Lecture 8

Norm function

- $||\cdot||$ is called the norm function and $||X||$ is called the norm of X.
- $||\cdot||_2$ is called euclidean norm. $||X||_2 = \sqrt{\sum_{i=1}^n |X_i|^2}$

If
$$
P = \begin{bmatrix} \uparrow & \uparrow & \cdots & \uparrow \\ X_1 & X_2 & \cdots & X_n \\ \downarrow & \downarrow & \cdots & \downarrow \end{bmatrix}
$$
 then $P^T = \begin{bmatrix} \leftarrow & X_1 & \rightarrow \\ \leftarrow & X_2 & \rightarrow \\ \vdots & \vdots & \vdots \\ \leftarrow & X_n & \rightarrow \end{bmatrix}$.
\n
$$
\Rightarrow P^T P = \begin{bmatrix} \leftarrow & X_1 & \rightarrow \\ \leftarrow & X_2 & \rightarrow \\ \vdots & \vdots & \vdots \\ \leftarrow & X_n & \rightarrow \end{bmatrix} \begin{bmatrix} \uparrow & \uparrow & \cdots & \uparrow \\ X_1 & X_2 & \cdots & X_n \\ \downarrow & \downarrow & \cdots & \downarrow \end{bmatrix}
$$
\n
$$
= \begin{bmatrix} X_1^T X_1 & X_1^T X_2 & \cdots & X_1^T X_n \\ X_2^T X_1 & X_2^T X_2 & \cdots & X_2^T X_n \\ \vdots & \vdots & \ddots & \vdots \\ X_n^T X_1 & X_n^T X_2 & \cdots & X_n^T X_n \end{bmatrix}
$$
\nBut when A is real symmetric matrix, $X_i^T X_j = 0$ for $i \neq j$.
\n
$$
\Rightarrow P^T P = \begin{bmatrix} X_1^T X_1 & 0 & \cdots & 0 \\ 0 & X_2^T X_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & X_n^T X_n \end{bmatrix}
$$
\nNow,
\n
$$
X_1^T X_1 = [x_1 \ x_2 \ \cdots \ x_n] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1^2 + x_2^2 + \cdots + x_n^2
$$
\n
$$
\Rightarrow X_1^T X_1 = ||X_1||^2
$$

But when A is real symmetrix matrix, $X_i^T X_j = 0$ for $i \neq j$.

$$
\Rightarrow P^{T}P = \begin{bmatrix} X_{1}^{T}X_{1} & 0 & \cdots & 0 \\ 0 & X_{2}^{T}X_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & X_{n}^{T}X_{n} \end{bmatrix}
$$

Now,

$$
\begin{bmatrix}\n\vdots \\
\leftarrow & X_n & \rightarrow\n\end{bmatrix}\n\begin{bmatrix}\n\downarrow & \downarrow & \downarrow & \downarrow\n\end{bmatrix}
$$
\n
$$
= \begin{bmatrix}\nX_1^T X_1 & X_1^T X_2 & \cdots & X_1^T X_n \\
X_2^T X_1 & X_2^T X_2 & \cdots & X_2^T X_n \\
\vdots & \vdots & \ddots & \vdots \\
X_n^T X_1 & X_n^T X_2 & \cdots & X_n^T X_n\n\end{bmatrix}
$$
\n
$$
\text{symmetry matrix, } X_i^T X_j = 0 \text{ for } i \neq j.
$$
\n
$$
\Rightarrow P^T P = \begin{bmatrix}\nX_1^T X_1 & 0 & \cdots & 0 \\
0 & X_2^T X_2 & \cdots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
0 & 0 & \cdots & X_n^T X_n\n\end{bmatrix}
$$
\n
$$
\therefore X_1^T X_1 = [x_1 \quad x_2 \quad \cdots \quad x_n] \begin{bmatrix}\nx_1 \\
x_2 \\
\vdots \\
x_n\n\end{bmatrix} = x_1^2 + x_2^2 + \cdots + x_n^2
$$
\n
$$
\Rightarrow X_1^T X_1 = ||X_1||^2
$$
\nor

$$
||X_1||=\sqrt{X_1^TX_1}
$$

For making $P^T P = I$ (in order to make matrix P orthogonal) we modify X as $\frac{X}{||X||}$ called the normalized vector P.

Thus $P = \begin{bmatrix} X_1 & X_2 & \cdots & X_n \\ \frac{\prod X_i}{\prod X_i} & \frac{\prod X_i}{\prod X_i} & \cdots & \frac{\prod X_n}{\prod X_i} \end{bmatrix}$. \perp ↑ ↑ ↑ $\frac{X_1}{||X_1||}$ $\frac{X_2}{||X_2||}$... $\frac{X_n}{||X_n||}$ ↓ ↓ ↓ \vert

We will get $P^T P = I$ now as $X_1^T X_1 = ||X_1||^2$. Now we can have $P^{-1} = P^T$. ⎢⎣ output and a search of the search of th

$$
P^{-1}AP=D\Rightarrow P^TAP=D
$$

This way diagonalization process of a matrix can be made simple.

#semester-1 #mathematics #matrices