

CS698C 2021 August Quiz 5

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TOTAL POINTS

100 / 100

QUESTION 1

1 Problem on dimensionality reduced
embedding **100 / 100**

+ **100** Point adjustment

🗨 Excellent.

1) Given $A \in \mathbb{R}^{n \times d}$, $B \in \mathbb{R}^{d \times p}$, $C \in \mathbb{R}^{n \times p}$
 a) $X \in \mathbb{R}^{d \times p}$

Minimize $\|AX - C\|_F$

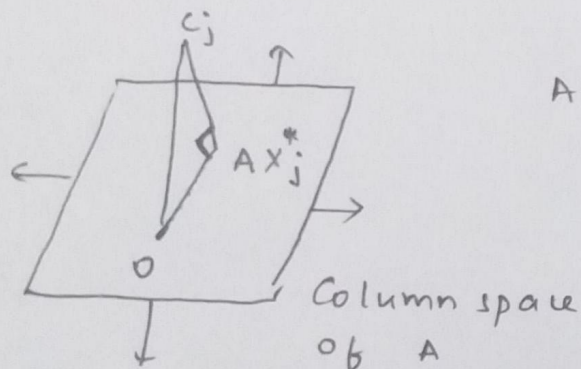
$$\exists Y \in \mathbb{R}^{d \times q}$$

$$YB = X$$

$$\|AX - C\|_F^2 = \sum_{j=1}^p \|AX_j - C_j\|_2^2$$

Consider $\min \|AX_j - C_j\|_2$ $0 \leq j < p$

This is similar to linear regression problem p times



$$\begin{aligned} \text{Arg Min}_{X_j} \|AX_j - C_j\|_2 \\ = A^- C_j \end{aligned}$$

$$\begin{aligned} A^T(AX_j^* - C_j) &= 0 \\ \text{Taking } A &= U_r \Sigma_r V_r^T \\ X_j^* &= V_r \Sigma_r^{-1} U_r^T C_j \\ &= A^- C_j \end{aligned}$$

$$\text{Argmin}_{X_j} \sum_{j=1}^p \|AX_j - C_j\|_2^2$$

$$= A^- [C_1 \dots C_p]$$

$$= A^- C$$

$$X^* = A^- C$$

$$\text{Given } YB = X$$

$$Y^* = A^- C B^-$$

$$\min \|Ax - c\|_F = \|A\bar{A}^+c\|_F \quad x^* = \bar{A}^+c$$

$$y^* = \bar{A}^+CB^+$$

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$$= \|A \underbrace{\bar{A}^+CB^+}_{y^*} - c\|_F$$

Minimum value of $\|Ax - c\|_F$

P.T.O

1 b)

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(3)

→ Embedding function Matrix is

used to solve the affine problem in lower dimension
and plug the solution in higher dimension

$$\|SAx - SB\|_F$$

$$\|Ax - B\|_F$$

→ No. of rows of the matrix depends on the distribution "D" from which it is drawn

Properties of embedding function are:

1) Both column^{space} wise embedding of A and row-wise embedding of B is done with $S \sim D_1$ and

$R \sim D_2$ matrices

$$\|A \cdot B - C\|_F \text{ to } \|SA \cdot BR - SCR\|_F$$

2) D_1 preserves colspace of A

$$P \left[\forall x \in \mathbb{R}^d \quad \|SAx\| \in (1 \pm \epsilon) \|Ax\| \right] \geq 1 - \delta$$

3) D_2 preserves rowspace of B

$$P \left[\forall y^T \in \mathbb{R}^e \quad \|y^T BR\| \in (1 \pm \epsilon/20) \|y^T B\| \right] \geq 1 - \delta$$

4) ~~Both $S \sim D_1$ $R \sim D_2$ satisfy~~

4) D_2 is an affine embedding

$$P \left[\forall y \in \mathbb{R}^{n \times e} \quad \|yBR - Ay^T R\|_F \in (1 \pm \epsilon) \|yB - Ay^T\| \right]$$

$$C^* = A A^T C$$

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$$5) P_{S/R} \left[\|S C^* R\|_F \in (1 \pm \epsilon) \|C^* R\|_F \right] \geq 1 - \delta$$

$$6) P \left[\|C^* R\|_F \in (1 \pm \epsilon) \|C^*\|_F \right] \geq 1 - \delta$$

$$7) P_{S/R} \left[\|S B^* R\|_F \in (1 \pm \epsilon) \|B^* R\|_F \right] \geq 1 - \delta$$

$$B^* = A A^T C (B^T B - I)$$

$$8) P \left[\|B^* R\|_F \in (1 \pm \epsilon) \|B^*\|_F \right] \geq 1 - \delta$$

