

Jack's reading list

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1 Recommendation Diversity and Group-Based Recommendation

Recommender systems [51] are useful tools that help people to filter and explore massive information. While the accuracy of recommender systems is important, many recent research indicated that focusing merely on accuracy is insufficient to fulfill the various contexts and objectives of recommender systems. My research (and this reading list) focus on two interesting topics: **Recommendation Diversity** [20] [56] [23] [10] [24] [28] [36] [21] [29] [4] [32] [45] [61] [17] [33] [60] [12] [1] [11] [53] [2] [34] [43] [47] [48] [50] [35] [44] and **Group-Based Recommendation** [31] [54] [5] [42] [30] [19] [7] [39] [46] [16] [9] [38] [25] [40] [57] [3] [6] [8] [41] [27] [52] [15] [37] [55] [59] [13] [22] [14] [26] [49] [58] [18] which should also be taken into consideration in order to extract more meaningful recommendations and fulfill user needs

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