## 21-241: Matrix Algebra – Summer I, 2006 Practice Exam 2

1. Let 
$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$
,  $\mathbf{v}_2 = \begin{pmatrix} 2 \\ 1 \\ 3 \end{pmatrix}$ ,  $\mathbf{v}_3 = \begin{pmatrix} 4 \\ 2 \\ 6 \end{pmatrix}$ , and  $\mathbf{w} = \begin{pmatrix} 3 \\ 1 \\ 2 \end{pmatrix}$ .

(a) Is w in  $\{v_1, v_2, v_3\}$ ? How many vectors are in  $\{v_1, v_2, v_3\}$ ?

- (b) How many vectors are in span  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ ?
- (c) Is **w** in the subspace spanned by  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ ? Why?

SOLUTION.

- (a) No.  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is a set containing only three vectors  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ . Apparently,  $\mathbf{w}$  equals none of these three, so  $\mathbf{w} \notin \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ .
- (b) span  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is the set containing ALL possible linear combinations of  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ . Particularly, any scalar multiple of  $\mathbf{v}_1$ , say,  $2\mathbf{v}_1, 3\mathbf{v}_1, 4\mathbf{v}_1, \cdots$ , are all in the span. This implies span  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  contains infinitely many vectors.
- (c) To determine whether **w** belongs to span  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ , we are to look to write **w** as a linear combination of  $\mathbf{v}_1$ ,  $\mathbf{v}_2$ ,  $\mathbf{v}_3$ . For this purpose, we need to find three scalars  $c_1, c_2, c_3$ , such that  $\mathbf{w} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3$ . This amounts to solve the system  $A\mathbf{c} = \mathbf{w}$  for  $\mathbf{c} = (c_1, c_2, c_3)^T$ , where matrix  $A = (\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3)$ . Note that actually we only need to determine if this system allows a solution. Now apply Gaussian to reduce the augmented matrix in the echelon form:

$$\begin{pmatrix} 1 & 2 & 4 & | & 3 \\ 0 & 1 & 2 & | & 1 \\ -1 & 3 & 6 & | & 2 \end{pmatrix} \xrightarrow{R_3 + R_1} \begin{pmatrix} 1 & 2 & 4 & | & 3 \\ 0 & 1 & 2 & | & 1 \\ 0 & 5 & 10 & | & 5 \end{pmatrix} \xrightarrow{R_3 - 5R_2} \begin{pmatrix} \boxed{1} & 2 & 4 & | & 3 \\ 0 & \boxed{1} & 2 & | & 1 \\ 0 & 0 & 0 & | & 0 \end{pmatrix}$$

The bottom row doesn't lead to inconsistency, so the system allows a solution (actually has infinitely many). This shows that **w** is in the subspace spanned by  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ .

2. Given subspaces H and K of a vector space V, the sum of H and K, written as H + K, is the set of all vectors in V that can be written as the sum of two vectors, one in H and the other in K; that is,

$$H + K = {\mathbf{w} | \mathbf{w} = \mathbf{u} + \mathbf{v} \text{ for some } \mathbf{u} \in H \text{ and some } \mathbf{v} \in K}$$

- (a) Show that H + K is subspace of V.
- (b) Show that H is a subspace of H + K and K is a subspace of H + K.

## Proof.

(a) Since H and K are subspaces of V, the zero vector **0** has to belong to them both. Taking  $\mathbf{u} = \mathbf{v} = \mathbf{0}$ , we have  $\mathbf{w} = \mathbf{0} + \mathbf{0} = \mathbf{0}$ , which, by definition, belongs to H + K. Next, we are to show H + K is closed under both addition and scalar multiplication. Suppose  $\mathbf{w}_1, \mathbf{w}_2$  are two vectors in H + K. By definition, they can be written as

$$\mathbf{w}_1 = \mathbf{u}_1 + \mathbf{v}_1, \quad \mathbf{w}_2 = \mathbf{u}_2 + \mathbf{v}_2, \quad \text{for some } \mathbf{u}_1, \mathbf{u}_2 \in H \text{ and some } \mathbf{v}_1, \mathbf{v}_2 \in K.$$