9) ~~For~~ Inner products

$$P \left| \left(S(Ax - Ax^*) R \right)_j, S(Ax^* R - CR)_j \right|$$

$$= (A \cup B - Ax^*) R_j (Ax^* - C) R_j^T$$

$$\leq \epsilon \|A \cup B - Ax^*\| \|Ax^* - C\| \geq 1 - \delta$$

$$10) P \left[\left(S A_j \right)^T (S C^* R_{j'}) \leq \epsilon \|A_j\| \|C^* R_{j'}\| \right] \geq 1 - \delta$$

$$11) P \left[\|C^* R\|_F \in (1 \pm \epsilon) \|C^*\|_F \right] \geq 1 - \delta$$

$$12) P \left[\|S A B^*\|_F \in (1 \pm \epsilon) \|A B^*\| \right] \geq 1 - \delta$$

Given,

$$A \in \mathbb{R}^{n \times d}$$

$$X \in \mathbb{R}^{d \times p}$$

$$C \in \mathbb{R}^{n \times p}$$

$$B \in \mathbb{R}^{2 \times p}$$

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$$\text{Min } \|AX - C\|_F$$

$$\exists Y \in \mathbb{R}^{d \times 2} \text{ s.t. } YB = X$$

$$\text{Min } \| \underbrace{A}_{n \times d} \underbrace{Y}_{d \times 2} \underbrace{B}_{2 \times p} - C \|_F$$

$n \times p$

Dimensionality reduction on both n and p is done by using random matrices S and R respectively. $S \sim D_1$, $R \sim D_2$

The problem $\|SAYBR - SCR\|_F$ needs to be solved

Required to prove: $\|SAYBR - SCR\|_F \overset{\text{embeddings}}{\uparrow} \overset{\text{belongs to}}{\downarrow} (1 \pm \epsilon) \|AYB - C\|_F$

$$\|SAYBR - SCR\|_F^2$$

$$S \quad k \times n$$

$$R \quad p \times l$$

$$= \|SAYBR - SCR - SA \underbrace{A^T C R}_{X^*} + SA \underbrace{A^T C R}_{X^*}\|_F^2$$

$$= \|SA(YB - X^*)R - SA X^* R - SCR + SA X^* R\|_F^2$$

$$= \|SA(YB - X^*)R + SA X^* R - SCR\|_F^2$$

$$= \|SA(YB - X^*)R\|_F^2 + \|SA X^* R - SCR\|_F^2$$

$$+ 2 \text{tr} \left[(SA(YB - X^*)R)^T (SA X^* R - SCR) \right]$$

$$X^* = A^T C$$

opening $\|\cdot\|_F$ norm

$$\begin{aligned} & \|A + B\|_F^2 \\ &= \|A\|_F^2 + \|B\|_F^2 \\ &+ 2 \sum_j A_j B_j \text{tr}(I_R) \end{aligned}$$

$$= \| SA(BR - X^*R) + SA(Y^*B - X^*B)R \|_F^2 \quad \begin{cases} X^* = A^*K \\ Y^* = A^*B \end{cases}$$

$$+ \| SAX^*R - SCR \|_F^2 + 2\text{tr}((S(A(YB - X^*B))R)^T (SAX^*R - SCR))$$

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$$= \| SA(Y - Y^*)BR + SA(Y^*B - X^*B)R \|_F^2$$

$$+ \| SAX^*R - SCR \|_F^2 + 2\text{tr}((S(A(YB - X^*B))R)^T (SAX^*R - SCR))$$

$$= \| SA(Y - Y^*)BR \|_F^2 + \| SA(Y^*B - X^*B)R \|_F^2 \quad \left. \begin{array}{l} \text{opening} \\ \| \|_F \end{array} \right\}$$

$$+ 2\text{tr}[(SA(Y - Y^*)BR)^T (SA(Y^*B - X^*B)R)]$$

$$+ \| SAX^*R - SCR \|_F^2 + 2\text{tr}[(S(A(YB - X^*B))R)^T (SAX^*R - SCR)]$$

Rearranging terms.

