

# Preliminaries

## Graph Convolutional Networks

- GCN is one of the most effective convolution models
  - Considered as a general “[message-passing](#)” architecture
  - $\mathbf{H}_k = M(\mathbf{A}, \mathbf{H}_{k-1}; \mathbf{W}_{k-1})$ : hidden feature matrix computed by  $k$ -th GCL
    - $\mathbf{A}$ : adjacency matrix
    - $\mathbf{H}_{k-1}$ : hidden feature matrix
    - $\mathbf{W}_{k-1}$ : trainable parameters
    - $M$ : message propagation function for GCN

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- $M$  defined in [1stChebNet](#) (Kipf and Welling 2017) as follow:

- $\mathbf{H}_k = M(\mathbf{A}, \mathbf{H}_{k-1}; \mathbf{W}_{k-1}) = \sigma(\hat{\mathbf{A}}\mathbf{H}_{k-1}\mathbf{W}_{k-1})$

- $\hat{\mathbf{A}} = \tilde{\mathbf{D}}^{-\frac{1}{2}}\tilde{\mathbf{A}}\tilde{\mathbf{D}}^{-\frac{1}{2}}$ : normalized adjacency matrix

- $\tilde{\mathbf{A}} = \mathbf{A} + \mathbf{I}_N$ : adding self-connection

- $\tilde{\mathbf{D}}_{ii} = \sum_j \tilde{\mathbf{A}}_{ij}$ : degree of the  $i$ -th node