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LIST OF PUBLICATIONS

Journal Papers

Naina Yadav, Sukomal Pal, Anil Kumar Singh, and Kartikey Singh (2022). Clus-DR: Cluster-based pre-trained model for diverse recommendation generation. Journal of King Saud University-Computer and Information Sciences, 34(8), 6385-6399. **IF-6.9, *SCI-Indexed***

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Naina Yadav, Sukomal Pal, and Anil Kumar Singh, ”Multimodal Deep Content-based Music Recommendation System Using MIDI and Lyrics (Computational Intelligence) **IF- 2.8, *SCI-Indexed* (Under Review)**

Naina Yadav, Sukomal Pal, and Anil Kumar Singh, “Diversified recommendation modeling using meta paths and graph coloring for heterogeneous information networks. (Data Mining and Knowledge Discovery) **IF- 4.8, *SCI-Indexed* (Under Review)**

6.3 Conference Paper

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