[1] Wu Z, Pan S, Chen F, et al. A comprehensive survey on graph neural networks[J]. arXiv preprint arXiv:1901.00596, 2019.

[2] Kipf T N, Welling M. Semi-supervised classification with graph convolutional networks[J]. arXiv preprint arXiv:1609.02907, 2016.

[3] Veličković P, Cucurull G, Casanova A, et al. Graph attention networks[J]. arXiv preprint arXiv:1710.10903, 2017.

[4] Klicpera J, Bojchevski A, Günnemann S. Predict then propagate: Graph neural networks meet personalized pagerank[J]. arXiv preprint arXiv:1810.05997, 2018.

[5] Xu K, Li C, Tian Y, et al. Representation learning on graphs with jumping knowledge networks[J]. arXiv preprint arXiv:1806.03536, 2018.

[6] Abu-El-Haija S, Kapoor A, Perozzi B, et al. N-gcn: Multi-scale graph convolution for semi-supervised node classification[J]. arXiv preprint arXiv:1802.08888, 2018.

[7] Li Q, Han Z, Wu X M. Deeper insights into graph convolutional networks for semi-supervised learning[C]//Thirty-Second AAAI Conference on Artificial Intelligence. 2018.

[8] Xu K, Hu W, Leskovec J, et al. How powerful are graph neural networks?[J]. arXiv preprint arXiv:1810.00826, 2018.

[9] Chiang W L, Liu X, Si S, et al. Cluster-gcn: An efficient algorithm for training deep and large graph convolutional networks[C]//Proceedings of the 25th ACM SIGKDD International Conference on Knowledge Discovery & Data Mining. 2019: 257-266.

[10] Huang B, Carley K M. Residual or gate? towards deeper graph neural networks for inductive graph representation learning[J]. arXiv preprint arXiv:1904.08035, 2019.

[11] Li G, Müller M, Thabet A, et al. Can GCNs Go as Deep as CNNs?[J]. arXiv preprint arXiv:1904.03751, 2019.

[12] Rong Y , Huang W , Xu T , et al. DropEdge: Towards Deep Graph Convolutional Networks on Node Classification[J]. 2019.

[13] Luan S, Zhao M, Chang X W, et al. Break the Ceiling: Stronger Multi-scale Deep Graph Convolutional Networks[C]//Advances in Neural Information Processing Systems. 2019: 10943-10953.

[14] He K, Zhang X, Ren S, et al. Deep residual learning for image recognition[C]//Proceedings of the IEEE conference on computer vision and pattern recognition. 2016: 770-778.

[15] Balduzzi D, Frean M, Leary L, et al. The shattered gradients problem: If resnets are the answer, then what is the question?[C]//Proceedings of the 34th International Conference on Machine Learning-Volume 70. JMLR. org, 2017: 342-350.

[16] Lin T Y, Dollár P, Girshick R, et al. Feature pyramid networks for object detection[C]//Proceedings of the IEEE conference on computer vision and pattern recognition. 2017: 2117-2125.

[17] 邱锡鹏. 神经网络与深度学习[M]. 第1版. 北京：机械工业出版社，2020：134-136.

[18] Zhao L, Akoglu L. PairNorm: Tackling Oversmoothing in GNNs[J]. arXiv preprint arXiv:1909.12223, 2019.

[19] Gilmer J, Schoenholz S S, Riley P F, et al. Neural message passing for quantum chemistry[C]//Proceedings of the 34th International Conference on Machine Learning-Volume 70. JMLR. org, 2017: 1263-1272.

[20] Hoang N T, Maehara T. Revisiting graph neural networks: All we have is low-pass filters[J]. arXiv preprint arXiv:1905.09550, 2019.

[21] Pei H, Wei B, Chang K C C, et al. Geom-gcn: Geometric graph convolutional networks[J]. arXiv preprint arXiv:2002.05287, 2020.

[22] Huang G, Liu S, Van der Maaten L, et al. Condensenet: An efficient densenet using learned group convolutions[C]//Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition. 2018: 2752-2761.