Hence,

$$\mathbf{w}_1 + \mathbf{w}_2 = (\mathbf{u}_1 + \mathbf{v}_1) + (\mathbf{u}_2 + \mathbf{v}_2) = (\mathbf{u}_1 + \mathbf{u}_2) + (\mathbf{v}_1 + \mathbf{v}_2),$$

where,  $\mathbf{u}_1 + \mathbf{u}_2 \in H$  because H is a subspace, thus closed under addition; and  $\mathbf{v}_1 + \mathbf{v}_2 \in K$  similarly. This shows that  $\mathbf{w}_1 + \mathbf{w}_2$  can be written as the sum of two vectors, one in H and the other in K. So, again by definition,  $\mathbf{w}_1 + \mathbf{w}_2 \in H + K$ , namely, H + K is closed under addition. For scalar multiplication, note that given scalar c,

$$c\mathbf{w}_1 = c(\mathbf{u}_1 + \mathbf{v}_1) = c\mathbf{u}_1 + c\mathbf{v}_1$$

where  $c\mathbf{u}_1 \in H$  because H is closed under scalar multiplication; and  $c\mathbf{v}_1 \in K$  parallelly. Now that  $c\mathbf{w}_1$  has been written as the sum of two vectors, one in H and the other in K, it's in H + K. That is, H + K is closed under scalar multiplication. And we are done.

- (b) Since H is a subspace of V, it's nonempty, closed under addition and scalar multiplication. We only need to show that H is a subset of H + K. This is derived from the fact that each vector in H can be written as the sum of itself, which belongs to H, and the zero vector, which belongs to K. A similar argument justifies K is a subspace of H + K, too.
- 3. Let **x** and **y** be linearly independent elements of a vector space V. Show that  $\mathbf{u} = a\mathbf{x} + b\mathbf{y}$  and  $\mathbf{v} = c\mathbf{x} + d\mathbf{y}$  are linearly independent if and only if  $ad bc \neq 0$ . Is the entire collection **x**, **y**, **u**, **v** linearly independent?

PROOF. Let 
$$A = (\mathbf{x} \ \mathbf{y}), B = (\mathbf{u} \ \mathbf{v}), C = \begin{pmatrix} a & c \\ b & d \end{pmatrix}$$
, then  

$$AC = (\mathbf{x} \ \mathbf{y}) \begin{pmatrix} a & c \\ b & d \end{pmatrix} = (a\mathbf{x} + b\mathbf{y} \quad c\mathbf{x} + d\mathbf{y}) = (\mathbf{u} \ \mathbf{v}) = B.$$

Two key facts we'll use later are that **u** and **v** (or, **x** and **y**) are linearly independent if and only if the homogeneous system  $B\mathbf{r} = \mathbf{0}$  (or,  $A\mathbf{r} = \mathbf{0}$ ) allows only trivial solution, denoted by Fact 1 (or, Fact 2). Now slow down, carefully think of the following deduction process, and make sure you really understand each step involved.

$\mathbf{u}, \mathbf{v}$ are linearly independent	
$\iff B\mathbf{r} = 0$ has only trivial solution	(by Fact 1)
$\iff (AC)\mathbf{r} = 0$ has only trivial solution	(since $AC = B$ )
$\iff A(C\mathbf{r}) = 0$ has only trivial solution	(by associativity)
$\iff C\mathbf{r} = 0$ has only trivial solution	(by Fact 2, replace $\mathbf{r}$ by $C\mathbf{r}$ )
$\iff C$ is nonsingular	
$\Longleftrightarrow \det C \neq 0$	
$\iff ad - bc \neq 0$	

The entire collection  $\mathbf{x}$ ,  $\mathbf{y}$ ,  $\mathbf{u}$ ,  $\mathbf{v}$  is linearly dependent, since we have four scalars, a, b, -1, 0, not all zero, such that the linear combination  $a\mathbf{x} + b\mathbf{y} + (-1)\mathbf{u} + 0\mathbf{v} = \mathbf{0}$ .

