

Math Camp: Problem Set 2 Solutions

Due Monday, August 26th

1. Show that the norm satisfies the triangle inequality: for any $x, y \in \mathbb{R}^n$:

$$\|x + y\| \leq \|x\| + \|y\|$$

Solution: This follows from Cauchy Schwarz:

$$\begin{aligned}\|x + y\|^2 &= \langle x + y, x + y \rangle \\ &= \langle x, x \rangle + 2\langle x, y \rangle + \langle y, y \rangle \\ &= \|x\|^2 + 2\langle x, y \rangle + \|y\|^2 \\ &\leq \|x\|^2 + 2\|x\|\|y\| + \|y\|^2 \\ &= (\|x\| + \|y\|)^2\end{aligned}$$

2. The Euclidean distance between two points $x, y \in \mathbb{R}^n$ is defined as $d(x, y) = \|x - y\|$. Show that the above result implies the following triangle inequality: for any $x, y, z \in \mathbb{R}^n$:

$$d(x, y) \leq d(x, z) + d(z, y)$$

Solution: We can use our standard trick of adding 0:

$$\begin{aligned}d(x, y) &= \|x - y\| \\ &= \|x - z + z - y\| \\ &\leq \|x - z\| + \|z - y\| \\ &= d(x, z) + d(z, y)\end{aligned}$$

3. Give an example of two matrices A and B such that A, B are non-zero, but $AB = 0$.

Solution: (Many possible answers). One example is $A = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ and $B = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$.

4. Let A be an $m \times n$ matrix and B a $n \times m$ matrix. Show that $\text{tr}(AB) = \text{tr}(BA)$.

Solution: For this we need to use the definition of matrix multiplication:

$$\begin{aligned}
 \text{tr}(AB) &= \sum_{i=1}^m (AB)_{ii} \\
 &= \sum_{i=1}^m \sum_{j=1}^n a_{ij} b_{ji} \\
 &= \sum_{j=1}^n \sum_{i=1}^m b_{ji} a_{ij} \\
 &= \sum_{j=1}^n (BA)_{jj} \\
 &= \text{tr}(BA)
 \end{aligned}$$

5. Find the rank of the following matrices

• $\begin{pmatrix} 1 & 3 & 4 \\ 2 & 0 & 1 \end{pmatrix}$

Solution: Subtracting twice the first row from the second gives

$$\begin{pmatrix} 1 & 3 & 4 \\ 0 & -6 & -7 \end{pmatrix}$$

This is in row-echelon form, so the rank is the number of nonzero rows; namely, 2.

• $\begin{pmatrix} 2 & 1 & 3 & 7 \\ -1 & 4 & 3 & 1 \\ 3 & 2 & 5 & 11 \end{pmatrix}$

Solution: Adding twice the second row to the first row and three times the second row to the third row gives:

$$\begin{pmatrix} 0 & 9 & 9 & 9 \\ -1 & 4 & 3 & 1 \\ 0 & 14 & 14 & 14 \end{pmatrix}$$

The first and third rows are linearly dependent; subtracting $14/9$ times the first row from the third gives

$$\begin{pmatrix} 0 & 9 & 9 & 9 \\ -1 & 4 & 3 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

By flipping the first and second rows we put the matrix in row-echelon form, so its rank is 2.

6. Let A and B be $n \times n$ matrices with $AB = I$. Show that $BA = I$. (Hint: what can we conclude about the rank of B ?)

Solution: We claim B is rank n . Note $Bx = 0 \Rightarrow ABx = 0 \Rightarrow x = 0$, so the columns of B are linearly independent, and therefore form a basis for \mathbb{R}^n .

Let $x \in \mathbb{R}^n$. Since the columns of B are a basis for \mathbb{R}^n , there exists an y such that $By = x$. We therefore see

$$BAx = B \underbrace{AB}_I y = By = x$$

Thus $BAx = x$ for every vector x , so $BA = I$ (convince yourself that this last step is true).

7. Let X be an $m \times n$ matrix with $m > n$. Show that $\text{rank}(X'X) = n$ iff $\text{rank}(X) = n$.

Solution: Suppose X is rank n . Since $m > n$, that implies the columns of X are linearly independent.

