

We want to consider a set  $G$  of simple graphs with  $N$  nodes

We can define  $P(G)$  the probability of observing a specific  $G$  network

$$\sum_{G \in \mathcal{G}} P(G) = 1$$

We can say: we want a network with some specific metrics (constraints), and ask ourselves what is the probability of observing a certain graph with those constraints.

We use an approach usually used in statistical physics:

$$S = \dots$$

we maximize the entropy with 1 constrain (sum of  $P(G) = 1$ ) + one constrain for each metrics we want respected

$$P(G) = \dots$$

$$e^{H(G)} \dots$$

$$E[y] = \dots$$

for any  $y$  quantity

Let's assume that:

$$H(G) = \beta \sum_{i=1}^m$$

lagrange multiplier

this is the average of a quantity we will see in a bit  
if we have only one constraint so  $i=1$

$$\Rightarrow P(G) = \frac{e^{\beta m}}{Z}$$

$$Z = \sum_{G \in \mathcal{G}} e^{\beta m}$$

↑

how do we compute such a sum over all the graphs in the set  $\mathcal{G}$ ?

We sum over every possible values of the elements of the adjacency matrix

$$m = \sum_{i,j} A_{ij} \equiv \text{the number of links in our graphs}$$

The only constrain here is a fixed number of links! It's erdos-renyi

(The number of nodes is fixed from the very start of this model, it's always fixed)

We can also derive  $\beta$  the lagrange multiplier

$$\underbrace{p_{vw}}_{\left\{ \right.} = \langle A_{vw} \rangle = \frac{1}{Z} \sum A_{vw} \exp \left( \beta \sum_{i,j} A_{ij} \right) = \dots = \frac{\langle m \rangle}{\binom{n}{2}}$$

This is the probability of having a link between node  $v$  and node  $w$

It's just like the prob that generate Erdos-Renyi with  $p = \frac{\langle m \rangle}{\binom{n}{2}}$

We can get also the configuration model:

Let's assume  $H(G) = \sum_{i=1}^N \beta_i k_i$   $k$  is the degree, so it's fixated for every node!

We compute  $Z$  as well

$$\Rightarrow p_{vw} = \langle A_{vw} \rangle = \dots = \frac{e^{(\beta_v + \beta_w)}}{1 + e^{(\beta_v + \beta_w)}}$$

if  $\beta \ll 1$

$$\Rightarrow p_{vw} = e^{(\beta_v + \beta_w)} = \dots = \underbrace{C^2}_{\frac{1}{2m}} \langle k_v \rangle \langle k_w \rangle$$

and we find the values of  $\beta$