Recitation - Week 4

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- 1. Suppose $v_1, \ldots, v_m \in \mathbb{R}^k$ are m non-zero orthogonal vectors. Prove that they are linearly independent.
- 2. Let U be the subspace of \mathbb{R}^n with orthonormal basis u_1, \ldots, u_k .
 - i. Prove that the projection of $x \in \mathbb{R}^n$ can be expressed as $P_U(x) = \langle u_1, x \rangle u_1 + \langle u_2, x \rangle u_2 + \dots + \langle u_k, x \rangle u_k$ (Use the fact that the orthonormal basis for a subspace of \mathbb{R}^n can be extended to obtain an orthonormal basis for \mathbb{R}^n)
 - ii. Prove that $||P_U(x)|| \le ||x||$
 - iii. Prove that $x P_U(x)$ is orthogonal to the subspace U
 - iv. Show that the linear transformation $P_U: \mathbb{R}^n \to \mathbb{R}^n$ satisfies $P_U^2 = P_U$ and $P_U^T = P_U$

1. Suppose $v_1, \ldots, v_m \in \mathbb{R}^k$ are m non-zero orthogonal vectors. Prove that they are linearly independent. Solution:

To check if they are linearly independent, we need to calculate $\alpha \in \mathbb{R}^m$ such that

$$\alpha_1 v_1 + \dots + \alpha_m v_m = 0$$

For any index $i \in [1, m]$,

$$<\alpha_1 v_1 + \ldots + \alpha_m v_m, v_i > = < 0, v_i > = 0$$

$$\Rightarrow \alpha_1 < v_1, v_i > + \cdots + \alpha_i < v_i, v_i > + \cdots + \alpha_m < v_m, v_i \ge 0$$

$$\Rightarrow \alpha_i < v_i, v_i > = 0$$

Since $v_i \neq 0$, $\langle v_i, v_i \rangle \neq 0 \Rightarrow \alpha_i = 0$ for all $i \in [1, m]$

2. Let U be the subspace of \mathbb{R}^n with orthonormal basis $u_1, ..., u_k$.

- i. Prove that the projection of $x \in \mathbb{R}^n$ can be expressed as $P_U(x) = \langle u_1, x \rangle u_1 + \langle u_2, x \rangle u_2 + \dots + \langle u_k, x \rangle u_k$ (Use the fact that the orthonormal basis for a subspace of \mathbb{R}^n can be extended to obtain an orthonormal basis for \mathbb{R}^n)
- ii. Prove that $||P_U(x)|| \le ||x||$
- iii. Prove that $x P_U(x)$ is orthogonal to the subspace U
- iv. Show that the linear transformation $P_U: \mathbb{R}^n \to \mathbb{R}^n$ satisfies $P_U^2 = P_U$ and $P_U^T = P_U$

Solution: The orthonormal basis of U is $u_1, ..., u_k$. This can be extended to form a basis $u_1, ..., u_n$ of \mathbb{R}^n

i.
$$x = \alpha_1 u_1 + \dots + \alpha_k u_k$$
 where $\alpha_i = \langle u_i, x \rangle \forall i \in [1, n]$

For projection, we need to solve

$$P_U(x) = argmin_{y \in U} ||x - y||$$

Since $y \in U$,

$$y = \beta_1 u_1 + \dots + \beta_k u_k$$
$$||x - y|| = ||(\alpha_1 - \beta_1)u_1 + \dots + (\alpha_k - \beta_k)u_k + \alpha_{k+1}u_{k+1} + \dots + \alpha_n u_n||$$
$$\Rightarrow ||x - y|| = \sqrt{(\alpha_1 - \beta_1)^2 + \dots + (\alpha_k - \beta_k)^2 + \alpha_{k+1}^2 + \dots + \alpha_n^2}$$

This will be minimized when $\beta_i = \alpha_i \forall i \in [1, k]$

$$\Rightarrow P_U(x) = \alpha_1 u_1 + \dots + \alpha_k u_k$$
$$\Rightarrow P_U(x) = \langle u_1, x \rangle u_1 + \dots + \langle u_k, x \rangle u_k$$

$$\text{ii.} \left| |P_U(x)| \right| = \sqrt{\alpha_1^2 + \dots + \alpha_k^2} \text{ and } \left| |x| \right| = \sqrt{\alpha_1^2 + \dots + \alpha_k^2 + \dots + \alpha_n^2} \text{ where } \alpha_i = < u_i, x >$$
$$\Rightarrow P_U(x) \le ||x||$$

iii. To show that a vector is orthogonal to a subspace U, it is enough to show that it is orthogonal to the Basis of U

$$x - P_U(x) = \alpha_{k+1}u_{k+1} + \dots + \alpha_n u_n$$

For any element u_i of the basis u_1, \ldots, u_k ,

$$< u_i, x - P_U(x) > = < u_i, \alpha_{k+1}u_{k+1} + \dots + \alpha_n u_n >$$

$$= \alpha_{k+1} < u_i, u_{k+1} > + \dots + \alpha_n < u_i, u_n >$$

$$= \alpha_{k+1}. 0 + \dots + \alpha_n. 0$$
 (given that u_1, \dots, u_n are orthonormal)
$$= 0$$

Thus, $\langle x - P_U(x), u_i \rangle = 0 \forall i \in [1, ..., k]$

iv. $P_U = VV^T$ where V = matrix with columns $u_1, ..., u_k$ $P_U^2 = P_U P_U = VV^T VV^T$

