Math 341: Final

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§1

For $c \in \mathbb{R}$, we define the matrix $\mathbf{A}_c \in \mathbb{R}^{3 \times 3}$ by

$$\mathbf{A}_{c} = \begin{bmatrix} 1 & -1 & 1 \\ 2 & 2 & 0 \\ 3 & c & 2 \end{bmatrix}$$

a. Compute $det(\mathbf{A}_c)$. Does it depend of c?

Proof.

$$det(\mathbf{A}_c) = 1 \cdot det \begin{bmatrix} 2 & 0 \\ c & 2 \end{bmatrix} - (-1) det \begin{bmatrix} 2 & 0 \\ 3 & 2 \end{bmatrix} + 1 \cdot det \begin{bmatrix} 2 & 2 \\ 3 & c \end{bmatrix}$$
$$= (4) + (4) + (2c - 6)$$
$$= 2c + 2$$

Yes, it depends on c.

b. For which c is the matrix \mathbf{A}_c invertible?

Proof. Corollary to Theorem 4.7 states that a matrix is invertible if and only its determinant does not equal 0. From (a),

$$\det(\mathbf{A}_{c}) = 2c + 2 = 0$$

$$c = -1$$

 $det(\mathbf{A}_c) = 0$ when c = -1. Hence, \mathbf{A}_c is invertible for all c except c = -1.

c. Compute \mathbf{A}_0^{-1} (i.e. when c = 0).

Proof. Using the proof to Theorem 3.2, we compute the inverse by constructing an agumented matrix $(\mathbf{A}_0|I_3)$ and applying elementary row operations to transform it into the form of

 $(I_3|\mathbf{A}_0^{-1}).$

$$(\mathbf{A}_0|I_3) = \begin{bmatrix} 1 & -1 & 1 & 1 & 0 & 0 \\ 2 & 2 & 0 & 0 & 1 & 0 \\ 3 & 0 & 2 & 0 & 0 & 1 \end{bmatrix}$$

$$R_2 \leftarrow R_2 - 2R_1 \quad R_3 \leftarrow R_3 - 3R_1 = \begin{bmatrix} 1 & -1 & 1 & 1 & 0 & 0 \\ 0 & 4 & -2 & -2 & 1 & 0 \\ 0 & 3 & -1 & -3 & 0 & 1 \end{bmatrix}$$

$$R_1 \leftarrow R_1 + \frac{1}{4}R_2 \quad R_3 \leftarrow R_3 - \frac{3}{4}R_2 = \begin{bmatrix} 1 & 0 & \frac{1}{2} & \frac{1}{2} & \frac{1}{4} & 0 \\ 0 & 4 & -2 & -2 & 1 & 0 \\ 0 & 0 & \frac{1}{2} & -\frac{3}{2} & -\frac{3}{4} & 1 \end{bmatrix}$$

$$R_1 \leftarrow R_1 - R_3 \quad R_2 \leftarrow R_2 + 4R_3 = \begin{bmatrix} 1 & 0 & 0 & 2 & 1 & -1 \\ 0 & 4 & 0 & -8 & -2 & 4 \\ 0 & 0 & \frac{1}{2} & -\frac{3}{2} & -\frac{3}{4} & 1 \end{bmatrix}$$

$$R_2 \leftarrow \frac{1}{4}R_2 \quad R_3 \leftarrow 2R_3 = \begin{bmatrix} 1 & 0 & 0 & 2 & 1 & -1 \\ 0 & 1 & 0 & -2 & -\frac{1}{2} & 1 \\ 0 & 0 & 1 & -3 & -\frac{3}{2} & 2 \end{bmatrix}$$

Therefore,

$$\mathbf{A}_0^{-1} = \begin{bmatrix} 2 & 1 & -1 \\ -2 & -\frac{1}{2} & 1 \\ -3 & -\frac{3}{2} & 2 \end{bmatrix}$$

d. Let $b = (1, -4, 2)^t$, find the solution of $\mathbf{A}_0 x = b$

Proof. Let $x = (x_1, x_2, x_3)^t$.

