

# Online Resource 1

immediate

March 1, 2019

## 1 Derivation of Neighborhood Proximity

To derive the measure neighborhood proximity, let us consider two nodes  $u$  and  $v$ , with  $N_{uv} \neq \emptyset$ , between which the closeness is to be determined. We define our measure called *neighborhood proximity* of  $u$  and  $v$ , denoted as  $\rho(u, v)$ , in such a way that it depends on the following points:

- (1) The total weight on the edges falling from the common neighbors  $w \in N_{uv}$  to the nodes  $u$  and  $v$ , i.e.,  $\sum_{w \in N_{uv}} (a_{uw} + a_{vw})$ .
- (2) The total weight on the edges falling among the common neighbors, i.e.,  $\sum_{w, x \in N_{uv}} a_{wx}$ .
- (3) The similarity between the weights of the edges falling from each common neighbor  $w \in N_{uv}$  to the nodes  $u$  and  $v$ .

The points (1) and (2) above can be combined into one factor  $\sum_{w, x \in N_{uv}} (a_{uw} + a_{vw} + a_{wx})$  which is smaller than or equal to  $\sum_{w \in N_{uv}} s_w$ . Thus we have

$$0 < \frac{\sum_{w, x \in N_{uv}} a_{uw} + a_{vw} + a_{wx}}{\sum_{w \in N_{uv}} s_w} \leq 1 \quad (1)$$

The upper bound in the Eq. (1) is achieved when every neighbor of a common neighbor  $w \in N_{uv}$  is either  $u$ ,  $v$  or some  $x \in N_{uv}$ .

To understand how point (3) affects  $\rho(u, v)$ , let us assume that for some  $w \in N_{uv}$ ,  $a_{uw}$  is much higher than  $a_{vw}$ . This means that the interaction between  $u$  and  $w$  is stronger while that between  $v$  and  $w$  is comparatively weak. So, in a sense  $w$  obstructs the interaction between  $u$  and  $v$ . If this situation occurs for each  $w \in N_{uv}$ , then the quality of interaction between  $u$  and  $v$  would worsen. So, to improve  $\rho(u, v)$ , the weights  $a_{uw}$  and  $a_{vw}$  must be as close as possible. For this reason, we consider the term  $\min(a_{uw}, a_{vw})$ . It has the highest value when both  $a_{uw}$  and  $a_{vw}$  are equal. Now taking the sum of these terms over all such  $w$ 's we get the term

$$\sum_{w \in N_{uv}} \min(a_{uw}, a_{vw})$$

Now if  $u$  and  $v$  are neighbors of each other, then  $a_{uv}$  would also improve  $\rho(u, v)$ . Hence, finally we consider the term,

$$\sum_{w \in N_{uv}} \min(a_{uw}, a_{vw}) + a_{uv}$$

Note that

$$\sum_{w \in N_{uv}} \min(a_{uw}, a_{vw}) + a_{uv} \leq \sum_{w \in N_u} a_{uw} = s_u$$

Similarly,

$$\sum_{w \in N_{uv}} \min(a_{uw}, a_{vw}) + a_{uv} \leq s_v$$

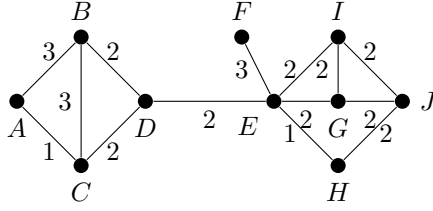


Fig. 1: A weighted network

Thus we have

$$0 < \frac{\sum_{w \in N_{uv}} \min(a_{uw}, a_{vw}) + a_{uv}}{\min(s_u, s_v)} \leq 1 \quad (2)$$

The upper bound in Eq. (2) is achieved when  $a_{uw} = a_{vw}$ , for all  $w \in N_{uv}$  and  $N_{uv} = N_u$  or  $N_{uv} = N_v$ . Now after understanding how the terms in Eq. (1) and (2) can affect the closeness of two nodes, defining  $\rho(u, v)$  is mere a matter of taking a suitable combination of these two terms, for example the average or the product. Here, we take the product as it gives a stricter definition of closeness. (the product is high only when both the factors are high.) Therefore, the definition of  $\rho(u, v)$  now takes the form,

$$\rho(u, v) = \left( \frac{\sum_{w, x \in N_{uv}} a_{uw} + a_{vw} + a_{wx}}{\sum_{w \in N_{uv}} s_w} \right) \left( \frac{\sum_{w \in N_{uv}} \min(a_{uw}, a_{vw}) + a_{uv}}{\min(s_u, s_v)} \right) \quad (3)$$

Here we have assumed that  $N_{uv} \neq \emptyset$ . For  $N_{uv} = \emptyset$ , we define  $\rho(u, v)$  as follows:

$$\rho(u, v) = \frac{a_{uv}}{\min(s_u, s_v)} \quad (4)$$

If any of the vertices  $u$  and  $v$  is isolated, we simply write  $\rho(u, v) = 0$ . Now the measure is well defined for each vertex pair  $(u, v)$  and assumes all possible values between 0 and 1 including the bounds.

Consider the graph in Fig. 1 to compute the neighborhood proximity of a few vertex pairs. Using the definition of  $\rho$ , we get  $\rho(A, D) = 3/4$ ,  $\rho(D, E) = 1/3$ ,  $\rho(D, F) = 1/3$ ,  $\rho(E, F) = 1$  and  $\rho(E, J) = 5/6$ .