$$= \underbrace{\| SA(Y - Y^*)BR \|_F^2}_{\text{term 1}} + \underbrace{\| SA(Y^*B - X^*B)R \|_F^2}_{\text{term 2}} + \underbrace{\| SAX^*R - SCR \|_F^2}_{\text{term 3}}$$

$$+ 2\text{tr}[(SA(Y - Y^*)BR)^T (SA(Y^*B - X^*B)R)] \leftarrow \text{term 4}$$

$$+ 2\text{tr}[\cancel{SA} [S(A(YB - X^*B))R]^T (SAX^*R - SCR)] \leftarrow \text{term 5}$$

(1)

term 1

$$\| SA(Y - Y^*)BR \|_F^2 \Rightarrow \sum_{j=1}^n \| SA(Y - Y^*)BR_j \|_2^2 \quad \text{Using}$$

$$(1 \pm \epsilon) \sum_{j=1}^n \| A(Y - Y^*)BR_j \|_2^2 \leq \| SA \|_F^2 = (1 \pm \epsilon) \| A \|_F^2$$

$$\Downarrow$$

$$= (1 \pm \epsilon) \| A(Y - Y^*)BR \|_F^2$$

$$= (1 \pm \epsilon)^2 \| A(Y - Y^*)B \|_F^2$$

$$= (1 \pm \epsilon) \| A(Y - Y^*)B \|_F^2 \quad \text{--- (2)}$$

$$\| AR \|_F^2 = (1 \pm \epsilon) \| A \|_F^2$$

$$(1 \pm \epsilon)^2 \approx (1 \pm \epsilon)$$

Similarly

term 2

$$\|SA(Y^*B - X^*)R\|_F^2$$

$$= (1 \pm \epsilon) \|A(Y^*B - X^*)\|_F^2$$

(3)

Assuming

$$\|SA(Y^*B - X^*)R\|_F^2$$

$$= (1 \pm \epsilon) \|A(Y^*B - X^*)\|_F^2$$

$$\|A(Y^*B - X^*)R\|_F^2$$

$$= (1 \pm \epsilon) \|A(Y^*B - X^*)\|_F^2$$

term 3

$$\|SA X^* R - SCR\|_F^2$$

$$= (1 \pm \epsilon) \|A X^* - C\|_F^2 \quad (4)$$

term 4

$$2 \text{tr} \left[(SA(Y - Y^*)BR)^T (SA(Y^*B - X^*)R) \right]$$

Consider the term

$$(A(Y - Y^*)B)^T (A(Y^*B - X^*))$$

$$= (A(Y - Y^*)B)^T (A(X^* \bar{B} B - X^*))$$

$$= \underbrace{(A(Y - Y^*)B)^T}_{\text{rowspan } B} \underbrace{(-AX^*(I - \bar{B}B))}_{\text{orthogonal to rowspan } B} = 0$$

Preserving inner product

Using $X^* \bar{B} = Y^*$

$\sum_j A_j^T B_j = \text{tr}(A^T B)$

$$\left((SA(Y - Y^*)BR)^T (SA(Y^*B - X^*)R) \right) - (A(Y - Y^*)B)^T (A(Y^*B - X^*))$$

$$\leq O(\epsilon) \|A(Y - Y^*)B\|_F$$

for some $0 \leq j < k$

$$((A(Y - Y^*)B)R_j)^T (A(Y^*B - X^*)R_j)$$

$$= R_j^T (A(Y - Y^*)B)^T (A(Y^*B - X^*)R_j) = 0 \quad (5)$$

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2 ||| 0 || (5)

$$\begin{cases} X^* = A \bar{C} \\ Y^* = A \bar{C} \bar{B} \\ X^* \bar{B} = Y^* \end{cases}$$

