

Let  $\mathbf{p} = (p_u)_{u \in \mathbf{X}}$  denote the PPR vector computed over  $G_{n,r}$  (where for ease of reading we suppress dependence on the hyperparameter  $\alpha$  and seed node  $v$ .)

**Lemma 1.** *Consider running Algorithm 1 with any  $r < \sigma$  and*

$$\frac{\Psi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}])}{10} \leq \alpha \leq \frac{\Psi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}])}{9}. \quad (\text{A.1})$$

*There exists a good set  $\mathcal{C}_\sigma[\mathbf{X}]^g \subseteq \mathcal{C}_\sigma[\mathbf{X}]$  with  $\text{vol}(\mathcal{C}_\sigma[\mathbf{X}]^g) \geq \text{vol}(\mathcal{C}_\sigma[\mathbf{X}])/2$  such that the following statements hold for all  $v \in \mathcal{C}_\sigma[\mathbf{X}]^g$ :*

- *For all  $u \in \mathcal{C}[\mathbf{X}]$ ,*

$$p_u \geq \frac{4}{5} \tilde{\pi}_{n,r}(u) - \frac{2\Phi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}])}{\Psi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}]) \tilde{D}_{\min}}$$

- *For all  $u' \in \mathcal{C}'_\sigma[\mathbf{X}]$ ,*

$$p_{u'} \leq \frac{2\Phi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}])}{\Psi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}]) \tilde{D}_{\min}}$$

*Proof.* For  $S \subseteq X_1^n$ , let  $I_S$  be a diagonal matrix where  $I_{jj} = 1$  if  $X_j \in S$  and 0 otherwise. Let  $D_S$  be the corresponding degree matrix for the subgraph induced by  $S$ .  $D_S$  is a diagonal matrix where for  $X_j \in S$ ,  $(D_S)_{jj} = \sum_{i: X_i \in S} A_{ij}$ , and if  $X_j \notin S$  then  $(D_S)_{jj} = 0$ . Then, define *leakage* and *soakage* vectors

$$\begin{aligned} \ell_t &:= e_v(WI_{\mathcal{C}_\sigma[\mathbf{X}]})^t (I - D^{-1}D_{\mathcal{C}_\sigma[\mathbf{X}]}) \\ \ell &:= \sum_{t=0}^{\infty} (1 - \alpha)^t \ell_t \\ s_t &:= e_v(WI_{\mathcal{C}_\sigma[\mathbf{X}]})^t (WI_{G/\mathcal{C}_\sigma[\mathbf{X}]}) \\ s &:= \sum_{t=0}^{\infty} (1 - \alpha)^t s_t \end{aligned}$$

Roughly, the proof will unfold in four steps. The first two will result in the lower bound of (??), while the latter two will imply the upper bound in (??).

1. For  $u \in \mathcal{C}'[\mathbf{X}]$ , use the results of [1] to lower bound  $p_v(u) \geq 4/5 \tilde{\pi}(u) - \tilde{p}_\ell(u)$ , where  $\tilde{p}$  is the PPR random walk over the subgraph induced by  $\mathcal{C}_\sigma[\mathbf{X}]$ , and  $\ell$  has bounded norm  $\|\ell\|_1 \leq 2 \frac{\Phi^{btw}(\mathcal{C}_\sigma[\mathbf{X}])}{\alpha}$ .
2. Since  $r < \sigma$ , there are no edges between  $u$  and  $G/\mathcal{C}_\sigma[\mathbf{X}]$ . Therefore, the page-rank vector  $\tilde{p}_\ell$  will not assign more than  $\|\ell\|_1/d_{\min}(\mathcal{C}_\sigma[\mathbf{X}])$  probability mass to any vertex in  $\mathcal{C}'[\mathbf{X}]$ . This will conclude our proof of (??).

