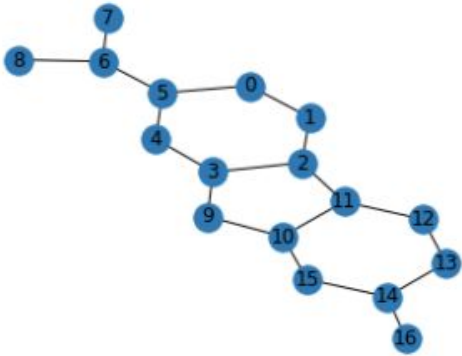


WPI

TreeRNN: Topology-Preserving Deep Graph Embedding and Learning

Yecheng Lyu, Ming Li, Xinming Huang,
Ulkuhan Guler, Patrick Schaumont, and
Ziming Zhang

Motivation



Graph



Image

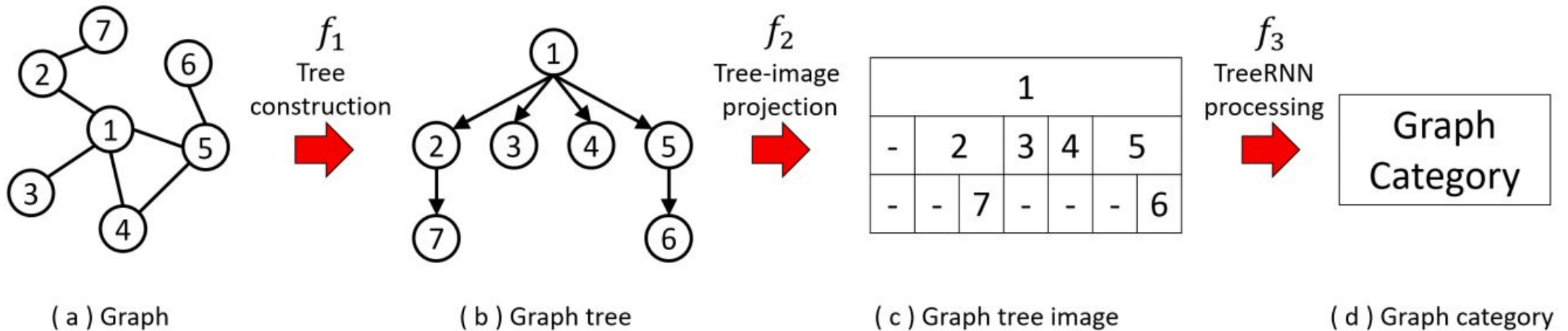


non-small cell
lung cancer ?

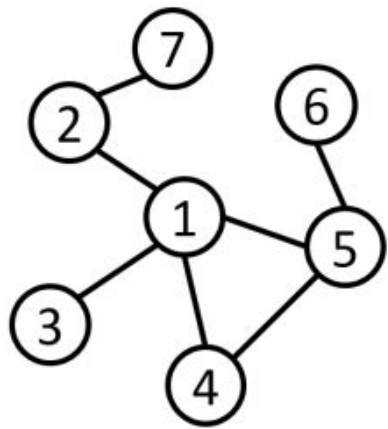
ovarian cancer ?

Classification

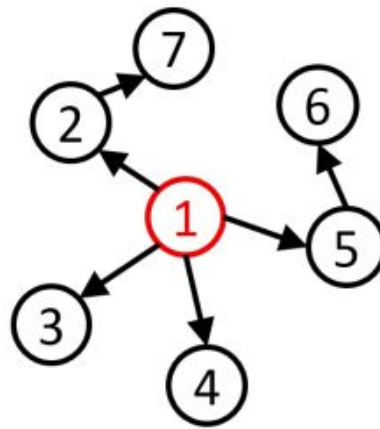
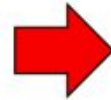
Our Solution



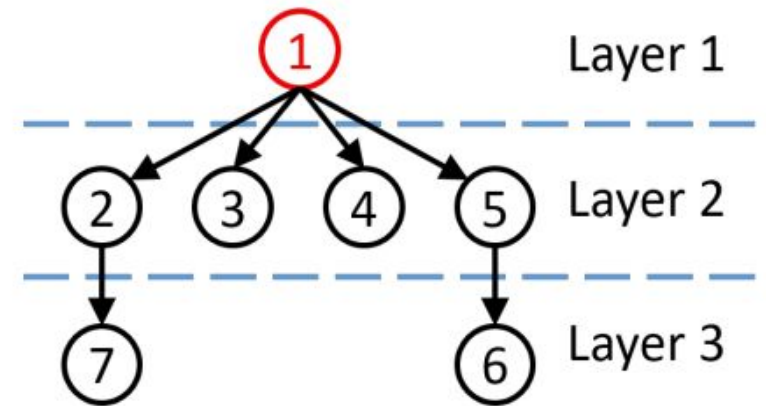
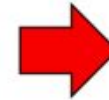
Step 1: Tree Construction from Graph



