

Supplemental Material

Yushan Zheng

Beijing Advanced Innovation Center for Biomedical Engineering, Beihang University,
Beijing 100191, China. yszheng@buaa.edu.com

This is the supplemental material for paper "Tracing Diagnosis Paths on Histopathology WSIs for Diagnostically Relevant Case Recommendation, MICCAI 2020". It includes the algorithm of the ROI feature extraction network.

Algorithm 1: The computation flowchart of the ROI feature extraction network $\mathbf{g} = \mathcal{F}_{graph}(\mathbf{A}, \mathbf{X})$, where $\mathcal{G}_{embed}^{(l)}$ and $\mathcal{G}_{pool}^{(l)}$ are GCN modules formulated in Algorithm2.

Input:

$\mathbf{A} \in \mathbb{R}^{n_p \times n_p} \leftarrow$ The adjacency matrix of the n_p patches in the given ROI.
 $\mathbf{X} \in \mathbb{R}^{d \times n_p} \leftarrow$ The patch features for the ROI.
 $L \leftarrow$ The number of Diffpool modules.
 $n_l \leftarrow$ The number of graph vetexes after the l -th pooling. Specifically, $n_0 = n_p$ and $n_{l+1} < n_l$.

1 $\mathbf{X}^{(0)} \leftarrow \mathbf{X}_i^T$;
2 $\mathbf{A}^{(0)} \leftarrow \mathbf{A}_i$;
3 $\mathbf{Z}^{(0)} \leftarrow \mathcal{G}_{embed}^{(0)}(\mathbf{A}^{(0)}, \mathbf{X}^{(0)}) \in \mathbb{R}^{n_0 \times d}$;
4 **for** $l = 0$ **to** $L - 1$ **do**
5 $\mathbf{S}^{(l)} \leftarrow softmax_r\left(\mathcal{G}_{pool}^{(l)}(\mathbf{A}^{(l)}, \mathbf{X}^{(l)})\right) \in \mathbb{R}^{n_{l+1} \times n_l}$;
6 $\mathbf{X}^{(l+1)} \leftarrow \mathbf{S}^{(l)T} \mathbf{Z}^{(l)} \in \mathbb{R}^{n_{l+1} \times d}$;
7 $\mathbf{A}^{(l+1)} \leftarrow \mathbf{S}^{(l)T} \mathbf{A}^{(l)} \mathbf{S}^{(l)} \in \mathbb{R}^{n_{l+1} \times n_{l+1}}$;
8 $\mathbf{Z}^{(l+1)} \leftarrow \mathcal{G}_{embed}^{(l+1)}(\mathbf{A}^{(l+1)}, \mathbf{X}^{(l+1)}) \in \mathbb{R}^{n_{l+1} \times d}$;
9 **end**
10 $\mathbf{g} \leftarrow Maxpool_r(\mathbf{X}^{(L)})$;
11 **return** \mathbf{g} ;

Algorithm 2: The computation flowchart of GCN $\mathbf{Z} \leftarrow \mathcal{G}(\mathbf{A}, \mathbf{X})$

Input:

- \mathbf{X} \leftarrow The features.
- \mathbf{A} \leftarrow The adjacency matrix.

- 1 $\tilde{\mathbf{A}} \leftarrow \mathbf{A} + \mathbf{I};$
- 2 $\tilde{\mathbf{D}} \leftarrow diag(\sum_j \tilde{\mathbf{A}}_{1j}, \sum_j \tilde{\mathbf{A}}_{2j}, \dots, \sum_j \tilde{\mathbf{A}}_{nj}),;$
- 3 $\mathbf{H}^{(0)} \leftarrow \mathbf{X};$
- 4 **for** $k = 1$ **to** K **do**
- 5 | $\mathbf{W}^{(k)}$ \leftarrow the trainable weighting matrix for the k -th step;
- 6 | $\mathbf{H}^{(k)} = ReLU(\tilde{\mathbf{D}}^{-\frac{1}{2}} \tilde{\mathbf{A}} \tilde{\mathbf{D}}^{-\frac{1}{2}} \mathbf{H}^{(k-1)} \mathbf{W}^{(k)});$
- 7 **end**
- 8 $\mathbf{Z} \leftarrow \mathbf{H}^{(K)};$
- 9 **return** $\mathbf{Z};$