3. For vertices  $u' \in G/\mathcal{C}_\sigma[\mathbf{X}]$ , we can upper bound  $p_v(u) \leq p_s(u')$ . In particular, this hold for all  $u' \in \mathcal{C}'[\mathbf{X}]_{pr}$ .
4. Since  $r < \sigma$ , there are no edges between  $u'$  and  $G/\mathcal{C}'[\mathbf{X}]_{pr}$ . Therefore, the page-rank vector  $p_s$  will assign no more than  $\|s\|_1/d_{\min}(\mathcal{C}_\sigma[\mathbf{X}])$  probability mass to any vertex in  $\mathcal{C}'[\mathbf{X}]$ . Additionally,  $s$  has bounded norm  $\|s\|_1 \leq \|\ell\|_1$ . This will conclude our proof of (??), and hence Proposition ??.

**Step 1** We will begin by restating the results of [1].

Denote by  $\tilde{p}$  the PageRank vector computed only over the subgraph induced by  $\mathcal{C}_\sigma[\mathbf{X}]$ .

$$\tilde{p}_v = \alpha e_v + (1 - \alpha) \tilde{p}_v \tilde{W} \quad (\text{A.2})$$

$$= \alpha \sum_{t=0}^{\infty} (1 - \alpha)^t (e_v \tilde{W}^t) \quad (\text{A.3})$$

and correspondingly for the leakage vector

$$\tilde{p}_\ell = \alpha \ell + (1 - \alpha) \tilde{p}_\ell \tilde{W}.$$

From Lemma 3.1 of [1] we have for all  $u \in \mathcal{C}_\sigma[\mathbf{X}]$

$$p_v(u) \geq \tilde{p}_v(u) - \tilde{p}_\ell(u).$$

with  $\|\ell\|_1 \leq \frac{2\Phi^{btw}(\mathcal{C}_\sigma[\mathbf{X}])}{\alpha}$ . Moreover if, as we have specified,  $\alpha \leq \frac{1}{9T_\infty(\mathcal{C}_\sigma[\mathbf{X}])}$ , Lemma 3.2 of [1] yields a lower bound on  $\tilde{p}$

$$\tilde{p}_v(u) \geq \frac{4}{5} \tilde{\pi}(u). \quad (\text{A.4})$$

**Step 2** We turn to upper bounding  $\tilde{p}_\ell(u)$ . We have

$$\begin{aligned}
\tilde{p}_\ell(u) &= \alpha \sum_{t=0}^{\infty} (1-\alpha)^t \left( \ell \widetilde{W}^t \right) [u] \\
&= \|\ell\|_1 \alpha \sum_{t=0}^{\infty} (1-\alpha)^t \left( \frac{\ell}{\|\ell\|_1} \widetilde{W}^t \right) [u] \\
&\stackrel{(i)}{=} \|\ell\|_1 \alpha \sum_{t=1}^{\infty} (1-\alpha)^t \left( \frac{\ell}{\|\ell\|_1} \widetilde{W}^t \right) [u] \\
&\stackrel{(ii)}{\leq} \|\ell\|_1 \frac{1}{d_{\min}(\mathcal{C}_\sigma[\mathbf{X}])}
\end{aligned} \tag{A.5}$$

where (i) follows from the fact that since  $r < \sigma$ ,  $E(\mathcal{C}'[\mathbf{X}], G/\mathcal{C}_\sigma[\mathbf{X}]) = 0$  and therefore  $\ell(u) = 0$ . To see (ii), let  $q = \frac{\ell}{\|\ell\|_1} \widetilde{W}^{t-1}$ , and then

$$\begin{aligned}
\left( \frac{\ell}{\|\ell\|_1} \widetilde{W}^t \right) [u] &= (qW) [u] \\
&\leq \|q\|_1 \|W(\cdot, u)\|_\infty \\
&\stackrel{(iii)}{\leq} \frac{1}{d_{\min}(C_n^\sigma)}.
\end{aligned}$$

where (iii) comes from the fact that  $\|u - v\| \leq r$  means  $v \in \mathcal{C}_\sigma[\mathbf{X}]$ . Combining (A.5) with (A.4), and since  $\|\ell\|_1 \leq 2^{\frac{\Phi^{btw}(\mathcal{C}_\sigma[\mathbf{X}])}{\alpha}}$ , we have

$$p_v(u) \geq \frac{4}{5} \tilde{\pi}(u) - 2 \frac{9\Phi^{btw}(\mathcal{C}_\sigma[\mathbf{X}])}{d_{\min}(\mathcal{C}_\sigma[\mathbf{X}])\alpha}.$$

