Name:	
Student Number:	

## SOLUTIONS TO MIDTERM #1 – LINEAR ALGEBRA AND MATRIX THEORY

	Multiple Choice	Problem 1	Problem 2	Problem 3	Problem 4	Problem 5	Total
	(10 max)	(25  max)	(15  max)	(20  max)	(15  max)	(15 max)	(100  max)
Score:							

1. Multiple Choice (10 points)

Let A, B be arbitrary  $n \times n$  matrices, and let  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$  be arbitrary vectors, and let  $c \in \mathbb{R}$  be a nonzero scalar. Which of the following are true in general? (2 points each.)

- (i) AB = BA
  - a) TRUE

- b) **FALSE**
- (ii) Only square matrices can be invertible.
  - a) TRUE

b) FALSE

(iii)  $\mathbf{x} \cdot \mathbf{y} = \mathbf{y} \cdot \mathbf{x}$ a) TRUE

b) FALSE

- (iv)  $(cA)^{-1} = cA^{-1}$ 
  - a) TRUE

- b) **FALSE**
- (v) If  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^2$  are nonzero vectors, and the angle between them is  $\theta$ , then  $\mathbf{x} \cdot \mathbf{y} = \sin(\theta) \|\mathbf{x}\| \|\mathbf{y}\|$ .
  - a) TRUE

- b) **FALSE**
- 2. Free Response

## Problem 1. (25 points).

(i) (10 points.) Let  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$  be any nonzero vectors. Using basic properties of the dot product (e.g. distributivity, non-negativity of the norm), prove the identity

$$-\|\mathbf{x}\| \|\mathbf{y}\| \le \mathbf{x} \cdot \mathbf{y} \le \|\mathbf{x}\| \|\mathbf{y}\|$$
 (Cauchy-Schwartz inequality)

(Hint: prove the identity first for the unit vectors  $\mathbf{u} = \frac{\mathbf{x}}{\|\mathbf{x}\|}$  and  $\mathbf{v} = \frac{\mathbf{y}}{\|\mathbf{y}\|}$  by distributivity for  $\|\mathbf{u} \pm \mathbf{v}\|^2$ , and factor out the scalar  $\|\mathbf{x}\| \|\mathbf{y}\|$  to prove the general identity.)

Answer to 1(i). We prove the statement first for the unit vectors **u** and **v**. We compute that

$$0 \le \|\mathbf{u} + \mathbf{v}\|^2 = \|\mathbf{u}\|^2 + 2\mathbf{u} \cdot \mathbf{v} + \|\mathbf{v}\|^2 = 2 + 2\mathbf{u} \cdot \mathbf{v}$$

by distributivity, and therefore by dividing both sides by 2, we get  $\mathbf{u} \cdot \mathbf{v} + 1 \ge 0$  and therefore  $\mathbf{u} \cdot \mathbf{v} \ge -1$ . Similarly, we compute that

$$0 \le \|\mathbf{u} - \mathbf{v}\|^2 = \|\mathbf{u}\|^2 - 2\mathbf{u} \cdot \mathbf{v} + \|\mathbf{v}\|^2 = 2 - 2\mathbf{u} \cdot \mathbf{v}$$

by distributivity, and again dividing both sides by 2, we get  $1 - \mathbf{u} \cdot \mathbf{v} \ge 0$  and therefore  $\mathbf{u} \cdot \mathbf{v} \le 1$ . Combining these inequalities gives us that

$$-1 \le \mathbf{u} \cdot \mathbf{v} = \frac{\mathbf{x}}{\|\mathbf{x}\|} \cdot \frac{\mathbf{y}}{\|\mathbf{y}\|} \le 1,$$

and multiplying both inequalities by  $\|\mathbf{x}\| \|\mathbf{y}\|$  gives the claim.

