Graph Representation Learning and Pre-Training

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Graphs in Society



Internet



Electrical Grid Network

Transportation

figure credit: Web 2

Graphs in Nature









Graphs are widely used for abstracting complex systems of sinteracting objects!



Food Web

Human Disease Networks

Protein-Protein Interactions

Graph Representation Learning



- Input: a graph G = (V, E)
- Output: $\mathbf{Z} \in \mathbb{R}^{|V| \times k}$, $k \ll |V|$, k-dim vector \mathbf{Z}_{v} for each node v. •

J. Malik

CVPR

How to Embed Graph Structures?

VS.

"Graph, a structure made of vertices and edges"





						0	٥							٥				
	0	0	0	0	0	12	0	11	20	127	27	0	102	147	04	0	0	0
	0	0	0	0	1	12	0	100	220	72/	225	102	727	197	P0	11	12	0
	0	0	1	0		0	41	100	250	200	200	102	200	238	200	11	12	0
		0	0	10	9	9	150	251	45	21	184	159	154	200	255	40	0	0
	10	0	0	0	0	0	145	146	3	10	0	11	124	253	255	107	0	0
	0	0	3	0	4	15	Z36	216	0	0	38	109	Z47	Z40	169	0	11	0
	1	0	2	0	0	0	253	253	23	62	224	241	255	164	0	5	0	0
	6	0	0	4	0	3	252	250	228	255	255	234	112	28	0	2	17	0
	0	2	1	- 4	0	21	255	253	251	255	172	31	8	0	1	0	0	0
	0	0	- 4	0	163	225	251	255	229	120	0	0	0	0	0	11	0	0
	0	0	21	162	255	255	254	255	126	6	0	10	14	6	0	0	9	0
	3	79	242	255	141	66	255	245	189	7	8	0	0	5	0	0	0	0
	26	221	237	98	0	67	251	255	144	0	8	0	0	7	0	0	11	0
1	25	255	141	0	87	244	255	288	3	0	0	13	0	1	0	1	0	0
1	45	248	228	116	235	255	141	34	0	11	0	1	0	0	0	1	3	0
	85	237	253	246	255	210	21	-1	0	1	0	0	6	-2	4	0	0	0
	6	23	112	157	114	32	0	0	0	0	2	0	8	0	7	0	0	0
1	00	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0



Network Embedding

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Heterogeneous Graphs



Graph Neural Networks



Graph Pre-Training and Self-Supervised Learning



Graph Pre-Training and Self-Supervised Learning



Open Data & Toolkit



Graph Representation Learning and Pre-Training



Pre-Train Graphs with Language/Image/Knowledge



Academic Graph



Microsoft Office Graph



LinkedIn Economic Graph



Facebook Entity Graph

Pre-Train for Cognitive Reasoning



Pre-Train for Cognitive Reasoning



Cause

Symptom

Treatment

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Thank you!

Papers & data & code available at <u>https://keg.cs.tsinghua.edu.cn/yuxiao/</u> yuxiaod@tsinghua.edu.cn