equal to h hops, including self-pairs, and counting all other pairs twice*, is plotted as the function of the number of hops h in log-log scale
 - The total number of pairs of nodes, $P(h)$, within h hops, is proportional to the number of hops to the power of a constant, H

$$P(h) \propto h^H, h \ll \delta$$

- where δ is the diameter of the graph.





Log-log plot of the number of pairs of nodes $P(h)$ within h hops versus number of hops h for Int-11-97



- If we plot the number of pairs of nodes, $P(h)$, within h hops as a function of the number of hops in log-log scale. For $h < \delta$, the slope of this plot is defined to be the hop-plot exponent H
- The number of pairs within h hops is as given in

$$P(h) = \begin{cases} ch^H, & h \ll \delta \\ |V|^2, & h \geq \delta \end{cases}$$

- where $c = |V| + 2|E|$ to satisfy internal conditions.



- Given a graph with $|V|$ nodes, $|E|$ edges and H hop-plot exponent, the effective diameter, δ_{ef} is defined as

$$\delta_{ef} = \left(\frac{|V|^2}{|V| + 2|E|} \right)^{1/H}$$

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- The average size of the neighborhood, $NN(h)$, within h hops as a function of the hop-plot exponent, H , for $h \ll \delta$, is as

$$NN(h) = \frac{c}{|V|} h^H - 1$$

- where $c = |V| + 2|E|$ to satisfy internal conditions



- The average out-degree estimate , $NN'(h)$, within h hops with average out-degree \bar{d} , is

$$NN'(h) = \bar{d}(\bar{d} - 1)^{h-1}$$



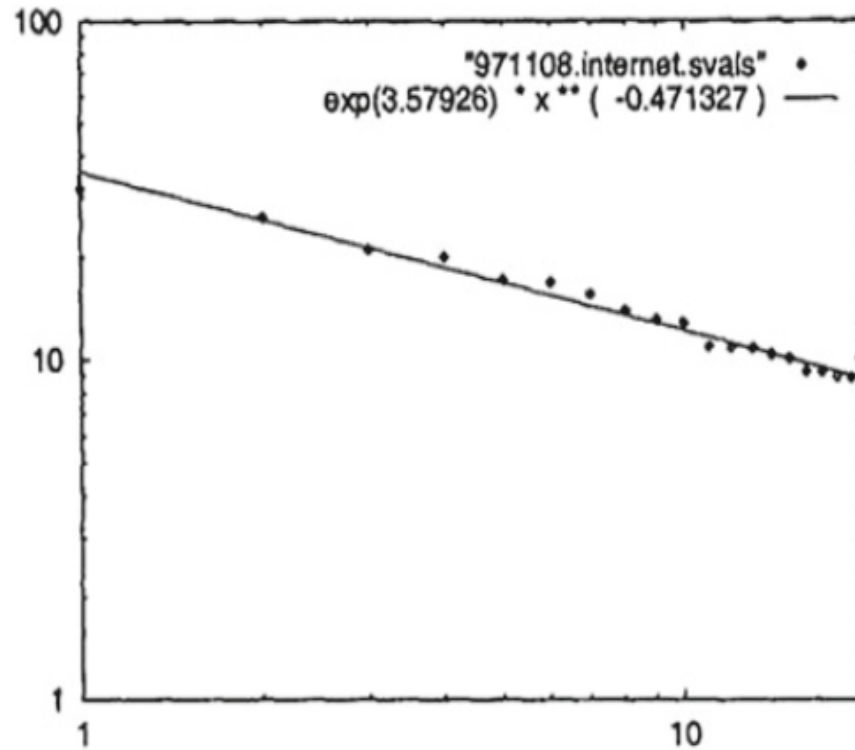
EIGEN EXPONENT ε

- The plot of the eigenvalue λ_i as a function of i in the log-log scale for the eigenvalues in the decreasing order
 - The eigenvalues, λ_i , of a graph are proportional to the order, i , to the power of a constant, ε

$$\lambda_i \propto i^\varepsilon$$

- The Eigen exponent ε is defined as the *slope of the plot of the sorted eigenvalues as a function of their order in the log-log scale.*





Log-log plot of the eigenvalues in decreasing order for Int-11-97