Suppose $X'X\beta = 0$. Multiplying on the left by β' we see:

$$\begin{aligned}\beta'X'X\beta &= 0 \\ (\Leftrightarrow)(X\beta)'X\beta &= 0 \\ (\Leftrightarrow)\|X\beta\|^2 &= 0 \\ (\Leftrightarrow)X\beta &= 0,\end{aligned}$$

which happens iff $\beta = 0$ since X is full rank. Therefore $X'X\beta = 0$ only has the trivial solution, so $X'X$ is full rank.

Now suppose $X'X$ is rank n . Then $X'X\beta = 0$ only has the trivial solution. Suppose $X\beta = 0$. Multiplying on the left by X' gives $X'X\beta = 0$, so $\beta = 0$. Therefore the columns of X are linearly independent, so X is full rank.

8. An $n \times n$ matrix A is said to be idempotent if $A^2 = A$. Let X be an $m \times n$ matrix such that $X'X$ is invertible. Show that $M = I_m - X(X'X)^{-1}X'$ is idempotent.

Solution: This is just a direct computation:

$$\begin{aligned}M^2 &= (I_m - X(X'X)^{-1}X')(I_m - X(X'X)^{-1}X') \\ &= I_m - 2X(X'X)^{-1}X' + X \underbrace{(X'X)^{-1}X'X}_{I_n} (X'X)^{-1}X' \\ &= I_m - X(X'X)^{-1}X' \\ &= M\end{aligned}$$

9. Let A be an idempotent matrix. Show that the eigenvalues of A must be 0 or 1.

Solution: Suppose λ is an eigenvalue of an idempotent matrix A . Then $Ax = \lambda x$ for some vector x . We calculate A^2x two ways:

$$\begin{aligned}A^2x &= Ax = \lambda x \\ A^2x &= A(Ax) = A(\lambda x) = \lambda Ax = \lambda^2 x\end{aligned}$$

The first line follows from idempotency, while the second follows from the vector that x is an eigenvector of A . Thus $\lambda = \lambda^2$, so $\lambda = 0$ or 1 .

10. Let V_1 and V_2 be vector subspaces of \mathbb{R}^n .

- Is $V_1 \cap V_2$ a vector subspace of \mathbb{R}^n ?

Solution: Yes. First note that $0 \in V_1$ and $0 \in V_2$, so $0 \in V_1 \cap V_2$.

Now we show closure under vector addition. Suppose $v, w \in V_1 \cap V_2$. Since $v, w \in V_1$ and V_1 is closed under vector addition, $v + w \in V_1$. By the same logic $v + w \in V_2$, so $v + w \in V_1 \cap V_2$.

Finally, we show closure under scalar multiplication. Let $v \in V_1 \cap V_2$ and $\alpha \in \mathbb{R}$. Then $\alpha v \in V_1$ because V_1 is closed under scalar multiplication. Similarly, $\alpha v \in V_2$, so $\alpha v \in V_1 \cap V_2$.

- Is $V_1 \cup V_2$ a vector subspace of \mathbb{R}^n ?

Solution: Not necessarily. Let V_1 be the x -axis in \mathbb{R}^2 and V_2 be the y -axis. Then $V_1 \cup V_2$ is not closed under vector addition, since $\begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ is not in $V_1 \cup V_2$.

- Define $V_1 + V_2 = \{v_1 + v_2 | v_1 \in V_1, v_2 \in V_2\}$. Is $V_1 + V_2$ a vector subspace of \mathbb{R}^n ?

Solution: Yes. Since $0 \in V_1$ and V_2 , $0 = 0 + 0$ is in $V_1 + V_2$.

To show closure under vector addition, let $v, w \in V_1 + V_2$. By definition, $v = v_1 + v_2$ where $v_1 \in V_1$ and $v_2 \in V_2$. Similarly $w = w_1 + w_2$. Then $v + w = (v_1 + w_1) + (v_2 + w_2)$. Since V_1 and V_2 are closed under vector addition, $v_1 + w_1 \in V_1$ and $v_2 + w_2 \in V_2$, meaning $v + w \in V_1 + V_2$.

Finally, we show closure under scalar multiplication. Let $v \in V_1 + V_2$. Again, we can write $v = v_1 + v_2$. Thus $\alpha v = \alpha v_1 + \alpha v_2$, which - since V_1 and V_2 are closed under scalar multiplication - is in $V_1 + V_2$.

