

Feb 18, 2020

$$A \geq 0$$

Last lecture: NMF

$$\|A - LR\|_F^2 + \lambda (\|L\|_F^2 + \|R\|_F^2)$$

$$\text{s.t. } L, R \geq 0$$

separable NMF: $X = WH$, $W = X(:, K)$

$$K \subseteq \{1, \dots, n\}$$

scale X to \bar{X} : column sums to 1

$$\bar{X} = \bar{X}(:, K) H$$

$$\bar{X} \Pi = \bar{X}(:, K) [I \ G] \leftarrow \text{Example of an "interpolative decomposition"}$$

$$\text{ID: } A \Pi = C [I \ T]$$

C is a subset of columns of A

At the end $A\Pi = QR$

$$\boxed{A} \boxed{\Pi} = \boxed{Q} \begin{matrix} R_1 \\ R_2 \end{matrix}$$

$$A\Pi_k \approx Q_k [R_1 \ R_2] \\ = \underbrace{Q_k R_1}_{\text{first } k \text{ cols of } A\Pi} [I \ R_1^{-1} R_2]$$

$$A\Pi_{:,1} \approx q_1 r_{11} = q_1 \|A(:,1)\|_2$$

$$A\Pi_{:,2} \approx q_1 r_{12} + q_2 r_{22}$$

$$\|A\Pi_{:,2}\|_2^2 = |r_{12}|^2 + |r_{22}|^2 \leq \|A\Pi_{:,1}\|_2^2 = |r_{11}|^2$$

$$|r_{22}|^2 \leq |r_{11}|^2 - |r_{12}|^2 \leq |r_{11}|^2$$

$|r_{k+1, k+1}|$ small \Rightarrow can truncate at k for an approx

$$\text{Also, } |r_{12}| \leq |r_{11}|$$

$A\Pi \approx C[I \ T]$ C subset of k columns of A

$$A\Pi = C[I \ R_1^{-1}R_2]$$

More stringent: $|T_{ij}| \leq 2$

PQR: some pathological where
 T has large entries

• randomized algs

CUR decomposition

$$\begin{matrix} & n \\ & \boxed{A} \\ m & \end{matrix} = \begin{matrix} & k \\ & \boxed{C} \\ m & \end{matrix} \begin{matrix} k \\ \boxed{U} \\ k \end{matrix} \begin{matrix} & n \\ & \boxed{R} \\ & \end{matrix}$$

C: columns of A

R: rows

How to select cols and rows?

Idea 1: ID/PQR on A \Rightarrow cols
 $A^T \Rightarrow$ rows

$$M^+ = (M^T M)^{-1} \cdot M^T$$

$$\min_U \|A - CUR\|_F^2$$

$$\min_U \|C^+ A - UR\|_F^2$$

$$\min_Z \|A - CZ\|_F^2$$

$$U = C^+ A R^+$$

$$Z = C^+ A$$

(Mahoney + Drineas 09)

Pre-process: $A \approx U_k \Sigma_k V_k^T$

$O(mnk)$

ColSelect(A, k, ϵ, c) $c = O(k \log k / \epsilon^2)$

(i) $\pi_j = \frac{1}{k} \|V_k^T(:, j)\|_2^2$

$O(nk)$

(ii) keep j th column w.p. $\min(1, c\pi_j)$ $O(n)$

(iii) return kept cols

CUR: $S = \text{ColSelect}(A, k, \epsilon, c)$

$O(nk)$
 $C = A(:, S)$

$T = \text{ColSelect}(A^T, k, \epsilon, c)$

$R = A(T, :)$

$U = C^+ A R^+ \quad O(|T|^2 n)$

$O(mk)$

$O(|S|^2 m)$

w.h.p. $|T|, |S|$ are $O(c) = O(k \log k / \epsilon^2)$

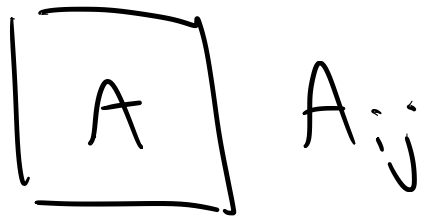
$$O(mnk)$$

Thm: $\|A - CUR\|_F$

$$\leq (2 + \epsilon) \underbrace{\|A - U_k \Sigma_k V_k^T\|_2}_{\text{OPT}}$$

OPT

Tensors are a generalization of matrices



 A_{ijk} = feature j of patient i in week k

A

$$A \approx \sum_{s=1}^r x_s y_s^T$$

$$A_{ij} \approx \sum_{s=1}^r (x_s)_i (y_s^T)_j$$

A

$$A_{ijk} \approx \sum_{s=1}^r (x_s)_i (y_s)_j (z_s)_k$$

$$A \approx \sum_{s=1}^r x_s \otimes y_s \otimes z_s$$

$$z_s = 1$$

$$\min_B \|A - B\|_F^2 \quad \text{s.t.} \quad B \text{ is rank-} r$$

rank is minimum r s.t. $B = \sum_{s=1}^r x_s \otimes y_s$

$$B = U_r \Sigma_r V_r^T$$

Tensor data: "best" low-rank approx

$$\min_{\underline{B}} \|\underline{A} - \underline{B}\|_F^2 \quad \text{s.t.} \quad \underline{B} \text{ is rank-} r$$

rank is min r s.t. $\underline{B} = \sum_{s=1}^r x_s \otimes y_s \otimes z_s$

Problem (de Silva and Lim 08)

Best low-rank approx can be ill-posed!

Might be no best rank- r approx to \underline{A}

$$\underline{A}_n = n \left(x_1 + \frac{1}{n} y_1 \right) \otimes \left(x_2 + \frac{1}{n} y_2 \right) \otimes \left(x_3 + \frac{1}{n} y_3 \right) \\ - n x_1 \otimes x_2 \otimes x_3 \quad \text{rank-2}$$

$$\lim_{n \rightarrow \infty} \underline{A}_n = \underline{A}^* = x_1 \otimes x_2 \otimes y_3 \\ + x_1 \otimes y_2 \otimes x_3 \\ + y_1 \otimes x_2 \otimes x_3 \quad \text{rank-3}$$

"border rank" = 2

A $2 \times 2 \times 2$ is rank 3

Then A does not have a best rank-2 approx

A $d_1 \times d_2 \times d_3$ $2 \leq r \leq \max(d_1, d_2, d_3)$

{ A | A do not have best rank- r approx, $r \leq r$ }

has positive volume