

# Networks: An Introduction (Newman)

## Chapter 8.1 - 8.4 discussion

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# The large-scale structure of real world networks

We will apply the ideas of chapter 6 and 7 to real world networks to get a sense of how they behave in general (mostly).

For today we will look at:

1. **Components**

- ▶ Undirected and
- ▶ Directed networks

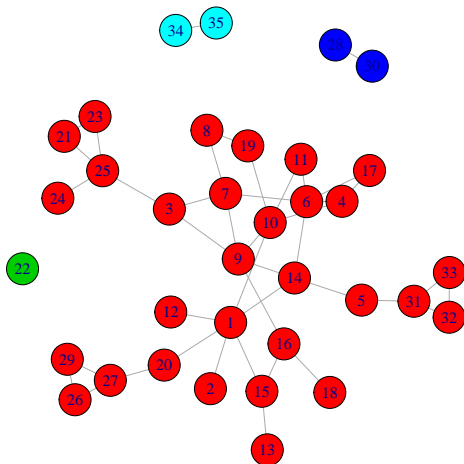
2. **Shortest Path and the Small-World Effect**

3. **Degree Distribution**

4. **Power Laws and Scale-Free Networks**

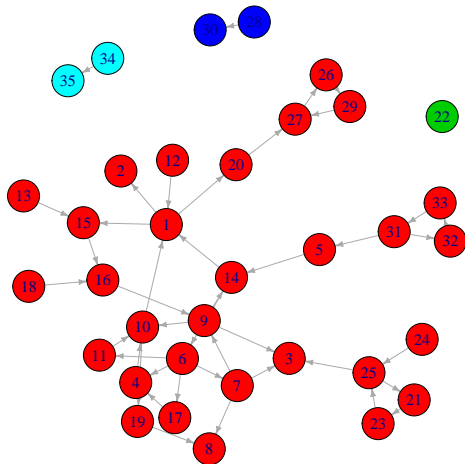
# Components (Undirected Networks)

Any two vertices in a component must be connected.



## Weakly Connected Components (Directed Networks)

Any two vertices in a component must be connected without regards to



# Some Examples of Large Components and Large Weakly Connected Components

We typically find that large components are usually more than half of vertex size ( $n$ ) and not infrequently over 90% .

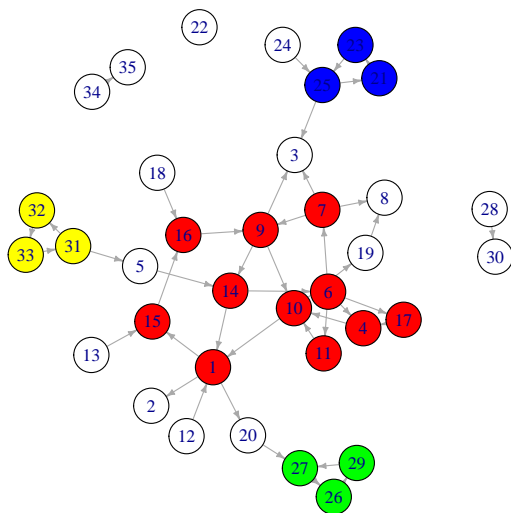
**Table :** (Table 18.1) Basic stats for some selected networks. Where  $n$ =no. of vertices,  $m$ =no. of edges,  $c$ =mean degree,  $S$ =% of largest component.

Network	Type	$n$	$m$	$c$	$S$
Marine food web	Directed	134	598	4.46	1.000
Software pack.	Directed	1,439	1,723	1.20	0.998
Metabolic	Undirected	765	3,686	9.64	2.56
Bio. co-auth.	Undirected	1,520,251	11,803,064	15.53	4.92
Phy. co-auth.	Undirected	52, 909	245,300	9.27	6.19
Protein int.	Undirected	2,115	2,240	2.12	6.80
Student dating	Undirected	573	477	1.66	16.01

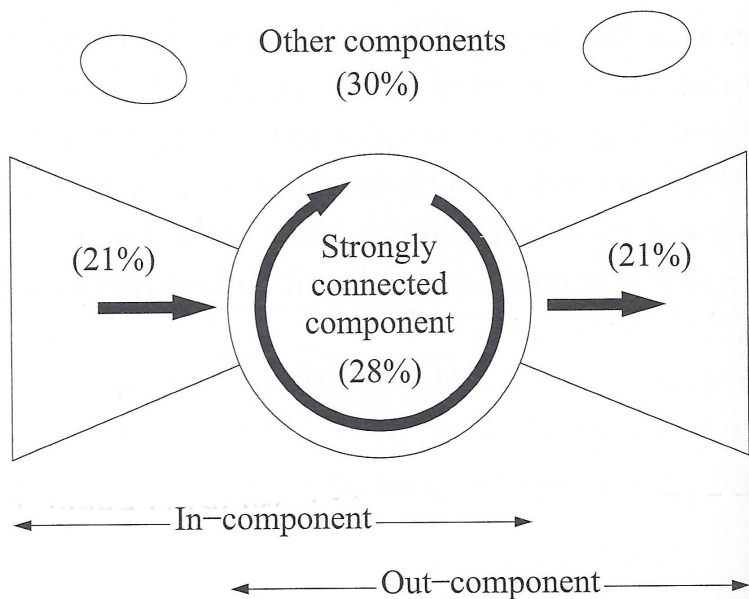
# Strongly Connected Components (Directed Networks)

- ▶ Any two vertices in a component must be connected (directed).
- ▶ Every strongly connected component generate an **in-component** and an **out-component**.

