

Group 22 - HW04

April 5, 2021

Exercise 3.6 You are given

$$W = \begin{bmatrix} w_1 & 0 & \dots & 0 \\ 0 & w_2 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & w_N \end{bmatrix}$$

where, for all $n \in \{1, \dots, N\}$, we have $w_n > 0$. Show that

$$\langle x, z \rangle_w = x^T W z$$

is an inner product on R^N . What happens if, for all $n \in \{1, \dots, N\}$, we have instead $w_n > 0$?

Solution We need to prove the following things:

- $\langle x, x \rangle_w \geq 0, \forall x \in R^N$
- $\langle x, x \rangle_w = 0 \iff x = 0$
- $\langle x, z \rangle_w = \langle z, x \rangle_w$
- $\langle \kappa z + \lambda y, x \rangle_w = \kappa \langle z, x \rangle_w + \lambda \langle y, x \rangle_w$
- We define, $X = [x_1 \dots x_N]$

$$\langle X, X \rangle_W = X^T W X = \begin{bmatrix} x_1 & \dots & x_N \end{bmatrix} \begin{bmatrix} w_1 & 0 & \dots & 0 \\ 0 & w_2 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & w_N \end{bmatrix} \begin{bmatrix} x_1 \\ \dots \\ x_N \end{bmatrix} =$$

$$= \begin{bmatrix} x_1 w_1 & \dots & x_N w_N \end{bmatrix} \begin{bmatrix} x_1 \\ \dots \\ x_N \end{bmatrix}^T = x_1^2 w_1 + \dots + x_N^2 w_N = \sum_{i=1}^N x_i^2 w_i \geq 0, \forall x \in R^N$$

- For the second property,

$$\langle X, X \rangle_W = \dots = \sum_{i=1}^N x_i^2 w_i = 0, \forall x = 0$$

- For the third property, we will use a property of the dot product ($a * b = ab^T$)

$$\langle X, Z \rangle_W = X^T W Z = X * (W Z) = (W Z) * X = (W Z)^T X = Z^T W^T X$$

but, $W^T = W$ so,

$$\langle X, Z \rangle_W = Z^T W X = \langle Z, X \rangle_W$$

- For the fourth property,

$$\langle \kappa z + \lambda y, x \rangle_w = (\kappa Z + \lambda Y)^T W X = (\kappa Z^T + \lambda Y^T) W X = \kappa Z^T W X + \lambda Y^T W X =$$

$$\kappa \langle Z, X \rangle_w + \lambda \langle Y, X \rangle_w$$

So, given that all properties are proven, $\langle x, z \rangle_w = x^T W z$ is an inner product on R^N .

Now if for all $n \in \{1, \dots, N\}$, we have instead $w_n > 0$, then the second property would not fulfill, since it would be valid to say that

$$\langle x, z \rangle_w = 0 \Leftrightarrow x = 0$$

but it would be wrong to say that

$$\langle x, z \rangle_w = 0 \Rightarrow x = 0$$

So, if for all $n \in \{1, \dots, N\}$, we have instead $w_n > 0$, then the $\langle x, z \rangle_w = x^T W z$ is NOT an inner product on R^N

Exercise 3.7 Continuing from the above, show that,

$$\langle X, Z \rangle_W = X^T W Z$$

is a norm on R^N . What happens if, for all $n \in \{1, \dots, N\}$, we have instead $w_n > 0$?

Solution We need to prove the following things:

- $\langle x, x \rangle \geq 0, \forall x \in R^N$
- $\langle x, x \rangle = 0 \Leftrightarrow x = 0$
- $\langle kx, kx \rangle = |k| \langle x, x \rangle$
- $\|U + V\|_W^2 \leq (\|U\|_W^2 + \|V\|_W^2)^2$
- For the first property,

$$\|X\|_W^2 = \langle X, X \rangle_W = X^T W X = \dots = \sum_{i=1}^N x_i^2 w_i \geq 0, \forall x \in R^N$$

- For the second property,

$$\|X\|_W^2 = \langle X, X \rangle_W = X^T W X = \dots = \sum_{i=1}^N x_i^2 w_i = 0, \forall x = 0$$

- For the third property,

$$\|kX\|_W = (kX)^T W (kX) = \dots = \sum_{i=1}^N (kx_i)^2 w_i = \sum_{i=1}^N (kx_i)^2 \sum_{i=1}^N w_i =$$

$$\sqrt{k^2 \left(\sum_{i=1}^N w_i \right) \|X\|^2} = |k| \|X\|_W$$

- For the fourth property,

$$\begin{aligned} \|U + V\|_W^2 &= (U + V)^T W (U + V) = (U^T + V^T) W (U + V) = \\ &= U^T W U + U^T W V + V^T W U + V^T W V \end{aligned}$$

On the other hand,

$$(\|U\|_W^2 + \|V\|_W^2)^2 = U^T W U + V^T W V + 2\|U\| \|V\|$$

So we need to show that $2\|U\| \|V\| \geq U^T W U + V^T W V$

But

$$U^T W U + V^T W V = U^T W U + U^T W U = 2U^T W U$$

So we need to show that $\|U\| \|V\| \geq U^T W V$ which is true because of the Cauchy-Schwarz inequality.