$$\mathbf{A}_{0}x = b$$

$$\begin{bmatrix} 1 & -1 & 1 \\ 2 & 2 & 0 \\ 3 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_{1} \\ x_{2} \\ x_{3} \end{bmatrix} = \begin{bmatrix} 1 \\ -4 \\ 2 \end{bmatrix}$$

$$\begin{cases} x_{1} - x_{2} + x_{3} = 1 \\ 2x_{1} + 2x_{2} = -4 \\ 3x_{1} + 2x_{3} = 2 \end{cases}$$

$$x_{1} = -4$$

$$x_{2} = 2$$

 $x_3 = 7$

e. Compute $det(\mathbf{A}_c^2)$

Proof. From Theorem 4.7, we know that

$$det(\mathbf{A}_c^2) = det(\mathbf{A}_c\mathbf{A}_c)$$

$$= det(\mathbf{A}_c) \cdot det(\mathbf{A}_c)$$

$$= (2c+2)(2c+2)$$

$$= 4c^2 + 8c + 4$$

f. Compute $det(5\mathbf{A}_c)$

Proof. Using the second "Properties of the Determinant" on pg 234,

$$det(5\mathbf{A}_{c}) = 5^{3} det(\mathbf{A}_{c})$$
$$= 5^{3}(2c + 2)$$
$$= 250c + 250$$

g. Compute $det(\mathbf{E}_k \mathbf{A}_c)$ where,

$$\mathbf{E}_{k} = \begin{bmatrix} 1 & 0 & 0 \\ k & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Proof.

$$\det(\mathbf{E}_{k}) = 1 \cdot \det \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - 0 \cdot \det \begin{bmatrix} k & 0 \\ 0 & 1 \end{bmatrix} + 0 \cdot \det \begin{bmatrix} k & 1 \\ 0 & 0 \end{bmatrix}$$
$$= 1$$

From Theorem 4.7, we know that,

$$det(\mathbf{E}_{k}\mathbf{A}_{c}) = det(\mathbf{E}_{k}) det(\mathbf{A}_{c})$$
$$= (1)(2c + 2)$$
$$= 2c + 2$$

h. Compute $det(\mathbf{D}_k \mathbf{A}_c)$ where,

$$\mathbf{D}_{k} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & k & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Proof.

$$\det(\mathbf{D}_{k}) = 1 \cdot \det \begin{bmatrix} k & 0 \\ 0 & 1 \end{bmatrix} - 0 \cdot \det \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + 0 \cdot \det \begin{bmatrix} 0 & k \\ 0 & 0 \end{bmatrix}$$
$$= k$$

From Theorem 4.7, we know that,

$$det(\mathbf{E}_{k}\mathbf{A}_{c}) = det(\mathbf{D}_{k}) det(\mathbf{A}_{c})$$
$$= (k)(2c + 2)$$
$$= 2kc + 2k$$

i. Compute $\det(\mathbf{A}_0^{-1})$

Proof. From (c) we know that,

$$\mathbf{A}_0^{-1} = \begin{bmatrix} 2 & 1 & -1 \\ -2 & -\frac{1}{2} & 1 \\ -3 & -\frac{3}{2} & 2 \end{bmatrix}$$

It follows that,

$$\begin{split} \det(\mathbf{A}_0^{-1}) &= 2 \cdot \det \begin{bmatrix} -\frac{1}{2} & 1 \\ -\frac{3}{2} & 2 \end{bmatrix} - 1 \cdot \det \begin{bmatrix} -2 & 1 \\ -3 & 2 \end{bmatrix} - 1 \cdot \det \begin{bmatrix} -2 & -\frac{1}{2} \\ -3 & -\frac{3}{2} \end{bmatrix} \\ &= 2(\frac{1}{2}) - 1(-1) - 1(\frac{3}{2}) \\ &= \frac{1}{2} \end{split}$$

j. Compute the eigenvalues of \mathbf{A}_0 . Can you diagonalize \mathbf{A}_0 ?