**Step 3** To get the corresponding upper bound on  $p_v(u')$ , we will use the soakage vectors  $s$  and  $s_t$ . We will first argue that  $s$  is a worse starting distribution – meaning it puts uniformly more mass outside the cluster – than simply starting at  $v$ .

**Lemma 2.** *For all  $u' \notin \mathcal{C}_\sigma[\mathbf{X}]$ ,*

$$p_v(u') \leq p_s(u'). \tag{A.6}$$

The proof of Lemma 2 is left to the supplement. It follows largely the same steps as Lemma 3.1 of [1], except over  $G/\mathcal{C}_\sigma[\mathbf{X}]$  rather than  $\mathcal{C}_\sigma[\mathbf{X}]$ .

**Step 4** Just as we upper bounded the probability mass  $\tilde{p}_\ell$  could assign to any one vertex, we can upper bound

$$\begin{aligned}
\tilde{p}_s(u) &= \alpha \sum_{t=0}^{\infty} (1-\alpha)^t \left( s \widetilde{W}^t \right) [u] \\
&= \|s\|_1 \alpha \sum_{t=0}^{\infty} (1-\alpha)^t \left( \frac{s}{\|s\|_1} \widetilde{W}^t \right) [u] \\
&= \|s\|_1 \alpha \sum_{t=1}^{\infty} (1-\alpha)^t \left( \frac{s}{\|s\|_1} \widetilde{W}^t \right) [u] \\
&\leq \|s\|_1 \frac{1}{d_{\min}(\mathcal{C}_\sigma[\mathbf{X}])}. \tag{A.7}
\end{aligned}$$

Finally, by the definition of  $s_t = e_v(WI_{\mathcal{C}_\sigma[\mathbf{X}]})^t(WI_{G/\mathcal{C}_\sigma[\mathbf{X}]})$ , letting  $q_t = e_v(WI_{\mathcal{C}_\sigma[\mathbf{X}]})^t$  for ease of notation, we have

$$\begin{aligned}
\|s_t\|_1 &= \|q_t(WI_{G/\mathcal{C}_\sigma[\mathbf{X}]})\|_1 \\
&= \sum_{u' \in G} \sum_{u \in G} q_t(u) (WI_{G/\mathcal{C}_\sigma[\mathbf{X}]})[u, u'] \\
&= \sum_{u' \in G/\mathcal{C}_\sigma[\mathbf{X}]} \sum_{u \in \mathcal{C}_\sigma[\mathbf{X}]} \frac{q(u)}{d(u)} I(e_{u,u'} \in G) \\
&= \sum_{u \in \mathcal{C}_\sigma[\mathbf{X}]} \frac{q(u)(d(u) - d_{\mathcal{C}_\sigma[\mathbf{X}]}(u))}{d(u)} \\
&= \|q_t(I - D^{-1}D_{\mathcal{C}_\sigma[\mathbf{X}]})\|_1 = \|\ell_t\|_1.
\end{aligned}$$

and as a result  $\|s\|_1 = \|\ell\|_1$ . Combining with  $\|\ell\|_1 \leq 2 \frac{\Phi^{btw}(\mathcal{C}_\sigma[\mathbf{X}])}{\alpha}$  and (A.7) yields the desired upper bound.  $\square$