Now if for all $n \in \{1, \dots, N\}$, we have instead $w_n \geq 0$, then the second property would not fulfill since $\langle x, z \rangle_w = 0$ does not mean that $x = 0$.

Exercise 3.8 Suppose A is an $M \times N$ matrix (with $M > N$), b is an $M \times 1$ vector and w is an $M \times 1$ vector with strictly positive components. Find the matrix \hat{A} which minimizes

$$\sum_{m=1}^M w_m \left(\sum_{n=1}^N A_{mn} x_n - b_m \right)^2$$

with respect to A .

Solution Firstly,

$$\sum_{m=1}^M w_m \left(\sum_{n=1}^N A_{mn} x_n - b_m \right)^2 = \sum_{m=1}^M w_m (Ax - b)_m^2$$

We define the MxM table W that has diagonal elements, the elements of vector w_n

$$W = \begin{bmatrix} w_1 & 0 & \dots & 0 \\ 0 & w_2 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & w_M \end{bmatrix}$$

Then,

$$\sum_{m=1}^M w_m (Ax - b)_m^2 = \|Ax - b\|_W^2$$

So, the same way with the norm $\|u\|_w$ from exercise 3.7,

$$J(A) = \|Ax - b\|_W^2 = (Ax - b)^T W (Ax - b)$$

In order to minimize J, we will take the derivative,

$$\nabla_A J(A) = \nabla_A = \nabla_A (Ax - b)^T W (Ax - b) = \frac{\partial((Ax - b)^T W (Ax - b))}{\partial A}$$

$$\nabla_A J(A) = 2W(Ax - b)x^T$$

So, we can find the \hat{A} that minimizes $J(A)$ from the equation,

$$J(\hat{A}) = 0 \Rightarrow 2W(\hat{A}x - b)x^T = 0 \Rightarrow W(\hat{A}x - b)x^T = 0$$

The table W is inversable, since $w_n > 0$ and it is a diagonal table, so,

$$\det(W) = w_1 w_2 \dots w_n > 0$$

So,

$$W(\hat{A}x - b)x^T = 0 \Rightarrow W^{-1}W(\hat{A}x - b)x^T = 0 \Rightarrow \hat{A}xx^T - bx^T = 0 \Rightarrow \hat{A}xx^T = bx^T$$

The multiplication xx^T is 1x1 and non-zero, so the $(xx^T)^{-1}$ exists. So,

$$\hat{A} = bx^T (xx^T)^{-1}$$

Exercise 3.11 Suppose A is an MN matrix (with $M > N$) and b is an M1 vector. Consider the following iteration

$$x^{(0)} = 0$$

$$x^{(t+1)} = x^{(t)} - \delta A^T (Ax^{(t)} - b)$$

Show that:

- if $x^{(t+1)} = x^{(t)}$ then $x^{(t)} = \hat{x} = (A^T A)^{-1} A^T b$
- there exists some δ_0 (find it) such that $\delta \in (0, \delta_0) \Rightarrow \lim_{t \rightarrow \infty} x^{(t)} = \hat{x}$

Solution

- When $x^{(t+1)} = x^{(t)}$ then,

$$\begin{aligned}\delta A^T(Ax^{(t)} - b) = 0 &\Rightarrow A^T(Ax^{(t)} - b) = 0 \Rightarrow A^T Ax^{(t)} - A^T b = 0 \Rightarrow \\ A^T Ax^{(t)} &= A^T b \Rightarrow x^{(t)} = (A^T A)^{-1} A^T b\end{aligned}$$