Proof. Using Theorem 5.2, we compute the eigenvalues of \mathbf{A}_0 by computing its characteristic polynomial.

$$\det(\mathbf{A}_0 - \lambda I) = \det \begin{bmatrix} 1 - \lambda & -1 & 1 \\ 2 & 2 - \lambda & 0 \\ 3 & 0 & 2 - \lambda \end{bmatrix}$$
$$= -(\lambda - 2)(\lambda^2 - 3\lambda + 1)$$

We compute when $-(\lambda-2)(\lambda^2-3\lambda+1)=0$. When $\lambda=2$, the characteristic polynomial equals 0. Using the quadratic formula, when $\lambda=\frac{3\pm\sqrt{5}}{2}$, the characteristic polynomial equals 0. Hence the eigenvalues of ${\bf A}_0$ are

$$2, \frac{3+\sqrt{5}}{2}, \frac{3-\sqrt{5}}{2}$$

We can rewrite the characteristic polynomial as

$$f(\lambda) = -(\lambda - 2)(\lambda - \frac{3 + \sqrt{5}}{2})(\lambda - \frac{3 - \sqrt{5}}{2})$$

Using the "Test for Diagonalization" outlined in pg 269, we determine if \mathbf{A}_0 can be diagonalized. It is clear that the first condition, the characteristic polynomial of T splits, holds. By Theorem 5.7, condition 2 is automatically satisfied for eigenvalues of multiplicity 1. Therefore, A is diagonalizable.

k. Compute the eigenvalues of \mathbf{A}_0^{-1}

Proof. Notice that since eigenvalues are non zero,

$$\mathbf{A}_0 x = \lambda x$$

$$\mathbf{A}_0^{-1} \mathbf{A}_0 x = \mathbf{A}_0^{-1} \lambda x$$

$$x = \mathbf{A}_0^{-1} \lambda x$$

$$\frac{1}{\lambda} x = \mathbf{A}_0^{-1} x$$

Thus the eigenvalues of \mathbf{A}_0^{-1} are

$$\frac{1}{2}$$
, $\frac{3+\sqrt{5}}{2}$, $\frac{3-\sqrt{5}}{2}$

§2

Consider the transformation $T: \mathbb{R}^3 \to \mathbb{R}^4$ given by

$$T(\mathbf{x}) = \begin{pmatrix} x_2 - x_1 + \alpha \left(x_2^2 + x_3^3\right) \\ x_1 + x_2 + 2gx_3 + \alpha x_1^2 \\ x_1 + x_2 + 2x_3 \\ hx_3 + q \end{pmatrix}, \text{ where } \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

in which, h, g, q and α are real numbers.

a. What are the condition on α and q such that the transformation is linear? Explain briefly.

Proof. We know that if T is linear then T(0)=0. Hence q=0 for T to be linear. Notice that we must not have any terms higher than degree 1 because for example if we have a transformation $U(x)=x^2$ and let $c\in\mathbb{R}$ then,

$$U(cx) = (cx)^{2}$$
$$= c^{2}x^{2}$$
$$\neq cU(x)$$

which means U is not a linear transformation. Thus, $\alpha=0$ so that there are no terms higher than degree 1.

b. From now we suppose that q=0 and $\alpha=0$. Write the representation matrix $A=[T]^{\epsilon_4}_{\epsilon_3}$.

Proof. We now define

$$T(\mathbf{x}) = \begin{pmatrix} x_2 - x_1 \\ x_1 + x_2 + 2gx_3 \\ x_1 + x_2 + 2x_3 \\ hx_3 \end{pmatrix}, \text{ where } \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

We now compute the representation matrix

$$T(e_1) = \begin{pmatrix} -1 \\ 1 \\ 1 \\ 0 \end{pmatrix} \quad T(e_2) = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 0 \end{pmatrix} \quad T(e_3) = \begin{pmatrix} 0 \\ 2g \\ 2 \\ h \end{pmatrix}$$

$$A = [[T(e_1)]_{\epsilon_4}[T(e_2)]_{\epsilon_4}[T(e_3)]_{\epsilon_4}]$$

$$= \begin{bmatrix} -1 & 1 & 0 \\ 1 & 1 & 2g \\ 1 & 1 & 2 \\ 0 & 0 & h \end{bmatrix}$$

c. What are the conditions on h and g such that the transformation T maps \mathbb{R}^3 onto \mathbb{R}^4 ? Explain briefly.