**Lemma 3.** *Let  $\mathcal{C}_\sigma$  satisfy the conditions of Theorem 4. For  $r < \sigma$ , the following statements hold with probability tending to one as  $n \rightarrow \infty$ :*

$$\begin{aligned}
D_{\min}(\mathcal{C}_\sigma[X]; \tilde{G}_{n,r}) &\geq \frac{1}{2} \nu_d r^d \lambda_\sigma \\
D_{\max}(\mathcal{C}_\sigma[X]; \tilde{G}_{n,r}) &\leq 2 \nu_d r^d \Lambda_\sigma \\
\widetilde{\text{vol}}_{n,r}(\tilde{G}_{n,r}) &\leq 2\nu(\mathcal{C}_\sigma) \Lambda_\sigma
\end{aligned}$$

where  $D_{\min}(\mathcal{C}_\sigma[X]; \tilde{G}_{n,r})$  is the minimum degree of any vertex  $v \in \mathcal{C}_\sigma[X]$  in the subgraph  $\tilde{G}_{n,r}$ , and analogously for  $D_{\max}(\mathcal{C}_\sigma[X]; \tilde{G}_{n,r})$ .

The statement follows immediately from Lemma ??.

## .1 Proof of Theorem 4

We note that by Theorems 1 and 2,

$$\kappa_2(\mathcal{C}) \geq \frac{\Phi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}])}{\Psi_{n,r}(\mathcal{C}_\sigma[\mathbf{X}])}.$$

As a result Lemma 1 implies

$$\begin{aligned} p_u &\geq \frac{4}{5} \tilde{\pi}_{n,r}(u) - \frac{18\kappa_2(\mathcal{C})}{\tilde{D}_{\min}} & (u \in \mathcal{C}[\mathbf{X}]) \\ p_{u'} &\leq \frac{18\kappa_2(\mathcal{C})}{\tilde{D}_{\min}} & (u' \in \mathcal{C}'[\mathbf{X}]) \end{aligned} \quad (\text{A.8})$$

We then have

$$\begin{aligned} \tilde{\pi}_{n,r}(u) &\geq \frac{D_{\min}(\mathcal{C}_\sigma[X]; \tilde{G}_{n,r})}{\widetilde{\text{vol}}_{n,r}(\tilde{G}_{n,r})} \\ &\geq \frac{D_{\min}(\mathcal{C}_\sigma[X]; \tilde{G}_{n,r})}{\tilde{n}\tilde{D}_{\max}} \end{aligned}$$

and application of Lemma 3 yields

$$\tilde{\pi}_{n,r}(u) \geq 8 \frac{\lambda_\sigma}{\nu(\mathcal{C}_\sigma)\Lambda_\sigma^2} \quad (\text{A.9})$$

and

$$\frac{1}{\tilde{D}_{\min}} \leq \frac{2}{\nu_d r^d \lambda_\sigma} \quad (\text{A.10})$$

with probability tending to 1 as  $n \rightarrow \infty$ , for all  $u \in \mathcal{C}[\mathbf{X}]$ .

Combining (A.8), (A.9) and (A.10), along with the requirement on  $\kappa_2(\mathcal{C})$  given by (17), we have

$$\begin{aligned} p_u &\geq 3/5 \frac{\lambda_\sigma}{\nu(\mathcal{C}_\sigma)\Lambda_\sigma^2} \\ p_{u'} &\leq 1/5 \frac{\lambda_\sigma}{\nu(\mathcal{C}_\sigma)\Lambda_\sigma^2} \end{aligned}$$

for any  $u \in \mathcal{C}$ ,  $u' \in \mathcal{C}'$ . As a result, if  $\pi_0 \in (2/5, 3/5) \cdot \frac{\lambda_\sigma}{\nu(\mathcal{C}_\sigma)\Lambda_\sigma^2}$ , as  $n \rightarrow \infty$  with probability tending to one any sweep cut of the form of (6), including the output set  $\hat{\mathcal{C}}$ , will successfully recover  $\mathcal{C}$  in the sense of (9).

## References

- [1] Zeyuan Allen Zhu, Silvio Lattanzi, and Vahab S Mirrokni. A local algorithm for finding well-connected clusters. In *ICML (3)*, pages 396–404, 2013.