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# Appendix A

## List of Publications

### First Author Publications

1. **Shivansh Mishra**, Shashank Sheshar Singh, Ajay Kumar, and Bhaskar Biswas, "ELP: Link prediction in social networks based on ego network perspective", in *Physica A: Statistical Mechanics and its Applications*, Volume 605, November 2022, ISSN: 0378-4371, **DOI:10.1016/j.physa.2022.128008 (SCI, IF:3.778)**.
2. **Shivansh Mishra**, Shashank Sheshar Singh, Ajay Kumar, and Bhaskar Biswas, "HOPLP-MUL: link prediction in multiplex networks based on higher order paths and layer fusion", in *Applied Intelligence*, May 2022, ISSN: 1573-7497, **DOI:10.1007/s10489-022-03733-8 (SCI, IF:5.019)**.
3. **Shivansh Mishra**, Shashank Sheshar Singh, Ajay Kumar, and Bhaskar Biswas, "MNERLP-MUL: Merged node and edge relevance based link prediction in multiplex networks", in *Journal of Computational Science*, Volume 60, April 2022, ISSN: 1877-7503, **DOI:10.1016/j.jocs.2022.101606 (SCI, IF:3.817)**.

4. **Shivansh Mishra**, Shashank Sheshar Singh, Ajay Kumar, and Bhaskar Biswas, "CLP-MUL: Clustering-based Link Prediction in Multiplex Networks" (Under Review)

## Co-Author Publications

1. Mukesh Kumar, **Shivansh Mishra**, and Bhaskar Biswas, "PQKLP: Projected Quantum Kernel based Link prediction in Dynamic Networks", in *Computer Communications*, Volume 196, December 2022, ISSN: 0140-3664, **DOI:10.1016/j.comcom.2022.10.006 (SCI, IF:5.047)**.
2. Mukesh Kumar, **Shivansh Mishra**, and Bhaskar Biswas, "PWF: Path Weight Aggregation Feature for link prediction in dynamic networks", in *Computer Communications*, Volume 191, July 2022, ISSN: 0140-3664, **DOI:10.1016/j.comcom.2022.05.019 (SCI, IF:5.047)**.
3. Mukesh Kumar, **Shivansh Mishra**, and Bhaskar Biswas, "CFLP: A new cost based feature for link prediction in dynamic networks", in *Journal of Computational Science*, Volume 62, July 2022, ISSN: 1877-7503, **DOI:10.1016/j.jocs.2022.101726 (SCI, IF:3.817)**.
4. Mukesh Kumar, **Shivansh Mishra**, and Bhaskar Biswas, "Features fusion based link prediction in dynamic networks", in *Journal of Computational Science*, Volume 57, January 2022, ISSN: 1877-7503, **DOI:10.1016/j.jocs.2021.101493 (SCI, IF:3.817)**.
5. Ajay Kumar, **Shivansh Mishra**, Shashank Sheshar Singh, Kuldeep Singh, and Bhaskar Biswas, "Link prediction in complex networks based on Significance of Higher-Order Path Index (SHOPI)", in *Physica A: Statistical Mechanics and its Applications*, Volume 545, May 2020, ISSN: 0378-4371, **DOI:10.1016/j.physa.2019.123790 (SCI, IF:3.778)**.

6. Shashank Sheshar Singh, **Shivansh Mishra**, Ajay Kumar, and Bhaskar Biswas, "CLP-ID: Community-based link prediction using information diffusion", in *Information Sciences*, Volume 514, April 2020, ISSN: 0020-0255, **DOI:10.1016/j.ins.2019.11.026 (SCI, IF:8.233)**.

## **Book Chapters**

1. Shashank Sheshar Singh, **Shivansh Mishra**, Ajay Kumar, and Bhaskar Biswas, "Link Prediction on Social Networks Based on Centrality Measures", in *Principles of Social Networking: The New Horizon and Emerging Challenges*, Springer Singapore, pp. 71–89, 2022, **DOI:10.1007/978-981-16-3398-0\_4**.