S $k \times n$
R $p \times 1$

$$2 \operatorname{tr} \left[(S A (Y - Y^*) B R)^T (S A (Y^* B - X^*) R) \right]$$

$$= 2 \sum_{j=1}^l (S A (Y - Y^*) B R)_j^T (S A (Y^* B - X^*) R)_j$$

$$= 2 \sum_{j=1}^l \left| (S A (Y - Y^*) B R)_j^T (S A (Y^* B - X^*) R)_j - \underbrace{(A (Y - Y^*) B R)_j^T (A (Y^* B - X^*) R)_j}_0 \right|$$

proven above

preserving inner product

$$\leq 2 \sum_{j=1}^l \epsilon \| (A (Y - Y^*) B R)_j \|_2 \| (A (Y^* B - X^*) R)_j \|_2$$

$$\leq \epsilon \sum_{j=1}^l \| (A (Y - Y^*) B R)_j \|_2 \sum_{j=1}^l \| (A (Y^* B - X^*) R)_j \|_2$$

$$\leq \epsilon \| A (Y - Y^*) B R \|_F \| A (Y^* B - X^*) R \|_F$$

$$\leq \epsilon \| A (Y - Y^*) B \|_F \| A (Y^* B - X^*) \|_F \quad \text{--- (6)}$$

Similarly

term 5

$$\operatorname{tr} \left[(S (A Y B - A X^*) R)^T (S A X^* R - S C R) \right]$$

$$\leq \epsilon \| A Y B - A X^* \|_F \| A X^* - C \|_F \quad \text{--- (7)}$$

Using Cauchy Schwarz inequality

$$2 \left| \sum_{i=1}^n u_i \bar{v}_i \right|^2 \leq \sum_{i=1}^n \|u_i\|^2 \sum_{i=1}^n \|v_i\|^2$$

$$2 \epsilon = \epsilon$$

Using

$$\|A R\|_F \leq (1 \pm \epsilon) \|A\|_F$$

Hence by combining the results of term(1) term(2) term(3) term(4) term(5) i.e.,

Answer

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(9)

$$Y^* = \bar{A}C$$

$$Y^* = \bar{A}CB^*$$

$$\|SA(Y - Y^*)BR\|_F^2 = (1 \pm \epsilon) \|A(Y - Y^*)B\|_F^2 \quad \text{--- (a)}$$

$$\|SA(Y^*B - X^*)R\|_F^2 = (1 \pm \epsilon) \|A(Y^*B - X^*)\|_F^2 \quad \text{--- (b)}$$

$$\|SA X^*R - SCR\|_F^2 = (1 \pm \epsilon) \|AX^* - C\|_F^2 \quad \text{--- (c)}$$

$$2\text{tr} \left[(SA(Y - Y^*)BR)^T (SA(Y^*B - X^*)R) \right] \leq \epsilon \|A(Y - Y^*)B\|_F \|A(Y^*B - X^*)\|_F \quad \text{--- (d)}$$

$$2\text{tr} \left[S(A Y^*B - AX^*)R^T (SA X^*R - SCR) \right] \leq \epsilon \|A Y^*B - AX^*\|_F \|AX^* - C\|_F \quad \text{--- (e)}$$

$$\begin{aligned} \|SA YBR - SCR\|_F^2 &= \|SA(Y - Y^*)BR\|_F^2 + \|SA(Y^*B - X^*)R\|_F^2 \\ &\quad + \|SA X^*R - SCR\|_F^2 + 2\text{tr} \left[(SA(Y - Y^*)BR)^T (SA(Y^*B - X^*)R) \right] \\ &\quad + 2\text{tr} \left[S(A Y^*B - AX^*)R^T (SA X^*R - SCR) \right] \end{aligned}$$

$$\begin{aligned} &\leq (1 \pm \epsilon) \|A(Y - Y^*)B\|_F^2 + (1 \pm \epsilon) \|A(Y^*B - X^*)\|_F^2 + (1 \pm \epsilon) \|AX^* - C\|_F^2 \\ &\quad + \epsilon \|A(Y - Y^*)B\|_F \|A(Y^*B - X^*)\|_F + \epsilon \|A Y^*B - AX^*\|_F \|AX^* - C\|_F \end{aligned}$$

