Appendix A

List of Publications

A.1 Journal Papers

- **A. K. Singh** and L. Kailasam, "*Link Prediction-Based Influence Maximization in Online Social Networks*" in **Neurocomputing**, 453, 2021, 151-163. (Published) (**SCI, IF: 5.719**) (https://doi.org/10.1016/j.neucom.2021.04.084)
- **A. K. Singh** and L. Kailasam, "*PILHNB: Popularity, Interests, Location used Hidden Naive Bayesian-based model for Link Prediction in Dynamic Social Networks*" in **Neurocomputing**, 2021, Elsevier. (https://doi.org/10.1016/j.neucom.2021.02.101) (Published) (**SCI, IF: 5.719**)
- **A. K. Singh,** L. Kailasam, Tribikram Pradhan and Deepti Gupta, "*Multifeature Analysis Based Influential Nodes Tracking in Dynamic Social Networks*" in **Neurocomputing**, 2021, Elsevier. (Revision Submitted) (**SCI, IF: 5.719**)

A.2 Conference Papers

- **A. K. Singh,** L. Kailasam, Tribikram Pradhan and Deepti Gupta, "*Context-aware Influential Nodes Tracking in Dynamic Social Networks*" in 25^{th} Pacific Asia Conference on Information Systems (PACIS-2021), Dubai, July 2021. (Published) (https://aisel.aisnet.org/pacis2021/177)

Appendix B

Sample Datasets

Table B.1: Example of check-in information in Gowalla dataset.

[user]	[check-in time]	[latitude]	[longitude]	[location id]
196514	2010-07-24T13:45:06Z	53.3648119	-2.2723465833	145064
196514	2010-07-24T13:44:58Z	53.360511233	-2.276369017	1275991
196514	2010-07-24T13:44:46Z	53.3653895945	-2.2754087046	376497
196514	2010-07-24T13:44:38Z	53.3663709833	-2.2700764333	98503
196514	2010-07-24T13:44:26Z	53.3674087524	-2.2783813477	1043431
196514	2010-07-24T13:44:08Z	53.3675663377	-2.278631763	881734
196514	2010-07-24T13:43:18Z	53.3679640626	-2.2792943689	207763
196514	2010-07-24T13:41:10Z	53.364905	-2.270824	1042822

Table B.2: Example of check-in information in Brightkite dataset.

[user]	[check-in time]	[latitude]	[longitude]	[location id]
58186	2008-12-03T21:09:14Z	39.633321	-105.317215	ee8b88dea22411
58186	2008-11-30T22:30:12Z	39.633321	-105.317215	ee8b88dea22411
58186	2008-11-28T17:55:04Z	-13.158333	-72.531389	e6e86be2a22411
58186	2008-11-26T17:08:25Z	39.633321	-105.317215	ee8b88dea22411
58187	2008-08-14T21:23:55Z	41.257924	-95.938081	4c2af967eb5df8
58187	2008-08-14T07:09:38Z	41.257924	-95.938081	4c2af967eb5df8
58187	2008-08-14T07:08:59Z	41.295474	-95.999814	f3bb9560a2532e
58187	2008-08-14T06:54:21Z	41.295474	-95.999814	f3bb9560a2532e
58188	2010-04-06T06:45:19Z	46.521389	14.854444	ddaa40aaa22411
58188	2008-12-30T15:30:08Z	46.522621	14.849618	58e12bc0d67e11
58189	2009-04-08T07:36:46Z	46.554722	15.646667	ddaf9c4ea22411
58190	2009-04-08T07:01:28Z	46.421389	15.869722	dd793f96a22411

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