11. Let A be an $m \times n$ matrix. Show that $\ker(A)$ is a vector subspace of \mathbb{R}^n .

Solution: We first note that $A0 = 0$, so $0 \in \ker(A)$. Now we show closure under vector addition. Let $x, y \in \ker(A)$. Then $A(x + y) = Ax + Ay = 0$, so $x + y \in \ker(A)$. Finally, we show closure under scalar multiplication. Let $\alpha \in \mathbb{R}$ and $x \in \ker(A)$. Then $A(\alpha x) = \alpha Ax = \alpha 0 = 0$, so $\alpha x \in \ker(A)$.

12. Find the eigenvalues and associated eigenvectors of the following matrices. Diagonalize these matrices.

- $\begin{pmatrix} 1 & 3 \\ 2 & 0 \end{pmatrix}$

Solution: The characteristics polynomial of the matrix is:

$$(1 - \lambda)(-\lambda) - 6 = \lambda^2 - \lambda - 6 = (\lambda + 2)(\lambda - 3)$$

Therefore the eigenvalues of this matrix are -2 and 3 . Since A has two distinct eigenvalues, it is diagonalizable.

The eigenvector associated with -2 satisfies $3x_1 + 3x_2 = 0$, or $x_1 = -x_2$, so an the eigenvector is $\begin{pmatrix} -1 \\ 1 \end{pmatrix}$ (or any scalar multiple of this).

The eigenvector associated with 3 satisfies $-2x_1 + 3x_2 = 0$, or $x_1 = \frac{3}{2}x_2$, so an eigenvector is $\begin{pmatrix} 1.5 \\ 1 \end{pmatrix}$.

To diagonalize A , we form P by stacking the eigenvectors of A and D by forming the diagonal matrix with the eigenvalues of A . Thus:

$$A = PDP^{-1} = \frac{2}{5} \begin{pmatrix} 1.5 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 3 & 0 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ -1 & 1.5 \end{pmatrix}$$

- $\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}$

Solution: First note this matrix is symmetric, so it is orthogonally diagonalizable. The characteristic polynomial of this matrix is:

$$P(\lambda) = (1 - \lambda)^3 - (1 - \lambda)$$

We can factor this as $P(\lambda) = \lambda(1 - \lambda)(\lambda - 2)$, so the eigenvalues of A are 0, 1, 2.

The eigenvector associated with 0 satisfies $x_1 + x_3 = 0$ and $x_2 = 0$, so any multiple of $\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ will do.

The eigenvector associated with 1 satisfies $x_1 + x_3 = x_1$ and $x_1 + x_3 = x_3$, so $x_1 = x_2 = 0$, so it is a multiple of $\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$.

The eigenvector associated with 2 satisfies $x_1 + x_3 = 2x_1$, so $x_1 = x_3$. It also satisfies $2x_2 = x_2$, so $x_2 = 0$, so it is a multiple of $\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$.

Define P as the 3×3 matrix where these vectors have been normalized to length 1. Since A is orthogonally diagonalizable,

$$A = P'DP$$

where $D = \text{diag}(0, 1, 2)$.

13. • Let A be an $n \times n$ square matrix. Show that A' has the same eigenvalues as A .

Solution: Note $\det(A - \lambda I) = \det((A - \lambda I)') = \det(A' - \lambda I)$, so λ is an eigenvalue of A if and only if λ is an eigenvalue of A' .

- Show that if $\lambda \neq 0$ is an eigenvalue of an invertible matrix A , λ^{-1} is an eigenvalue of A^{-1} .

Solution: Let $\lambda \neq 0$ be an eigenvalue of A . Then $Ax = \lambda x$ for some nonzero vector x . Multiplying on the left by A^{-1} we see $x = \lambda A^{-1}x$, so $A^{-1}x = \lambda^{-1}x$, so λ^{-1} is an eigenvalue of A^{-1} (with the same eigenvector associated to λ for A).

14. Let A be a symmetric, invertible $n \times n$ matrix. Show that A^{-1} is symmetric.

Solution: Let A be symmetric, and suppose $AB = I$. Transposing, we see $B'A' = I$. Since A is symmetric, $B'A = I$. From question 5, we know that B and B' are inverses for A , and by the uniqueness of the inverse, $B = B'$, so B is symmetric. That is, A^{-1} is symmetric.