Using $\frac{a^2 + b^2}{2} \geq \sqrt{a^2 b^2}$

Using $AM \geq GM$

$$\leq (1+\epsilon) \|A(Y-Y^*)B\|_F^2 + (1+\epsilon) \|A(Y^*B-X^*)\|_F^2 + (1+\epsilon) \|AX^*-C\|_F^2$$

$$+ \epsilon \left(\frac{1}{2}\right) \|A(Y-Y^*)B\|_F^2 + \epsilon \left(\frac{1}{2}\right) \|A(Y^*B-X^*)\|_F^2 + \epsilon \|AX^*-C\|_F^2$$

$$\leq (1\pm\epsilon) \left[\|A(Y-Y^*)B\|_F^2 + \|A(Y^*B-X^*)\|_F^2 + \|AX^*-C\|_F^2 \right]$$

$$Y^* = A^+CB^+$$

$$X = A^+C$$

Consider

$$\begin{aligned} \|AYB - C\|_F^2 &= \|AYB - AY^*B + AY^*B - C\|_F^2 \\ &= \|AYB - AY^*B\|_F^2 + \|AY^*B - C\|_F^2 \\ &= \|AYB - AY^*B\|_F^2 + \|AY^*B - AX^* + AX^* - C\|_F^2 \\ &= \|AYB - AY^*B\|_F^2 + \|A(Y^*B - X^*)\|_F^2 + \|AX^* - C\|_F^2 \end{aligned}$$

trace terms are \perp so zero

$$\leq (1\pm\epsilon) \|AYB - C\|_F^2$$

Now Solution

$$Y = \frac{SA^+BR}{t}$$

$$t = (SA)^+ S^+ R (SB)^+$$

1c) $\|A \Upsilon B - C\|_F$ reduced to

$$\|S A \Upsilon B R - S C R\|_F \quad \begin{matrix} S \sim D_1 \\ R \sim D_2 \end{matrix}$$

S	$k \times d$	A	$n \times d$	B	$q \times p$
R	$p \times d$	Υ	$d \times q$	C	$n \times p$

1) Consider dimensions of S matrix $k \times n$.

S is in column space of A .

$$\| \underbrace{S A \Upsilon B R}_{k_1} - \underbrace{S C R}_{k_2} \|_F$$

$$\|S A k_1 - S k_2\|_F.$$

We need to preserve column space of A
and columns of C .

Considering $S_{ij} \sim N(0, \frac{1}{k})$
 $R_{ij} \sim N(0, \frac{1}{p})$

Using γ net argument for A
and p times J-L lemma for C .

$$k \geq 0 \left(\frac{d}{\epsilon^2} \log \frac{1}{\delta} + \frac{1}{\epsilon^2} \log \frac{p}{\delta} \right).$$

2) Consider dimensions of R matrix $p \times d$.
 R is in row space of $S A \Upsilon B$. q rows.
Preserving row space of $S A \Upsilon B$ gives
 $O\left(\frac{q}{\epsilon^2} \log \frac{1}{\delta}\right).$

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Preserving n rows of B

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$$\frac{1}{\epsilon^2} \log \frac{n}{\delta}$$

$$l = O\left(\frac{2}{\epsilon^2} \log \frac{1}{\delta} + \frac{1}{\epsilon^2} \log \frac{n}{\delta}\right)$$

Optimal solution

$$\text{of } \|SAYBR - SCR\|_F$$

$$= \|SAXR - SCR\|_F$$

~~X~~
Taking $\|SAX - SC\|_F \Rightarrow \|S \underbrace{AYBR}_X - \underbrace{SCR}_C\|_F$ ^{ignoring R}

$$\hat{Y} = (SA)^T S \underbrace{C}_{n \times p}$$

$$\|S \underbrace{AYBR}_X - \underbrace{SCR}_C\|_F$$
 ^{ignoring S}

Taking $\|$

$$\hat{Y} = R^T (CR)^T (AYB)$$

$$\hat{Y} = (SA)^T S C (CR)^T (AYB)$$

~~Time complexity is~~

Using $\|XAB - B\|$
 $B^T A$

1 Problem on dimensionality reduced embedding 100 / 100

+ 100 Point adjustment

Excellent.