[1] Lawrence Page, Sergey Brin, Rajeev Motwani, and Terry Winograd. The pagerank citation ranking: Bringing order to the web. Stanford Digital Libraries Working Paper, 1998, 9(1):1-14.

[2] Jon M Kleinberg. Hubs, authorities, and communities. ACM computing surveys (CSUR), 1999, 31(4es): 5,

[3] Andrew Y Ng, Michael I Jordan, and Yair Weiss. On spectral clustering: Analysis and an algori thm//Proceedings of the Advances in neural information processing systems. Vancouver, Canada, 2002: 849–856.

[4] Jinshan Qi, Xun Liang, Zhiyu Li, Yanfang Chen, and Yuan Xu. Representation learning of lar ge-scale complex information network: concepts, methods and challenges. Chinese Journal of Computers, 2018, 41(10): 2394–2419(in Chinese)

(齐金山, 梁循, 李志宇等. 大规模复杂信息网络表示学习：概念、方法与挑战. 计算机学报, 2018, 41(10): 2394-2419).

[5] Bryan Perozzi, Rami Al-Rfou, and Steven Skiena. Deepwalk: Online learning of social represe ntations//Proceedings of the 20th ACM SIGKDD international conference on Knowledge discovery and data mining. New York, USA, 2014: 701–710.

[6] Jian Tang, Meng Qu, Mingzhe Wang, Ming Zhang, Jun Yan, and Qiaozhu Mei. Line: arge-s cale information network embedding//Proceedings of the 24th international conference on world wide web. Florence, Italy, 2015: 1067–1077.

[7] Aditya Grover and Jure Leskovec. node2vec: Scalable feature learning for networks//Proc eedings of the 22nd ACM SIGKDDinternational conference on Knowledge discovery and data mi ning. San Francisco, USA, 2016: 855–864.

[8] Joan Bruna, Wojciech Zaremba, Arthur Szlam, and Yann Lecun. Spectral networks and locally connected networks on graphs// Proceedings of the International Conference on Learning Representations (ICLR2014). Banff, Canada, 2014: URL http://arxiv.org/abs/1312.6203.

[9] Michaël Defferrard, Xavier Bresson, and Pierre Vandergheynst. Convolutional neural networks on graphs with fast localized spectral filtering// Proceedings of the Advances in Neural Information Processing Systems. Barcelona, Spain, 2016: 3844–3852.

[10] Thomas N Kipf and Max Welling. Semi-supervised classification with graph convolutional networks. arXiv preprint arXiv: 1609.02907, 2016.

[11] Felix Wu, Tianyi Zhang, Amauri Holanda de Souza Jr, Christopher Fifty, Tao Yu, and Kilian Q Weinberger. Simplifying graph convolutional networks. arXiv preprint arXiv: 1902.07153, 2019.

[12] Rianne van den Berg, Thomas N Kipf, and Max Welling. Graph convolutional matrix completion. arXiv preprint arXiv:1706.02263, 2017.

[13] John Boaz Lee, Ryan A Rossi, Xiangnan Kong, Sungchul Kim, Eunyee Koh, and Anup Rao. Higher-order graph convolutional networks. arXiv preprint arXiv:1809.07697, 2018.

[14] Hongwei Wang, Fuzheng Zhang, Jialin Wang, Miao Zhao, Wenjie Li, Xing Xie, and Minyi Guo. Exploring high-order user preference on the knowledge graph for recommender systems. ACM Transactions on Information Systems (TOIS), 2019, 37(3):32,.

[15] Xiao Wang, Houye Ji, Chuan Shi, Bai Wang, Yanfang Ye, Peng Cui, and Philip S. Yu. Graph neural networks for social recommendation// Proceedings of the 2019 World Wide Web Confere nce, San Francisco, USA, 2019: 417-426

[16] Yaguang Li, Rose Yu, Cyrus Shahabi, and Yan Liu. Diffusion convolutional recurrent neural network: Data-driven traffic forecasting//Proceedings of the International Conference on Learning Representations. Vancouver, BC, Canada, 2018.

[17] Zhiyong Cui, Kristian Henrickson, Ruimin Ke, and Yinhai Wang. Traffic graph convolutional recurrent neural network: A deep learning framework for network-scale traffic learning and forecasting. arXiv: 1802.07007, 2018.

[18] Jiani Zhang, Xingjian Shi, Junyuan Xie, Hao Ma, Irwin King, and Dit-Yan Yeung. Gaan: Gated attention networks for learning on large and spatiotemporal graphs//Proceedings of International Conference on Uncertainty in Artificial Intelligence, California, USA, 2018: 339-349.

[19] Bing Yu, Haoteng Yin, and Zhanxing Zhu. Spatio-temporal graph convolutional networks: A deep learning framework for traffic forecasting // Proceedings of the International Joint Conference on Artificial Intelligence, Stockholm, Sweden, 2018: 3634-3640.

[20] Yaqing Wang and Quanming Yao. Few-shot learning: A survey.arXiv preprint arXiv: 1904.05046, 2019.

[21] Victor Garcia and Joan Bruna. Few-shot learning with graph neural networks//Proceedings of the International Conference on Learning Representations. Vancouver, BC, Canada, 2018.

[22] Kenneth Marino, Ruslan Salakhutdinov, and Abhinav Gupta. The more you know: Using knowledge graphs for image classification// Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, Honolulu, HI, USA, 2017: 20-28

[23] Chung-Wei Lee, Wei Fang, Chih-Kuan Yeh, and Yu-Chiang Frank Wang. Multi-label zero-shot learning with structured knowledge graphs// Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, Salt Lake City,USA, 2018: 1576-1585.

[24] Michael Kampffmeyer, Yinbo Chen, Xiaodan Liang, Hao Wang, Yujia Zhang, and Eric P Xing. Rethinking knowledge graph propagation for zero-shot learning// Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, Long Beach, USA, 2019: 11487-11496.

[25] Xiaojuan Qi, Renjie Liao, Jiaya Jia, Sanja Fidler, and Raquel Urtasun. 3D graph neural networks for rgbd semantic segmentation// Proceedings of the IEEE International Conference on Computer Vision, Venice, Italy, 2017: 5209-5218.

[26] Yue Wang, Yongbin Sun, Ziwei Liu, Sanjay E Sarma, Michael M Bronstein, and Justin M Solomon. Dynamic graph cnn for learning on point clouds. arXiv preprint arXiv:1801.07829, 2018.

[27] Loic Landrieu and Martin Simonovsky. Large-scale point cloud semantic segmentation with superpoint graphs// Proceedings of the IEEE Conference on Computer Vision and Pattern Reco gnition, Salt Lake City, USA, 2018: 4558-4567.

[28] Damien Teney, Lingqiao Liu, and Anton van den Hengel. Graph-structured representations for visual question answering//Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, Honolulu, USA, 2017: 3233-3241.

[85] Xinlei Chen, Li-Jia Li, Li Fei-Fei, and Abhinav Gupta. Iterative visual reasoning beyond convolutions// Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition, Salt Lake City, USA, 2018: 7239-7248.