

1 Proof of Theorem 3.1

We first consider the objective function for $\lambda > 0$:

$$J(\mathbf{s}) = \|\mathbf{s} - \mathbf{r}_g\|^2 + \lambda \mathbf{s}^\top ((1 - \alpha - \beta)L_u + \alpha L_g + \beta L_{\text{uni}}) \mathbf{s} \quad (1)$$

. To find the optimal \mathbf{s}_g in the sense of minimizing $J(\mathbf{s})$, we take the gradient of $J(\mathbf{s})$ with respect to \mathbf{s} and set it to zero:

$$\nabla_{\mathbf{s}} J(\mathbf{s}) = 2(\mathbf{s} - \mathbf{r}_g) + 2\lambda ((1 - \alpha - \beta)L_u + \alpha L_g + \beta L_{\text{uni}}) \mathbf{s} = \mathbf{0}, \quad (2)$$

where $\mathbf{0}$ is the all-zeros vector. Eq. (2) can be rewritten as

$$(I + \lambda ((1 - \alpha - \beta)L_u + \alpha L_g + \beta L_{\text{uni}})) \mathbf{s} = \mathbf{r}_g. \quad (3)$$

Due to the fact that $L_u = I - \bar{P}_u$, $L_g = I - \bar{P}_g$, and $L_{\text{uni}} = I - \bar{P}_{\text{uni}}$, we have

$$\begin{aligned} A &\triangleq I + \lambda ((1 - \alpha - \beta)(I - \bar{P}_u) + \alpha(I - \bar{P}_g) + \beta(I - \bar{P}_{\text{uni}})) \\ &= (1 + \lambda)I - \lambda ((1 - \alpha - \beta)\bar{P}_u + \alpha\bar{P}_g + \beta\bar{P}_{\text{uni}}). \end{aligned} \quad (4)$$

Let $H = (1 - \alpha - \beta)\bar{P}_u + \alpha\bar{P}_g + \beta\bar{P}_{\text{uni}}$. Then, it follows that

$$A = (1 + \lambda)I - \lambda H, \quad (5)$$

which results in

$$\begin{aligned} \mathbf{s} &= ((1 + \lambda)I - \lambda H)^{-1} \mathbf{r}_g \\ &= \frac{1}{1 + \lambda} \left(I - \frac{\lambda}{1 + \lambda} H \right)^{-1} \mathbf{r}_g. \end{aligned} \quad (6)$$

Letting $\mu = \frac{\lambda}{1 + \lambda}$, we have

$$\mathbf{s} = (1 - \mu)(I - \mu H)^{-1} \mathbf{r}_g. \quad (7)$$

Assuming $0 < \alpha, \beta < 1$, since $0 < \mu < 1$ and H is a combination of three matrices \bar{P}_u , \bar{P}_g , and \bar{P}_{uni} , the spectral radius of μH is less than 1. Thus, we can expand the inverse of $I - \mu H$ using the Neumann series:

$$(I - \mu H)^{-1} = \sum_{n=0}^{\infty} (\mu H)^n. \quad (8)$$

Thus, Eq. (7) can be rewritten as

$$\mathbf{s} = (1 - \mu) \sum_{n=0}^{\infty} (\mu H)^n \mathbf{r}_g. \quad (9)$$

This infinite series represents a polynomial function of H applied to \mathbf{r}_g . Truncating the series at a finite K , we have

$$\mathbf{s} \approx (1 - \mu) \sum_{n=0}^K (\mu H)^n \mathbf{r}_g, \quad (10)$$

which indicates that \mathbf{s} can be approximated by applying a polynomial graph filter to \mathbf{r}_g , where the filter coefficients are determined by $1 - \mu$ and μ^n .

Given that H is a weighted sum of \bar{P}_u , \bar{P}_g , and \bar{P}_{uni} , and each item similarity graph can be represented using their respective polynomial filters $f_1(\bar{P}_u)$, $f_2(\bar{P}_g)$, and $f_3(\bar{P}_{\text{uni}})$, Eq. (10) can be finally expressed as

$$\mathbf{s} \approx \mathbf{r}_g \left((1 - \alpha - \beta) f_1(\bar{P}_u) + \alpha f_2(\bar{P}_g) + \beta f_3(\bar{P}_{\text{uni}}) \right), \quad (11)$$

where the polynomial filters approximate the series expansion. Therefore, we conclude that Group-GF's prediction is an approximate solution to the optimization problem in Eq.(11) under the condition that the spectral radius of μH is less than 1, which holds when $0 < \alpha, \beta < 1$. This completes the proof of Theorem 3.1.