Proof. Since $T: \mathbb{R}^3 \to \mathbb{R}^4$, there are no h and g such that T is onto. This is because T is onto if and only if the range of T equals the codomain, but in this linear transformation, the dimension of codomain is greater than the domain so by rank nullity, T cannot be onto. \square

d. What are the conditions on h and g such that the transformation T is one to one? Explain briefly.

Proof. By Theorem 2.4, T is one to one if and only if $N(T) = \{0\}$. Using the rank nullity theorem, this is equivalent to saying that rank(T) = 3, i.e.

$$dim(\mathbb{R}^3) = rank(T) + nullity(T)$$

 $3 = rank(T)$

We know that $rank(T) = rank([T]_{\epsilon_3}^{\epsilon_4})$ by Theorem 3.3. Additionally by Theorem 3.5., the rank of a matrix is the number of linearly independent columns. It is clear that A will have rank 3 if $h \neq 0$. There are no conditions for q.

e. Suppose that h = 0, then

$$A = \begin{bmatrix} -1 & 1 & 0 \\ 1 & 1 & 2g \\ 1 & 1 & 2 \\ 0 & 0 & 0 \end{bmatrix}$$
 and let $b = \begin{bmatrix} 0 \\ 2 \\ r \\ 0 \end{bmatrix}$

What are the conditions on r and g such that the system Ax = b has a solution? When is the solution unique?

Proof. We begin by putting the augmented matrix (A|b) in its reduced form.

$$(A'|b') = \begin{bmatrix} -1 & 1 & 0 & 0 \\ 0 & 2 & 2g & 2 \\ 0 & 0 & 2 - 2g & r - 2 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

By Theorem 3.11 and 3.13, a system is consistent if and only if rank(A') = rank(A'|b'). So for the system to have as solution, $2 - 2g \neq 0$ if $r - 2 \neq 0$. If r - 2 = 0, there is no conditions on g.

In summary, for the system to have as solution if $r \neq 2$ then $g \neq 1$. If r = 2, there are no conditions on g.

Combining Theorem 3.10 and the corollary to Theorem 4.7, we get that a system has a unique solution if and only if $det(A) \neq 0$. The determinant of an upper triangular matrix is the product of its diagonal entries by property 4 of determinants in section 4.4. Using this

fact we can compute the condition of g as such that

$$-1 * 2 * (2 - 2g) \neq 0$$
$$4g - 4 \neq 0$$
$$g \neq 1$$

There is no conditions for r when the solution is unique.

f. Suppose that h=0, g=1, r=2. Solve $\mathbf{A}\mathbf{x}=\mathbf{b}$ and give the answer in parametric form.

Proof. We define

$$\mathbf{A} = \begin{bmatrix} -1 & 1 & 0 \\ 1 & 1 & 2 \\ 1 & 1 & 2 \\ 0 & 0 & 0 \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} 0 \\ 2 \\ 2 \\ 0 \end{bmatrix}$$

We can compute a solution space to $\mathbf{A}\mathbf{x} = \mathbf{b}$ as outlined in Theorem 3.9. We start by first computing the solution set to $\mathbf{A}\mathbf{x} = 0$ denoted by K_H . It is clear that $rank(\mathbf{A}) = 2$ because the first two columns are linearly independent and the third column is the sum of the first two columns. By Theorem 3.8, $dim(K_H) = 3 - 2 = 1$. Thus any nonzero solution constitutes a basis for K. For example, since

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

is a solution to the $\mathbf{A}\mathbf{x}=0$, it is a basis for K_H by Corollary 2 of Theorem 1.10. So a solution set to K_H would be

$$\mathcal{K}_{\mathcal{H}} = \left\{ t \begin{pmatrix} 1 \\ 1 \\ -1 \\ 0 \end{pmatrix} : t \in \mathbb{R} \right\}$$

A solution to $\mathbf{A}\mathbf{x} = \mathbf{b}$ is

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

Therefore, by Theorem 3.9 we compute the solution space as

$$K = \left\{ \begin{pmatrix} 1 \\ 1 \\ 0 \\ 0 \end{pmatrix} + t \begin{pmatrix} 1 \\ 1 \\ -1 \\ 0 \end{pmatrix} : t \in \mathbb{R} \right\}$$