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(ii) (5 points.) Use the Cauchy-Schwartz inequality to prove Minkowski's inequality (also called the triangle inequality):

$$\|\mathbf{x} + \mathbf{y}\| \le \|\mathbf{x}\| + \|\mathbf{y}\|$$
 (Minkowski's inequality)

for any nonzero vectors  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ . (Hint: square both sides and use distributivity.)

Answer to 1(ii). We square the term on the left hand side and compute

$$\|\mathbf{x} + \mathbf{y}\|^2 = \|\mathbf{x}\|^2 + 2\mathbf{x} \cdot \mathbf{y} + \|\mathbf{y}\|^2 \le \|\mathbf{x}\|^2 + 2\|\mathbf{x}\|\|\mathbf{y}\| + \|\mathbf{y}\|^2 = (\|\mathbf{x}\| + \|\mathbf{y}\|)^2,$$

where we have used distributivity of the dot product and the Cauchy-Schwartz inequality to bound the second equation. Taking square roots of both sides proved Minkowski's inequality (since both sides are non-negative).

(iii) (5 points.) Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n \in \mathbb{R}^n$  be pair-wise orthogonal unit vectors. (That is, assume  $\mathbf{u}_i \cdot \mathbf{u}_j = 0$  whenever  $i \neq j$ , and  $\|\mathbf{u}_i\| = 1$  for all i.) Let  $\mathbf{y} \in \mathbb{R}^n$  be a vector expressible in the form

$$\mathbf{y} = \sum_{i=1}^{n} \alpha_i \mathbf{u}_i$$

for some coefficients  $\alpha_i \in \mathbb{R}$ . Prove the identity

$$\|\mathbf{y}\|^2 = \sum_{i=1}^n (\mathbf{y} \cdot \mathbf{u}_i)^2.$$
 (Parseval's identity)

Answer to 1(iii). We expand the left hand side as

$$\|\mathbf{y}\|^2 = \left(\sum_{i=1}^n \alpha_i \mathbf{u}_i\right) \cdot \left(\sum_{j=1}^n \alpha_j \mathbf{u}_j\right) = \sum_{i=1}^n \sum_{j=1}^n \alpha_i \alpha_j \mathbf{u}_i \cdot \mathbf{u}_j = \sum_{i=1}^n \alpha_i^2,$$

where we have used distributivity of the dot product, and where we have used the assumption that  $\mathbf{u}_i \cdot \mathbf{u}_j = 0$  when  $i \neq j$  and  $\mathbf{u}_i \cdot \mathbf{u}_i = 1$  for all i. But we can also compute the dot product

$$\mathbf{y} \cdot \mathbf{u}_i = \left(\sum_{j=1}^n \alpha_j \mathbf{u}_j\right) \cdot \mathbf{u}_i = \sum_{j=1}^n \alpha_j \mathbf{u}_j \cdot \mathbf{u}_i = \alpha_i$$

for any i, where we have again used the fact that  $\mathbf{u}_i \cdot \mathbf{u}_j = 0$  when  $i \neq j$  and  $\mathbf{u}_i \cdot \mathbf{u}_i = 1$  for all i. Plugging in this formula for  $\alpha_i$  completes the proof.

(iv) (3 points.) Define the matrix  $P = [\mathbf{u}_1 \ \mathbf{u}_2 \ \cdots \ \mathbf{u}_n]$ , where we consider  $\mathbf{u}_i$  as column vectors. Prove that  $P^{\top}P = I_n$ . Does this imply that P is invertible?

(Hint: use the definition of matrix multiplication in terms of dot products of columns and rows.)

**Answer to 1(iv).** Considering the vectors  $\mathbf{u}_i$  as column vectors (i.e.  $n \times 1$  matrices), we consider

$$P^{\top} = \begin{bmatrix} \mathbf{u}_1^{\top} \\ \vdots \\ \mathbf{u}_n^{\top} \end{bmatrix}$$
 as an  $n \times n$  matrix. But then the  $(i,j)$  entry of  $P^{\top}P$  is the dot product of the  $i$ 'th row

vector of  $P^{\top}$  and the j'th column vector of P, but this is precisely  $\mathbf{u}_i \cdot \mathbf{u}_j$ , which is 1 exactly when i = j and 0 otherwise. But these are precisely the entries of the identity matrix  $I_n$ , and therefore  $P^{\top}P = I_n$ . Since P is square and has a left inverse, it is therefore invertible.

(v) (2 points.) Give an example of nonsquare matrices A, B such that  $AB = I_n$ . This problem shows that nonsquare matrices can have left or right inverses.

(Hint: choose appropriate row/column vectors from the previous question.)

Answer to 1(v). If  $\mathbf{u} = \begin{bmatrix} u_1 \\ \vdots \\ u_n \end{bmatrix}$  is any column vector with unit norm, then  $\mathbf{u}^\top \cdot \mathbf{u} = [1]$  is the  $1 \times 1$ 

identity matrix, since  $\mathbf{u}^{\top} \cdot \mathbf{u} = [u_1^2 + u_2^2 + \dots + u_n^2] = [\|\mathbf{u}\|^2] = [1].$ 

Problem 2. (15 points). Systems of equations depending on a parameter. In this problem, we consider the system AX = B, where A and B are given by the augmented matrix

$$[A|B] = \begin{bmatrix} 1 & 0 & t & 0\\ 4 & 1 & 2 & 4\\ -2 & 4 & 8 & -2 \end{bmatrix},$$

where  $t \in \mathbb{R}$  is a parameter. In this question, you only need to give the answer, not the proof.

(i) (5 points.) For which  $t \in \mathbb{R}$  does the system AX = B have no solutions? When does it have a unique solution? When does it have infinitely many solutions? (Note that it's possible that only some of these possibilities will occur.)

**Answer to 2(i).** We reduce the augmented matrix [A|B] to row echelon form, by computing

$$\begin{bmatrix} 1 & 0 & t & | & 0 \\ 4 & 1 & 2 & | & 4 \\ -2 & 4 & 8 & | & -2 \end{bmatrix} \xrightarrow{\langle 2 \rangle \mapsto \langle 2 \rangle - 4\langle 1 \rangle} \begin{bmatrix} 1 & 0 & t & | & 0 \\ 0 & 1 & 2 - 4t & | & 4 \\ -2 & 4 & 8 & | & -2 \end{bmatrix} \xrightarrow{\langle 3 \rangle \mapsto \langle 3 \rangle + 2\langle 1 \rangle} \begin{bmatrix} 1 & 0 & t & | & 0 \\ 0 & 1 & 2 - 4t & | & 4 \\ 0 & 4 & 8 + 2t & | & -2 \end{bmatrix}$$

$$\xrightarrow{\langle 3 \rangle \mapsto \langle 3 \rangle - 4\langle 2 \rangle} \begin{bmatrix} 1 & 0 & t & | & 0 \\ 0 & 1 & 2 - 4t & | & 4 \\ 0 & 0 & 18t & | & -18 \end{bmatrix}$$

When t=0, we get the inconsistent system  $\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 2 & 4 & 4 \\ 0 & 0 & 0 & 0 & -18 \end{bmatrix}$  and we see that there are no solutions.

When  $t \neq 0$ , by the row operation  $\langle 3 \rangle \mapsto \frac{1}{18t} \langle 3 \rangle$ , we get the row echelon form  $\begin{bmatrix} 1 & 0 & t & 0 \\ 0 & 1 & 2 - 4t & 4 \\ 0 & 0 & 1 & -1/t \end{bmatrix}$  which we see has a unique solution.

(ii) (5 points.) Specify the solution set to AX = B for each  $t \in \mathbb{R}$ .

Answer to 2(ii). When t = 0, we have just shown that there are no solutions, so the solution set is empty. When  $t \neq 0$ , we solve for the unique solution  $X = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$ . We see immediately that  $x_3 = -1/t$ , and we can solve for

$$x_2 + (2 - 4t)x_3 = x_2 - \frac{2 - 4t}{t} = x_2 - \frac{2}{t} + 4 = 4$$

to find that  $x_2 = 2/t$ . Similarly, we can solve for

$$x_1 + tx_3 = x_1 - \frac{t}{t} = x_1 - 1 = 0$$

to find that  $x_1 = 1$ , so that we have the unique solution  $X = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2/t \\ -1/t \end{bmatrix}$  whenever  $t \neq 0$ .

(iii) (5 points.) Compute the rank of A for each  $t \in \mathbb{R}$ .

Answer to 2(iii). Using the matrices in row echelon form that we computed in 2(i), we see that A has two nonzero rows (or equivalently, two pivot entries) when t = 0, and has three nonzero rows (or equivalently, three pivot entries) whenever  $t \neq 0$ . Therefore A has rank 2 when t = 0, and has rank 3 whenever  $t \neq 0$ .

Hint: Reduce the augmented matrix [A|B] to row echelon form, and separate into two cases, depending on  $t \in \mathbb{R}$ . In both cases, the solution set and rank can then be computed explicitly.

Problem 3. (20 points). In this problem, we define the matrices

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 4 & 3 & 2 \\ 6 & 4 & 3 \end{bmatrix}, \qquad B = \begin{bmatrix} 3 \\ 1 \\ -2 \end{bmatrix}, \qquad C = \begin{bmatrix} 1 & -2 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

You may use the fact that A is invertible, with inverse

$$A^{-1} = \begin{bmatrix} 1 & -3 & 2 \\ 0 & 3 & -2 \\ -2 & 2 & -1 \end{bmatrix}.$$

(i) (5 points.) Specify the unique solution for AX = B.

**Answer to 3(i).** We use that A is invertible, and left-multiplying the equation by  $A^{-1}$  gives us that  $X = A^{-1}AX = A^{-1}B$ , and so we can compute that

$$X = A^{-1}B = \begin{bmatrix} 1 & -3 & 2 \\ 0 & 3 & -2 \\ -2 & 2 & -1 \end{bmatrix} \begin{bmatrix} 3 \\ 1 \\ -2 \end{bmatrix} = \begin{bmatrix} -4 \\ 7 \\ -2 \end{bmatrix}.$$

(ii) (5 points.) Specify the unique solution for  $A^{\top}X = B$ .

**Answer to 3(ii).** Since A is invertible, we know that its transpose  $A^{\top}$  is invertible with inverse  $(A^{-1})^{\top}$ , and therefore we can compute that

$$X = (A^{-1})^{\top} A^{\top} X = (A^{-1})^{\top} B = \begin{bmatrix} 1 & 0 & -2 \\ -3 & 3 & 2 \\ 2 & -2 & -1 \end{bmatrix} \begin{bmatrix} 3 \\ 1 \\ -2 \end{bmatrix} = \begin{bmatrix} 7 \\ -10 \\ 6 \end{bmatrix}$$

(iii) (3 points.) By row reduction or otherwise, compute the inverse matrix  $C^{-1}$ .

**Answer to 3(iii).** We do a row reduction on the augmented matrix  $[C|I_3]$  to compute that

$$\begin{bmatrix} 1 & -2 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{\langle 1 \rangle \mapsto \langle 1 \rangle + 2\langle 2 \rangle} \begin{bmatrix} 1 & 0 & 0 & 1 & 2 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix}$$

so we have compute the inverse matrix

$$C^{-1} = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

(iv) (7 points.) Specify the unique solution for ACX = B.

**Answer to 3(iv).** We note that the matrix AC is a product of invertible  $n \times n$  matrices, and therefore has inverse  $C^{-1}A^{-1}$ . Therefore, we can left-multiply by  $C^{-1}A^{-1}$  to find that

$$X = C^{-1}A^{-1}ACX = C^{-1}A^{-1}B = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & -3 & 2 \\ 0 & 3 & -2 \\ -2 & 2 & -1 \end{bmatrix} \begin{bmatrix} 3 \\ 1 \\ -2 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -4 \\ 7 \\ -2 \end{bmatrix} = \begin{bmatrix} 10 \\ 7 \\ -2 \end{bmatrix}.$$

Hint: while problems (i) and (ii) can be solved by row reduction, this would be time-consuming. It is much quicker to use the formula for the matrix  $A^{-1}$ . Problem (iv) is similar.

Problem 4 (15 points). Computations.

Let  $\mathbf{a} = [-1, 1, -1]$  and  $\mathbf{b} = [0, 1, 4]$  be vectors.

(i) (3 points.) Compute the dot product  $\mathbf{a} \cdot \mathbf{b}$ .

Answer to 4(i). We compute that

$$\mathbf{a} \cdot \mathbf{b} = [-1, 1, -1] \cdot [0, 1, 4] = (-1) \cdot 0 + 1 \cdot 1 + (-1) \cdot 4 = 1 - 4 = -3.$$

(ii) (3 points.) Compute the projection proj<sub>a</sub> b. (I.e., the projection of b onto a)

**Answer to 4(ii).** We recall the general formula for a projection, that

$$\operatorname{proj}_{\mathbf{a}} \mathbf{b} = \frac{\mathbf{a} \cdot \mathbf{b}}{\|\mathbf{a}\|^2} \mathbf{a},$$

and computing the norm  $\|\mathbf{a}\| = \sqrt{(-1)^2 + 1^2 + (-1)^2} = \sqrt{3}$ , we find that

$$\operatorname{proj}_{\mathbf{a}} \mathbf{b} = \frac{\mathbf{a} \cdot \mathbf{b}}{\|\mathbf{a}\|^2} \mathbf{a} = \frac{-3}{\sqrt{3}^2} [-1, 1, -1] = (-1)[-1, 1, -1] = [1, -1, 1].$$

- (iii) (9 points.) Specify the solution sets for each of the following augmented matrices. (No proof is required for this part.)
  - (a) (3 points.) Specify the solution set associated to:

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

Answer to 4(iii)(a). Since the bottom row contains the equation 0 = 1, the system is inconsistent, and therefore there are no solutions, and the solution set is empty.

(b) (3 points.) Specify the solution set associated to:

$$\begin{bmatrix} 1 & 0 & 3 & 0 \\ 0 & 1 & 2 & 7 \\ 0 & 0 & 1 & 3 \end{bmatrix}$$

Answer to 4(iii)(b). Since the matrix to the left of the augmentation bar is square with full rank, it is invertible, so we have a unique solution. This system is simple enough to compute by hand (though row reductions also work). We see immediately that  $x_3 = 3$ , and using the equation  $x_2 + 2x_3 = 7$ , we compute that  $x_2 = 7 - 2x_3 = 7 - 2 \cdot 3 = 1$ . Similarly, we can use the equation  $x_1 + 3x_3 = 0$  to compute that  $x_1 = -3x_3 = -3 \cdot 3 = -9$ , which gives us the solution set

$$\left\{ \begin{bmatrix} -9\\1\\3 \end{bmatrix} \right\},\,$$

which is a set with a single vector, corresponding to the unique solution to this problem.

(c) (3 points.) Specify the solution set associated to:

$$\begin{bmatrix} 1 & 0 & 3 & 2 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Answer to 4(iii)(c). This is a system in reduced row echelon form, and so we can read off the solution set immediately. Letting  $x_3$  be the free variable, we solve for the variables  $x_1$  and  $x_2$ , and we use the equations  $x_1 + 3x_3 = 2$  and  $x_2 + 2x_3 = 3$  to compute that  $[x_1, x_2, x_3] = [2 - 3x_3, 3 - 2x_3, x_3]$  provides a solution for any  $x_3$ , and since we have solved for the dependent variables  $x_1$  and  $x_2$ , these are the only possible solutions. Therefore, we have found the solution set to be

$$\left\{ \begin{bmatrix} 2 - 3x_3 \\ 3 - 2x_3 \\ x_3 \end{bmatrix} : x_3 \in \mathbb{R} \right\}.$$

**Problem 5.** (15 points). Let  $\mathbf{a} \in \mathbb{R}^n$  be any nonzero vector.

(i) (7 points.) Let P be the  $n \times n$  matrix such that  $PX = \operatorname{proj}_{\mathbf{a}} X$  for any vector  $X \in \mathbb{R}^n$ . Specify the (i, j)-entry of P. Using this formula, or otherwise, prove that  $P^2 = P$ , and prove that P is symmetric.

Answer to 5(i). Recall from Lecture 6 that we derived the formula

$$P_{ij} = \frac{a_i a_j}{\left\|\mathbf{a}\right\|^2}$$

for the projection matrix P associated to the vector  $\mathbf{a}$ . We can therefore compute the matrix product as

$$[P^{2}]_{ik} = \sum_{j=1}^{n} P_{ij} P_{jk} = \sum_{j=1}^{n} \frac{a_{i} a_{j} a_{j} a_{k}}{\|\mathbf{a}\|^{4}} = \frac{a_{i} a_{k}}{\|\mathbf{a}\|^{4}} \sum_{j=1}^{n} a_{j}^{2} = \frac{a_{i} a_{k}}{\|\mathbf{a}\|^{4}} \|\mathbf{a}\|^{2} = \frac{a_{i} a_{k}}{\|\mathbf{a}\|^{2}} = [P]_{ik}$$

for any indices i and k, and therefore  $P^2 = P$ . To prove symmetry of P, we see that  $P_{ij} = P_{ji}$  by the formula for the projection matrix, but by the definition of the matrix transpose, we have  $[P^{\top}]_{ij} = P_{ji} = P_{ij}$  for any indices i and j, and therefore  $P = P^{\top}$ , so P is a symmetric matrix.

(ii) (3 points.) Define the reflection matrix in the **a** direction as  $R = I_n - 2P$ , where P is given in (i). Using the answer to (i), show that  $R^2 = I_n$ .

Answer to 5(ii). We compute that

$$R^{2} = (I_{n} - 2P)^{2} = I_{n}^{2} - 4P + 4P^{2} = I_{n} - 4P + 4P = I_{n},$$

where we have used the property  $P = P^2$  proved in 5(i), and this proves the claim.

(iii) (5 points.) Let  $\mathbf{a} = [1, -1, 0]$ . Compute the associated projection matrix  $P \in \mathcal{M}_{3,3}$ , and find its reduced row echelon form. What is the rank of P? Is P invertible?

Answer to 5(iii). We use the formula for the projection matrix in part (i), and compute that

$$P = \frac{1}{\left\|\mathbf{a}\right\|^2} \begin{bmatrix} a_1a_1 & a_1a_2 & a_1a_3 \\ a_2a_1 & a_2a_2 & a_2a_3 \\ a_3a_1 & a_3a_2 & a_3a_3 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & -1 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1/2 & -1/2 & 0 \\ -1/2 & 1/2 & 0 \\ 0 & 0 & 0 \end{bmatrix},$$

but we can compute its reduced row echelon form to be

$$\begin{bmatrix} 1/2 & -1/2 & 0 \\ -1/2 & 1/2 & 0 \\ 0 & 0 & 0 \end{bmatrix} \overset{\langle 2 \rangle \mapsto \langle 1 \rangle + \langle 2 \rangle}{\sim} \begin{bmatrix} 1/2 & -1/2 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \overset{\langle 1 \rangle \mapsto 2 \langle 1 \rangle}{\sim} \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix},$$

which has a single nonzero row, and therefore P has rank 1. Since P is a  $3 \times 3$  matrix with non-maximum rank, it is singular, i.e. not invertible.