

# PEG-PYG Description

Haorui Wang

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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  $\hat{D}_{ii} = \sum_{j=0} \hat{A}_{ij}$  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')
- **update\_coors:** (bool, optional) Whether to update positional encodings. (default: 'False')
- **use\_formerinfo:** (bool, false) Whether to use previous layer's output to update node features. (default: 'False')
- **norm\_coors:** Whether to normalize positional encodings. Only used when `update_coors = True`. (default: 'False')
- **\*\*kwargs:** (optional) Additional arguments of: `class:'torch_geometric.nn.conv.MessagePassing'`.

## 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) → torch.Tensor`