

Inversion Homework #1

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Appendix A

Exercise 4

Let

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 2 & 1 & 3 \\ 4 & 6 & 7 & 11 \end{bmatrix} \tag{1}$$

Find bases for $\mathcal{N}(\mathbf{A})$, $\mathcal{R}(\mathbf{A})$, $\mathcal{N}(\mathbf{A}^T)$, and $\mathcal{R}(\mathbf{A}^T)$. What are the dimensions of the four subspaces?

Solution:

To find $\mathcal{N}(\mathbf{A})$, we solve the system of equations $\mathbf{A}\mathbf{x} = 0$,

$$\mathbf{A}\mathbf{x} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 2 & 1 & 3 \\ 4 & 6 & 7 & 11 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$
 (2)

Put the system of equations into an augmented matrix,

$$\begin{bmatrix}
1 & 2 & 3 & 4 & 0 \\
2 & 2 & 1 & 3 & 0 \\
4 & 6 & 7 & 11 & 0
\end{bmatrix}$$
(3)

and then find the reduced row echelon form (RREF),

$$\begin{bmatrix}
1 & 0 & -2 & -1 & 0 \\
0 & 1 & \frac{5}{2} & \frac{5}{2} & 0 \\
0 & 0 & 0 & 0 & 0
\end{bmatrix}$$
(4)

We can find that

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 2x_3 + x_4 \\ -\frac{5}{2}x_3 - \frac{5}{2}x_4 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 2 \\ -\frac{5}{2} \\ 1 \\ 0 \end{bmatrix} x_3 + \begin{bmatrix} 1 \\ -\frac{5}{2} \\ 0 \\ 1 \end{bmatrix} x_4$$
 (5)

So,

$$\mathcal{N}(\mathbf{A}) = \mathbf{space} \begin{pmatrix} \begin{bmatrix} 2 \\ -\frac{5}{2} \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 \\ -\frac{5}{2} \\ 0 \\ 1 \end{bmatrix} \end{pmatrix}, \tag{6}$$

and the dimension of $\mathcal{N}(\mathbf{A})$ is 2.

To find $\mathcal{R}(\mathbf{A})$, the equations becomes $\mathbf{A}\mathbf{x} = \mathbf{b}$,

$$\mathbf{A}\mathbf{x} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 2 & 1 & 3 \\ 4 & 6 & 7 & 11 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \mathbf{b}$$
 (7)

Because the \mathbf{RREF} of \mathbf{A} is

$$\begin{bmatrix} 1 & 0 & -2 & -1 \\ 0 & 1 & \frac{5}{2} & \frac{5}{2} \\ 0 & 0 & 0 & 0 \end{bmatrix}$$
 (8)

We can write **b** as a linear combination of the first two columns of **A**:

$$\mathbf{b} = x_1 \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} + x_2 \begin{bmatrix} 2 \\ 2 \\ 6 \end{bmatrix}, \tag{9}$$

i.e.

$$\mathcal{R}(\mathbf{A}) = \mathbf{space} \begin{pmatrix} \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \\ 6 \end{bmatrix} \end{pmatrix}, \tag{10}$$

and the dimension of $\mathcal{R}(\mathbf{A})$ is 2.

To find $\mathcal{N}(\mathbf{A}^T)$ and $\mathcal{R}(\mathbf{A}^T)$, we first calculate the **RREF** of \mathbf{A}^T .

$$\mathbf{A}^T = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 2 & 6 \\ 3 & 1 & 7 \\ 4 & 3 & 11 \end{bmatrix},\tag{11}$$

RREF of A is
$$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$
 (12)

So,

$$\mathbf{x} = \begin{bmatrix} -2\\ -1\\ 0 \end{bmatrix} x_3,\tag{13}$$

i.e.

$$\mathcal{N}(\mathbf{A}^{T}) = \mathbf{space} \begin{pmatrix} \begin{bmatrix} -2\\ -1\\ 0 \end{bmatrix} \end{pmatrix},$$

$$\mathcal{R}(\mathbf{A}^{T}) = \mathbf{space} \begin{pmatrix} \begin{bmatrix} 1\\ 2\\ 3\\ 4 \end{bmatrix} \begin{bmatrix} 2\\ 2\\ 1\\ 3 \end{bmatrix} \end{pmatrix}.$$
(14)

And the dimension of $\mathcal{N}(\mathbf{A}^T)$ is 1, the dimension of $\mathcal{R}(\mathbf{A}^T)$ is 2.

Exercise 10

Show that if $\mathbf{x} \perp \mathbf{y}$, then

$$\|\mathbf{x} + \mathbf{y}\|_{2}^{2} = \|\mathbf{x}\|_{2}^{2} + \|\mathbf{y}\|_{2}^{2}.$$
 (15)

Solution:

Exercise 11

In this exercise, we will derive the formula (A.88) for the 1-norm of a matrix. Begin with the optimization problem

$$\|\mathbf{A}\|_1 = \max_{\|\mathbf{x}\|_1 = 1} \|\mathbf{A}\mathbf{x}\|_1. \tag{16}$$

(a) Show that if $\|\mathbf{x}\|_1 = 1$, then

$$\|\mathbf{A}\mathbf{x}\|_{1} \le \max_{j} \sum_{i=1}^{m} |A_{i,j}|.$$
 (17)

Solution:

(b) Find a vector \mathbf{x} such that $\|\mathbf{x}\|_1 = 1$, and

$$\|\mathbf{A}\mathbf{x}\|_1 = \max_j \sum_{i=1}^m |A_{i,j}|.$$
 (18)

Solution:

(c) Conclude that

$$\|\mathbf{A}\|_{1} = \max_{\|\mathbf{x}\|_{1}=1} \|\mathbf{A}\mathbf{x}\|_{1} = \max_{j} \sum_{i=1}^{m} |A_{i,j}|.$$
 (19)

Solution:

Appendix B

Exercise 6

Suppose that $\mathbf{X} = (X_1, X_2)^T$ is a vector composed of two random variables with a multivariate normal distribution with expected value μ and covariance matrix \mathbf{C} , and that \mathbf{A} is a 2 by 2 matrix. Use properties of expected value and covariance to show that $\mathbf{y} = \mathbf{A}\mathbf{x}$ has expected value $\mathbf{A}\mu$ and covariance \mathbf{ACA}^T

Solution:

Exercise 9

Using MATLAB, repeat the following experiment 1000 times. Generate five exponentially distributed random numbers from the exponential probability density function (B.10) with means $\mu = 1/\lambda = 10$. You may find the library function **exprand** to be useful here. Use (B.74) to calculate a 95% confidence interval for the 1000 mean determinations. How many times out of the 1000 experiments did the 95% confidence interval cover the expected value of 10? What happens if you instead generate 50 exponentially distributed random numbers at time? Discuss your results.

Solution:

Appendix C

Exercise 1

Let

$$f(\mathbf{x}) = x_1^2 x_2^2 - 2x_1 x_2^2 + x_2^2 - 3x_1^2 x_2 + 12x_1 x_2 - 12x_2 + 6.$$
 (20)

Find the gradient, $\nabla f(\mathbf{x})$, and Hessian, $\mathbf{H}(f(\mathbf{x}))$. What are the critical points of f? Which of these are minima and maxima of f?

Solution:

Exercise 2

Find a Taylor's series approximation for $f(\mathbf{x} + \Delta \mathbf{x})$, where

$$f(\mathbf{x}) = e^{-(x_1 + x_2)^2} \tag{21}$$

is near the point

$$\mathbf{x} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}. \tag{22}$$

Solution: