Carelland Care	Data set type	Data set	Link (https://tinyurl.com/)	Classification	Clustering	Link Prediction	Recommendation	Regression	Other	Input setting	Used by
Margor Cape	Bookmarks										
Cancer	Biology / Chemistry	CMP CTD [131] Drug-Drug Interactions [5] Drug-Target Interactions FAERS iAF1260 [30] iJO1366 [80]	pmh-data-ch ctd-data - dt-inter drug-faers iaF1260 iJO1366-ecoli	✓ ✓ ✓		✓		✓		Het, NU, D Het, NU Het, NU Hom, NU, MC Het, (N)U Het, NU, D Het, NU, D	[120] [81] [5] [122] [5, 49, 61, 104, 135, 139] [124] [124]
Cacordation DACM [13] accordate Description Desc	Categorical	Cancer [95] Chess [93] Congress [91] Covertype [10] Letter [32] Nursery [79] Scene [12] Yeast [27]	cancer-db chess-cat-db congress-votes covertype letterrecog nursery-db scenedb yeastdataset	\ \ \ \ \ \ \ \ \	\ \ \					Hom, NU Hom, NU, A	[69, 88] [88] [88] [85] [85] [88] [96]
Amazon [cf. 7a, 128] amazon-parol amazon-pa	Co-authorship / Co-citation	ACM [113] DBLP* [34, 128] Cora* [92] Citeseer [87] PubMed* [92]	acm-data dblp-net dblp-bip aminer-dblp cora-cit-network cora-coauth citeseer pubmed-db pb-diabetes	✓ ✓ ✓ ✓	√	\ \ \ \ \	✓			Het, NU Hom/Het, NU, MC Hom/Het, NU, MC Hom, NU, MC Hom/Het, NU, MC	[36] [3, 18, 26, 29, 36, 38, 39, 45, 46, 50, 97, 98, 101, 115, 123, 124] [2, 3, 18, 31, 35, 36, 48, 50, 55, 82, 85, 97, 98, 115, 123, 125] [2, 3, 18, 26, 33, 35, 48, 50, 55, 85, 115, 123, 125] [2, 3, 18, 26, 31, 35, 48, 50, 82, 97, 123, 125]
Gowalla [19] gowalla-dan	Purchase	Amazon* [64, 76, 128] Diginetica Tmall	diginetica tmall-ijcai15				1			Hom/Het, (N)U, D, MC Hom/Het, (N)U Hom/Het, NU	[16, 17, 65, 85, 101, 110, 118, 122] [17, 25, 83, 111, 121] [83, 112, 119, 124]
Marges	GPS		gowalla-db	✓							
HG data collection [8] scholp-data	Images	Botswana FERET [84] Houses Indian Pines [4] JAFFE [24] KSC LFW-A [114] ModelNet40 [117] MS-COCO [67] NUS-WIDE [22] NTU [13] ORL [89] Pascal VOC [28] Pavia University Visual Genome [62]	botswana-img color-feret house-img indian-pines jaffe-db ksc-db-img lfw-db-img modelnet40 mscoco-db nus-wide-db http://3d.csic.ntu.edu.tw/ faces-orl pascal-voc-2007 pavia-uni visual-genome	√	✓	✓			\frac{1}{\sqrt{1}}	Hom, NU, A, MC Hom, NU, A Hom, (N)U, Dir Hom, NU, A, MC Hom, NU, A, MC Hom, NU, A, MC Hom, NU, A Hom/Het, NU, MC Hom, NU Het, NU Het, NU Hom/Het, NU, MC Hom, NU, A Hom/NU, A Hom, NU, A, MC Hom, NU, A Hom, NU, A Hom, NU, A	[71, 74, 100, 132] [51] [66, 86] [70, 73, 74, 100, 132] [51] [71, 73, 74] [51] [1, 18, 31, 35, 47, 48, 72, 77, 125] [116] [20, 116] [1, 18, 31, 35, 47, 48, 72, 77, 125] [51] [66, 116] [70, 74, 100] [116]
MDB* Indb-movie-db V V Hom/Het, NJ, MC [29, 45, 46, 58, 115] Indb-movie-decomoda Micro-Video 1.7M [15] micro-Video Movic-lens 42 movie-lens V V Hom, NU, D [43] Hom/Het, (N)U, MC, D [35, 43, 49, 53, 61, 68, 103, 104, 118, 135, 139] Movic-lens 42 movie-lens V V Hom, NU, A [96]	Miscellanea										
Review Goodreads* [109, 110] goodreads-db movie-reviews-db yelp-db	Movies	IMDB* LDOS-CoMoDa [78] Micro-Video 1.7M [15]	imdb-movie-db ldos-comoda micro-video				\ \ \ \			Hom/Het, NU, MC Het, NU Hom, NU, D	[29, 45, 46, 58, 115] [129] [43]
Movie review Yelp' movie-reviews-db yelp-db	Q&A	Yahoo* [60, 105]	yahoo-db	✓						Hom, NU, A	[96]
Social networks	Review	Movie review	movie-reviews-db			√				Het, NU	[110]
Last.FM last-fm-db ✓ Hom/Het, NU, MC [68, 101, 103, 130] Spotify [9] million-songs spotify-db ✓ Het, NU, Dir [63, 107] Text Ohsumed Reuters oh-r8-r52 ✓ Het, NU [110]	Social networks	Baidu (feed & news) [53] Foursquare [126] Douban Douban-Weibo [14] Facebook-Twitter [14] Friendster [128] LiveJournal [128] MaFengWo Sina Twitter YouTube [128] Weeplace [90]	foursquare-net douban-db friendster-net livejournal-net	✓		\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \	\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \			Hom, NU, MC Het, U Hom/Het, (N)U, MC Hom, NU Hom, NU Hom, NU Hom, NU Hom, NU Hom, NU Hom, NU, D Hom, NU, D Hom, NU, D Hom, NU Hom, NU, D	[53] [52, 126, 127] [29, 40, 41, 44, 130, 134] [14] [101] [101] [101] [54] [99] [99] [101] [134]
	Songs	Last.FM	last-fm-db				✓			Hom/Het, NU, MC	[68, 101, 103, 130]
	Text					✓					

Table 1. Most commonly used public datasets by the articles reviewed. Input setting: Hom/Het stands for homogeneous/heterogeneous, (N)U for (non-)uniform, MC for multi-channel, D for dynamic, and Dir for directed. The symbol * means that the specific dataset is available in multiple versions. † The NTU dataset is not reachable from our location.

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