

L18: Random Projections

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Data Mining

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Input $X = \{x_1, x_2, \dots, x_n\} \subset \mathbb{R}^d$

goal $G = \{g_1, g_2, \dots, g_n\} \subset \mathbb{R}^R$

$\forall$ all pairs $x_i, x_j \in X$

d extremely large

≈ 100 million

$n \ll d$

$n \approx 500$

$$(1-\epsilon) \|x_i - x_j\| \leq \|g_i - g_j\| \leq (1+\epsilon) \|x_i - x_j\|$$

error tolerance

$\epsilon \in (0, 1)$

$\epsilon \approx 0.1$
or 0.01

Johnson Lindenstrauss

Random Projections

Determining

mapping $u: \mathbb{R}^d \rightarrow \mathbb{R}^k$
w/ distortion $(1 \pm \epsilon)$

Define

set

$v_1, v_2, \dots, v_k$

$v_j \in \mathbb{R}^d$

$$\|v_j\| = 1 \cdot \frac{\sqrt{d}}{\sqrt{k}}$$

$$g = u(x) = \tilde{\Pi}_{v_1, \dots, v_k}(x)$$

$$g_j = \langle x, v_j \rangle$$

ifn^g
coord of g

How do we choose v_j

generate $g \in \mathbb{R}^d$

$$g_j \sim \mathcal{N}(0, 1)$$

$$g \sim \mathcal{N}_d(0, I)$$

$$v_j = v = \frac{g}{\|g\|} \cdot \frac{\sqrt{d}}{\sqrt{k}}$$

Why does random projection work?

$$v \sim \mathcal{N}_d(0, 1)$$

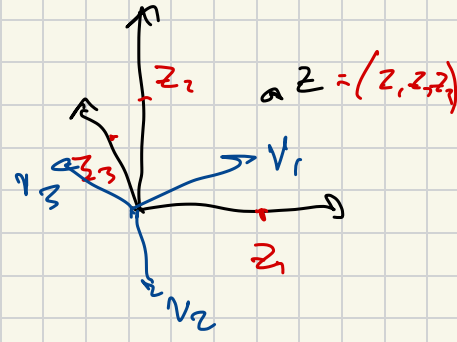
$$\bar{v} = \frac{v}{\|v\|}$$

$$x_i - x_j = z \in \mathbb{R}^d$$

$$\mathbb{E}[\langle z, \bar{v} \rangle^2] = \frac{1}{d} \|z\|^2$$

$$\|z\|^2 = \sum_{j=1}^d z_j^2$$

$$\|z\|^2 = \sum_{j=1}^d \langle v_j, z \rangle^2$$



$$\|v_j\| = 1 \quad \langle v_1, v_2 \rangle = 0$$

pick one
 $v_i = v$
at random

unbiased

$$\mathbb{E}[\langle z, \bar{v} \rangle^2] = \|z\|^2$$

$$\mathbb{E}\left[\frac{1}{d} \sum_{j=1}^d \langle z, v_j \rangle^2\right] = \|z\|^2$$

$$\mathbb{E}\left[\frac{d}{k} \|z \cdot U\|^2\right] = \|z\|^2$$

$$U = [v_1, v_2, \dots, v_k]$$

Chernoff-Hoeffding bound k R.V. X_i
 $M = \frac{1}{k} \sum_{i=1}^k X_i$ subgaussian

$$\mathbb{P}_r[|M - \mathbb{E}[M]| > \varepsilon] \leq \delta$$

$\Rightarrow$ if $k = O\left(\frac{1}{\varepsilon^2} \log \frac{1}{\delta}\right)$

Holds for any one $z = X_i - X_j$ pair.

For all distances $\delta' = \frac{\delta}{n^2}$ $n^2 \approx \# \text{ pairs}$

$$\hookrightarrow k = O\left(\frac{1}{\varepsilon^2} \log \frac{n^2}{\delta}\right) = O\left(\frac{1}{\varepsilon^2} \log \frac{n}{\delta}\right)$$