

L23: Graph Embeddings

+ Ethics Assessment

Nov 19, 2025

Data Mining

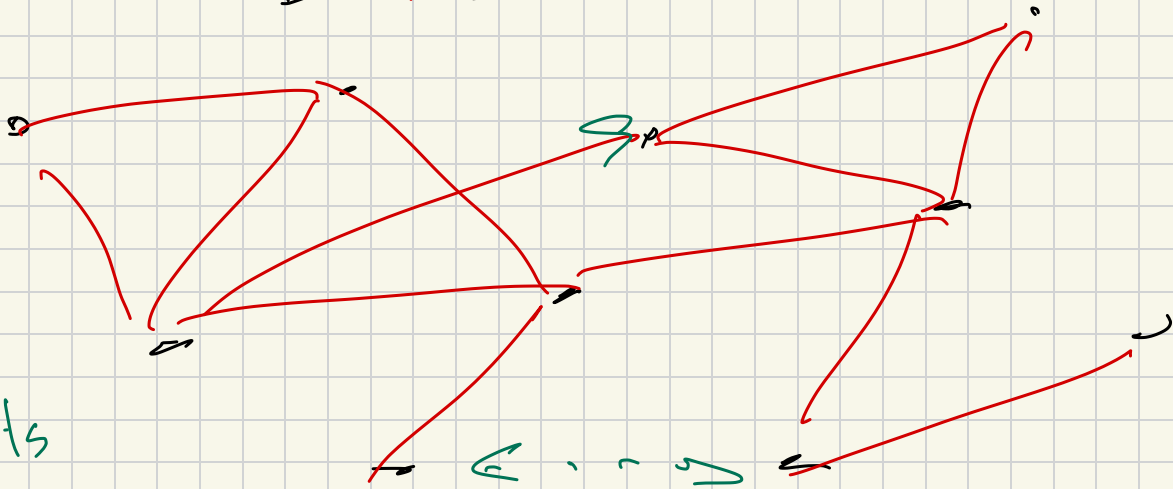
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Data graph $G = (V, E)$

$V = \text{vertices} = \text{webpages, computers, people, products}$

$E = \text{edges} = \text{connections}$



Similarity

$$S(v_1, v_2) = e_{12} \\ \in [0, 1]$$

Graph Embedding

$$G = (V, E) \Rightarrow V \in \mathbb{R}^d$$

$$\text{dist}(v_1, v_2) \approx f(E)$$

Input

list vertices, $V = \{v_1, v_2, \dots\}$

list edges $E = \{e_1 = (v_1, v_2), e_2 = (v_2, v_3), \dots\}$

$d = 2 \Rightarrow$ draw graph

$d = k \Rightarrow$ clustering classification

MDS

input

V

, distance function

$$D(v_1, v_2) = \begin{cases} 0 & \text{if edge } e=(v_1, v_2) \in E \\ 1 & \text{o.w.} \end{cases}$$

not commonly used
blunt

$$D^{(2)} \in \mathbb{R}^{n \times n}$$

$$D_{ij}^{(2)} = D(v_i, v_j)^2$$

$$\text{eigs}(D^{(2)})$$

$\Rightarrow$ eigen vectors $v_1, v_2, \dots$

project $\phi(v_i) \in \mathbb{R}^R$

$$\phi(v_1(i)), \phi(v_2(i)), \dots, \phi(v_n(i))$$

Laplacian Matrices

$$|V| = n \leq 10,000$$

D = diagonal matrix degrees

$$\begin{bmatrix} \deg(v_1) & 0 & \dots & 0 \\ 0 & \deg(v_2) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & \dots & \deg(v_n) \end{bmatrix}$$

$$L_0 = D - A$$

$$L = I - D^{-1/2} A D^{-1/2}$$

A = adjacency matrix

$$A_{ij} = \begin{cases} 1 & \text{if edge } e_{ij} \in E \\ 0 & \text{otherwise} \end{cases}$$

Graph Neural Network

DeepWalk, node2vec

Learn embedding

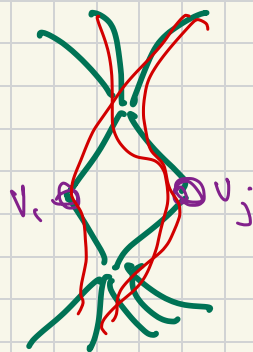
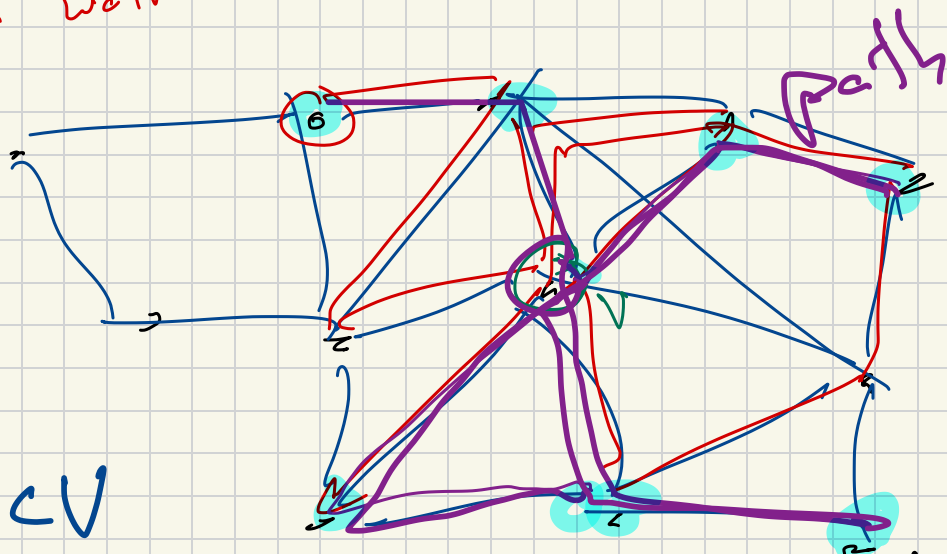
$$G = (V, E) \Rightarrow \phi(v) \in \mathbb{R}^k$$

Random Walk

Neighborhood
Length

$$L = 2$$

neighborhood
 $N(v_i, P) \subset V$



$$\text{Sim}(v_i, v_j) = \text{Similarity} \left(\bigcup_{P \in \text{Paths}} N(v_i, P), \bigcup_{P \in \text{Paths}} N(v_j, P) \right)$$

Is Proctree

Scalable

↳ optimize $\phi(v) \in \mathbb{R}^k$

with SGD

goal

$$\langle \phi(v_i), \phi(v_j) \rangle \approx \text{Sim}(N(v_i), N(v_j))$$