Now,
$$V^T V = \begin{bmatrix} \cdots & u_1^T & \cdots \\ & \vdots & \\ \cdots & u_k^T & \cdots \end{bmatrix} \begin{bmatrix} \vdots & & \vdots \\ u_1 & \cdots & u_k \\ \vdots & & \vdots \end{bmatrix} = \begin{bmatrix} u_1^T u_1 & \cdots & u_1^T u_n \\ \vdots & & \vdots \\ u_n^T u_1 & \cdots & u_n^T u_n \end{bmatrix} = \begin{bmatrix} 1 & \cdots & 0 \\ \vdots & & \vdots \\ 0 & \cdots & 1 \end{bmatrix} = Id_{k \times k}$$
$$\Rightarrow P_U^2 = VIV^T = VV^T = P_U$$

 $P_U^T = (VV^T)^T = VV^T = P_U$

- 1. Which of the following are an inner product for $x, y \in \mathbb{R}^3$:
 - *i.* $f(x, y) = x_1y_2 + x_2y_3 + x_3y_1$ *ii.* $f(x, y) = x_1^2y_1^2 + x_2^2y_2^2 + x_3^2y_3^2$

iii. $f(x, y) = x_1y_1 + x_2y_2$

- 2. Let $x = (cos\theta_1, sin\theta_1) \in \mathbb{R}^2$ and $y = (cos\theta_2, sin\theta_2) \in \mathbb{R}^2$ be two vectors on unit circle (||x|| = 1 = ||y||). What does $x^T y$ represent geometrically?
- 3. For $x, y \in \mathbb{R}^n$, prove that $||x + y|| \le ||x|| + ||y||$ (Triangular inequality). When does ||x + y|| = ||x|| + ||y||?
- 4. For any $A \in \mathbb{R}^{m \times n}$ and $x \in \mathbb{R}^{n}$, show that

$$||Ax|| \le ||x|| \sqrt{\sum_{i=1}^{m} \sum_{j=1}^{n} A_{ij}^2}$$

1. Which of the following are an inner product for $x, y \in \mathbb{R}^3$:

i.
$$f(x, y) = x_1y_2 + x_2y_3 + x_3y_1$$

ii. $f(x, y) = x_1^2 y_1^2 + x_2^2 y_2^2 + x_3^2 y_3^2$

iii.
$$f(x, y) = x_1y_1 + x_2y_2$$

Solution:

- i. Not an inner product. Take x = y = (1, -1, 0); f(x, x) = -1 < 0
- ii. Not an inner product: $f(\alpha x, y) = \alpha^2 f(x, y) \neq \alpha f(x, y)$
- iii. Not an inner product: Take $x = y = (0,0,-1) \Rightarrow f(x,x) = 0$ and $x \neq 0$

2. Let $x = (cos\theta_1, sin\theta_1) \in \mathbb{R}^2$ and $y = (cos\theta_2, sin\theta_2) \in \mathbb{R}^2$ be two vectors on unit circle (||x|| = 1 = ||y||). What does $x^T y$ represent geometrically?

Solution:

$$x^{T}y = \cos\theta_{2}\cos\theta_{1} + \sin\theta_{2}\sin\theta_{1}$$
$$\Rightarrow x^{T}y = \cos(\theta_{2} - \theta_{1})$$

3. For $x, y \in \mathbb{R}^n$, prove that $||x + y|| \le ||x|| + ||y||$ (Triangular inequality). When does ||x + y|| = ||x|| + ||y||? Solution:

$$||x + y||^{2} = \langle x + y, x + y \rangle = \langle x, x \rangle + \langle y, y \rangle + \langle x, y \rangle + \langle y, x \rangle$$
$$\Rightarrow ||x + y||^{2} = ||x||^{2} + ||y||^{2} + 2 \langle x, y \rangle$$

Using Cauchy-Schwarz inequality,

$$\Rightarrow ||x + y||^{2} \le ||x||^{2} + ||y||^{2} + 2||x||||y||$$
$$\Rightarrow ||x + y||^{2} \le (||x|| + ||y||)^{2}$$

Since norms are always positive,

$$||x+y|| \le \left||x|\right| + \left||y|\right|$$

For equality to hold, we should have equality in Cauchy-Shwarz i.e. $||x||||y|| = \langle x, y \rangle$ which is true iff x and y are collinear

4. For any $A \in \mathbb{R}^{m \times n}$ and $x \in \mathbb{R}^n$, show that

$$||Ax|| \le ||x|| \sqrt{\sum_{i=1}^{m} \sum_{j=1}^{n} A_{ij}^2}$$

Solution: Let a_1^T , ..., a_m^T be the columns of A

$$||Ax||^2 = (a_1^T x)^2 + (a_2^T x)^2 + \dots + (a_m^T x)^2$$

Using Cauchy-Schwarz,

$$\begin{aligned} \left| |Ax| \right|^{2} &\leq \left| |a_{1}| \right|^{2} ||x||^{2} + \dots + \left| |a_{m}| \right|^{2} ||x||^{2} \\ &\Rightarrow \left| |Ax| \right| \leq \left(\left| |a_{1}| \right|^{2} ||x||^{2} + \dots + \left| |a_{m}| \right|^{2} \right)^{0.5} \\ &\Rightarrow \left| |Ax| || \leq \left| |x| \right| \left(\left| |a_{1}| \right|^{2} + \dots + \left| |a_{m}| \right|^{2} \right)^{0.5} \end{aligned}$$

Since $a_{1}^{T}, \dots, a_{m}^{T}$ are rows of A, $\left| |a_{i}| \right|^{2} = a_{i,1}^{2} + \dots + a_{i,n}^{2}$ assuming $a_{i} = [a_{i,1}, \dots, a_{i,n}]$
 $\Rightarrow \left| |Ax| || \leq \left| |x| \right| \left(\sum_{i=1}^{m} \sum_{j=1}^{n} A_{ij}^{2} \right)^{0.5} \end{aligned}$