# Example: Strongly Connected Components



# The Bowtie Effect





# The Shortest Path and the Small-World Effect

- ▶ **Small-world effect:** the observation that in most real world networks the distances between vertices are small.
- ▶ Recall:  $l_i = \frac{1}{n} \sum_j d_{ij}$ , where  $d_{ij}$  is the shortest path between  $i$  and  $j$ ; is the mean geodesic distance for vertex  $i$ .

The mean geodesic distance of the network is defined as:

$$l = \frac{1}{n} \sum_i l_i.$$

- ▶ For random graphs (Erdos-Reyni graphs):  $l \propto \log(n)$
- ▶ For scale free networks:  $l \propto \log(\log(n))$

(The definition of a scale-free network is below)

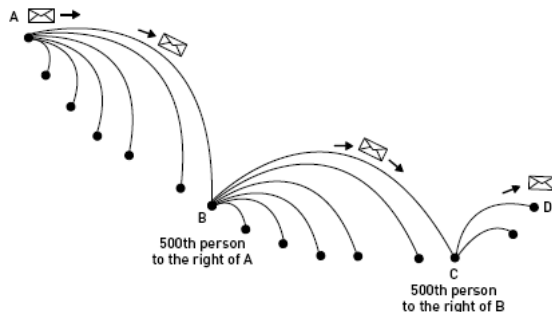
# The Shortest Path and the Small-World Effect

**Table :** (Table 18.1) Basic stats for some selected networks. Where  $l$ =mean geodesic distance connected vertex pairs.

Network	Type	n	m	c	$l$
Metabolic	Undirected	765	3,686	9.64	2.56
Bio. co-auth.	Undirected	1,520,251	11,803,064	15.53	4.92
Phy. co-auth.	Undirected	52, 909	245,300	9.27	6.19
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# Funneling and the Sociometric Superstars

- ▶ (Suggested by Milgram) **Funneling**: The idea that a given vertex has one or two connecting vertices that are well connected (i.e. have a high degree).



# Degree Distributions

- ▶ The **distribution of degrees** of a given network is a *KEY* fundamental property.
- ▶ Consider a simple undirected network with vertices,  $\{1, 2, \dots, n\}$ .
- ▶ If we pick a vertex  $i \in \{1, 2, \dots, n\}$  at random, can we guess what its degree  $k_i$  will be ?

No: But we know that it can be any counting number between 0 and  $n - 1$ .

Also depending on what natural phenomena the network is describing, we can say something sensible (on average).

# Ways to Characterize A Degree Distribution

Table : Degree Distribution in table form

degree:	0	1	2	...	n-1
probability:	$p_0$	$p_1$	$p_2$	...	$p_{n-1}$

Or

- ▶ Mathematical Function (aka. probability mass function, pmf):

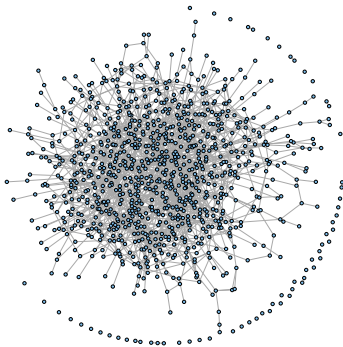
$$p(k), \quad k = 0, 1, 2, \dots, n-1.$$

When  $p(k)$  follows the Power Law we say that we have a **scale-network**.

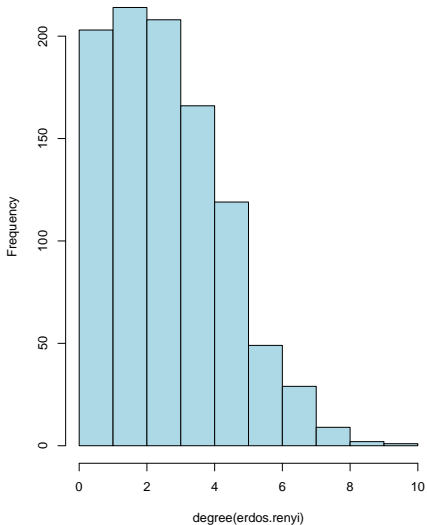
When a network has random connections (with a fixed probability of connection) then we have the **Erdos-Renyi network**. In this case  $p(k)$  will follow a Binomial Law (usually approximated by a Poisson Law).

# Example: Erdos-Renyi (random)

Erdos & Renyi (n=1000, p=0.002)

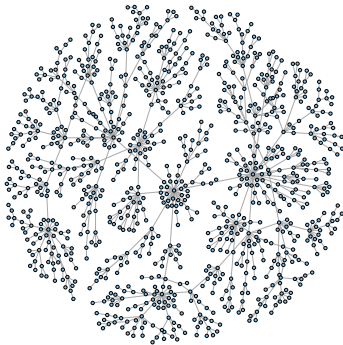


Histogram of degree(erdos.renyi)

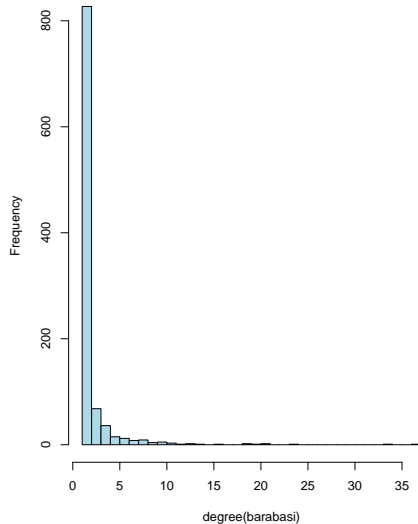


# Example: Barabasi and Albert (scale free)

Barabasi & Albert (n=1000)



Histogram of degree(barabasi)



# Directed Networks

For directed networks we can talk about the in-degree distribution and the out-degree distribution, separately.

Ideally we would like to know the joint distribution instead.  
This is an on-going research topic.

Next time we will look more closely at the Power Law and some of its consequences.