# PEG-PYG Description

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The PEG layer from the "Equivariant and Stable Positional Encoding for More Powerful Graph Neural Networks"

The PEG layer:

$$X^{'},Z^{'}=(\sigma[(\hat{A}\odot M)XW],Z)$$

where  $M_{uv} = MLP(||Z_u - Z_v||), \forall u, v \in V$ .  $\hat{A} = \hat{D}^{-1/2}(A+I)\hat{D}^{-1/2}$  is the normalized adjacent matrix and  $\hat{D}_{ii} = \Sigma_{j=0}\hat{A}_{ij}$  is diagonal degree matrix.  $\odot$  denotes Hadamard product and Z is the positional encoding. The adjacency matrix can include other values than 1 representing edge weights via the optional edge\_weight tensor.

#### PARAMETERS:

- in\_feats\_dim: (int) Size of each input node feature sample
- pos\_dim: (int) Size of each input positional encoding sample. Notice in PEG we do not update the positional encodings.
- out\_feats\_dim: (int) Size of each output node embedding sample.
- edge\_mlp\_dim: (int) We use MLP to make one to one mapping between the relative information and edge weight. edge\_mlp\_dim represents the hidden units dimension in the MLP. (default: 32)
- improved: (bool, optional) If set to :obj: 'True', the layer computes  $\hat{A}$ ' as A + 2I. (default: 'False')
- cached: (bool, optional) If set to: True, the layer will cache the computation of  $\hat{D}^{-1/2}\hat{A}\hat{D}^{-1/2}$  on first execution, and will use the cached version for further executions. This parameter should only be set to: 'True' in transductive learning scenarios. (default: 'False')
- add\_self\_loops: (bool, optional) If set to: 'False', will not add self-loops to the input graph. (default: 'True')
- normalize: (bool, optional) Whether to add self-loops and compute symmetric normalization coefficients on the fly. (default: 'True')

- bias: (bool, optional) If set to: 'False', the layer will not learn an additive bias. (default: 'True')
- use\_formerinfo: (bool, false) Whether to use previous layer's output to update node features. (default: 'False')
- $\bullet \ \ **kwargs: (optional) \ Additional \ arguments \ of: \ class: `torch\_geometric.nn.conv. Message Passing'.$

### **SHAPES**

- input: node features :( $|\mathcal{V}|$ ,  $F_{in}$ ), positional encodings: ( $|\mathcal{V}|$ ,  $P_{in}$ ), edge indices:' $(2,|\mathcal{E}|)$ , edge weights: ( $|\mathcal{E}|$ )'(optional)
- output: node features: ( $|\mathcal{V}|$ ,  $F_{out}$ ), positional encodings: ( $|\mathcal{V}|$ ,  $P_{out}$ ) reset\_parameters()

forward(x: torch.Tensor, edge\_index: Union[torch.Tensor, torch\_sparse.tensor.SparseTensor], edge\_weight: Optional[torch.Tensor] = None)  $\rightarrow$  torch.Tensor