(a) Graph

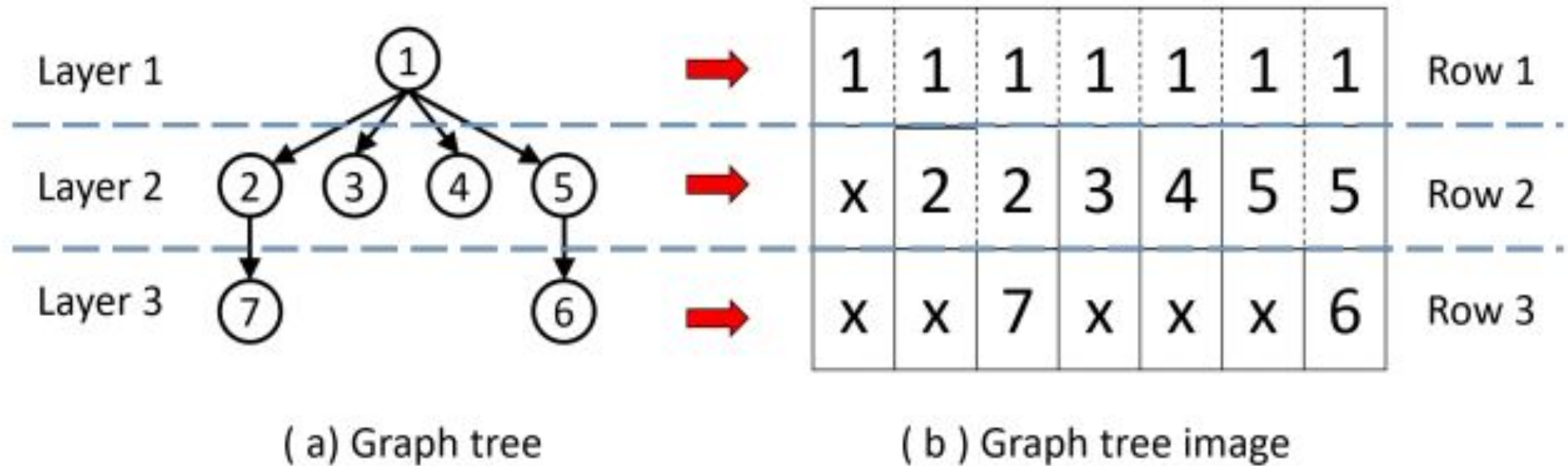


(b) Breadth first search (BFS)



(c) Graph tree

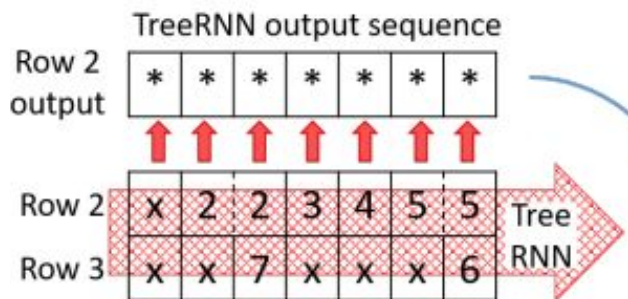
Step 2: Projection a tree to image space



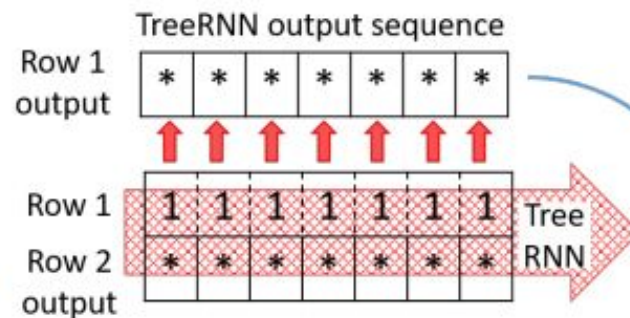
Step 3: TreeRNN Graph Classification

Row 1	1	1	1	1	1	1
Row 2	x	2	2	3	4	5
Row 3	x	x	7	x	x	6

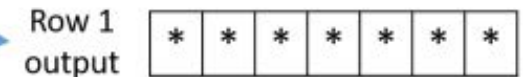
(a) Input feature map



(b) TreeRNN step on bottom two rows



(c) TreeRNN step on next two rows



(d) TreeRNN layer output

Results

TABLE II

GRAPH CLASSIFICATION RESULTS (%) IN MUTAG, PTC-MR AND NCI1. NUMBERS IN RED ARE THE BEST IN THE COLUMN, AND NUMBERS IN BLUE ARE THE SECOND BEST.

Category	Method	MUTAG	PTC-MR	NCI1
Graph Convolution	GraphConv [7]	86.1	-	76.2
	GINConv [6]	95.00 \pm 4.61	72.94 \pm 6.28	80.32 \pm 1.73
	ECConv [3]	89.44	-	83.80
	DGCNN [4]	85.83 \pm 1.66	58.59 \pm 2.47	74.44 \pm 0.47
	GIC [9]	94.44 \pm 4.30	77.64 \pm 6.98	84.08 \pm 1.07
Graph Embedding	PSCN [12]	88.95 \pm 4.37	62.29 \pm 5.68	76.34 \pm 1.68
	DDGK [11]	91.58 \pm 6.74	63.14 \pm 6.57	68.10 \pm 2.30
	WKPI [28]	85.8 \pm 2.5	62.7 \pm 2.7	87.5 \pm 0.5
	Ours	94.74 \pm 5.55	74.69 \pm 5.78	84.96 \pm 4.81

Thanks!



**WPI
VIS-LAB**



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Xinming Huang

Github Repository:

<https://github.com/YechengLyu/TreeRNN>