- For the second question,

$$\begin{aligned}x^{(t+1)} &= x^{(t)} - \delta A^T(Ax^{(t)} - b) \Rightarrow \\ x^{(t+1)} &= x^{(t)} - \delta A^T Ax^{(t)} + \delta A^T b \Rightarrow \\ x^{(t+1)} &= x^{(t)}[I - \delta A^T A] + \delta A^T b \Rightarrow \\ \lim_{t \rightarrow \infty} x^{(t+1)} &= \lim_{t \rightarrow \infty} x^{(t)}[I - \delta A^T A] + \lim_{t \rightarrow \infty} \delta A^T b \Rightarrow \\ \lim_{t \rightarrow \infty} x^{(t)} &= \lim_{t \rightarrow \infty} (x^{(t)})[I - \delta A^T A] + \delta A^T b \Rightarrow \\ \lim_{t \rightarrow \infty} x^{(t)}[I - I + \delta A^T A] &= \delta A^T b \Rightarrow \\ \lim_{t \rightarrow \infty} x^{(t)}[\delta A^T A] &= \delta A^T b\end{aligned}$$

So, when δ is non-zero

$$\begin{aligned}\lim_{t \rightarrow \infty} x^{(t)}[A^T A] &= A^T b \Rightarrow \\ \lim_{t \rightarrow \infty} x^{(t)} &= (A^T A)^{-1} A^T b \Rightarrow \\ \lim_{t \rightarrow \infty} x^{(t)} &= \hat{x}, \forall \delta \in R^*\end{aligned}$$

So, δ should be greater than zero, $\delta > 0$.

Exercise 3.12 Suppose that we have $1 \times N$ vectors $a_1 a_2 \dots a_n$ and scalars $b_1 b_2 \dots b_n$. Now, for every $m \in N$ define

$$A^{(m)} = [a_1 a_2 \dots a_m]$$

$$x = [x_1 x_2 \dots x_N]$$

$$b^{(m)} = [b_1 b_2 \dots b_m]$$

Let \hat{x} be the vector that minimizes $\|A^{(m)}x - b^{(m)}\|^2$. Consider the following iterations.

$$G^{(N)} = (A^{(N)})^T A^{(N)}, G^{(m+1)} = G^{(m)} + a_m^T a_m$$

$$h^{(N)} = (h^{(N)})^T b^{(N)}, h^{(m+1)} = h^{(m)} + a_m^T b_m$$

$$x^{(N)} = (G^{(N)})^{-1} h^{(N)}, x^{(m+1)} = (G^{(m)})^{-1} h^{(m)}$$

Show that, for every $m \in N$, \hat{x} minimizes $\|A^{(m)}x - b^{(m)}\|^2$. Give an interpretation

Solution From theory, we know that the \hat{x} that minimizes $\|Ax - b\|^2$ is

$$\hat{x} = (A^T A)^{-1} A^T b$$

For $m = \{N, N + 1\}$
 $m = N$

$$G^{(N+1)} = G^{(N)} + a_{N+1}^T a_N$$

$$h^{(N+1)} = h^{(N)} + a_{N+1}^T a_N$$

$m = N + 1$

$$G^{(N+2)} = G^{(N+1)} + a_{N+2}^T a_{N+2} = G^{(N)} + a_{N+1}^T a_{N+1} + a_{N+2}^T a_{N+2}$$

$$h^{(N+2)} = h^{(N+1)} + a_{N+2}^T b_{N+2} = h^{(N)} + a_{N+1}^T b_{N+1} + a_{N+2}^T b_{N+2}$$

For $m = k, k \in [N, \infty]$

$$G^{(k+1)} = G^{(k)} + a_{k+1}^T a_{k+1} = (A^{(N)})^T A^{(N)} + a_{N+1}^T a_{N+1} + \dots + a_{k+1}^T a_{k+1}$$

$$h^{(k+1)} = h^{(k)} + a_{k+1}^T a_{k+1} = (A^{(N)})^T b^{(N)} + a_{N+1}^T b_{N+1} + \dots + a_{k+1}^T b_{k+1}$$

But

$$(A^{(N)})^T A^{(N)} = a_1^T a_1 + \dots + a_N^T a_N$$

$$(A^{(N)})^T b^{(N)} = a_1^T b_1 + \dots + a_N^T b_N$$

So,

$$G^{(m+1)} = a_1^T a_1 + \dots + a_{m+1}^T a_{m+1} = (A^{(m+1)})^T A^{(m+1)}$$

$$h^{(m+1)} = a_1^T b_1 + \dots + a_{m+1}^T b_{m+1} = (A^{(m+1)})^T b^{(m+1)}$$

So,

$$x^{(m+1)} = (G^{(m+1)})^{-1} h^{(m+1)} \Rightarrow$$

$$x^{(m+1)} = [(A^{(m+1)})^T A^{(m+1)}]^{-1} (A^{(m+1)})^T b^{(m+1)}$$

which shows that for every $m \in N$, \hat{x} minimizes $\|A^{(m)}x - b^{(m)}\|^2$.