4. Find bases for the column space (range) and null space (kernel) of the matrix  $A = \begin{pmatrix} -2 & 4 & -2 & -4 \\ 2 & -6 & -3 & 1 \\ -3 & 8 & 2 & -3 \end{pmatrix}$ .

SOLUTION. To find the basis for column space, we need to find pivot column(s). To find the basis for null space, we need to find general solution to the homogeneous system  $A\mathbf{x} = \mathbf{0}$ . Both can be achieved

by reducing the matrix in the echelon form.

$$\begin{pmatrix} -2 & 4 & -2 & -4 \\ 2 & -6 & -3 & 1 \\ -3 & 8 & 2 & -3 \end{pmatrix} \xrightarrow{R_2+R_1} \begin{pmatrix} -2 & 4 & -2 & -4 \\ 0 & -2 & -5 & -3 \\ 0 & 2 & 5 & 3 \end{pmatrix} \xrightarrow{R_3+R_2} \begin{pmatrix} \hline -2 & 4 & -2 & -4 \\ 0 & \hline -2 & -5 & -3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

We see that the first two columns are pivot columns, so the first two column of the ORIGINAL MATRIX A, namely,  $\{(-2, 2, -3)^T, (4, -6, 8)^T\}$ , form a basis for Col A. The last two columns are free, and we can easily read the general solution from the echelon form:

$$x_2 = -\frac{5}{2}x_3 - \frac{3}{2}x_4, \quad x_1 = 2x_2 - x_3 - 2x_4 = -6x_3 - 5x_4, \quad x_3, x_4$$
 free

Written in vector form,

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} -6x_3 - 5x_4 \\ -\frac{5}{2}x_3 - \frac{3}{2}x_4 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} -6x_3 \\ -\frac{5}{2}x_3 \\ x_3 \\ 0 \end{pmatrix} + \begin{pmatrix} -5x_4 \\ -\frac{3}{2}x_4 \\ 0 \\ x_4 \end{pmatrix} = x_3 \begin{pmatrix} -6 \\ \frac{5}{2} \\ 1 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} -5 \\ -\frac{3}{2} \\ 0 \\ 1 \end{pmatrix}$$

Thus,  $\{(-6, \frac{5}{2}, 1, 0)^T, (-5, \frac{3}{2}, 0, 1)^T\}$  form a basis for Nul A.

5. Show that  $\left\{ \mathbf{u}_1 = \begin{pmatrix} 3 \\ -5 \end{pmatrix}, \mathbf{u}_2 = \begin{pmatrix} -4 \\ 6 \end{pmatrix} \right\}$  is a basis for  $\mathbb{R}^2$ . Let  $\mathbf{x} = \begin{pmatrix} 2 \\ -6 \end{pmatrix}$ . Find the coordinate vector for  $\mathbf{x}$  with respect to this basis.

SOLUTION. First of all,  $\mathbf{u}_1$  and  $\mathbf{u}_2$  are linearly independent because they are not multiples of each other. Next, we are to characterize vectors in span  $\{\mathbf{u}_1, \mathbf{u}_2\}$ . Suppose vector  $\mathbf{b} \in \mathbb{R}^2$  belongs to span  $\{\mathbf{u}_1, \mathbf{u}_2\}$ , then the linear system  $A\mathbf{y} = \mathbf{b}$  is consistent, where matrix  $A = (\mathbf{u}_1 \quad \mathbf{u}_2)$ . Applying Gaussian to the augmented matrix, we get

$$\begin{pmatrix} 3 & -4 & b_1 \\ -5 & 6 & b_2 \end{pmatrix} \xrightarrow{R_2 + \frac{5}{3}R_1} \begin{pmatrix} 3 & -4 & b_1 \\ 0 & -\frac{2}{3} & b_2 + \frac{5}{3}b_1 \end{pmatrix}$$

The system has a pivot in each row, thus is always consistent for all possible  $\mathbf{b} \in \mathbb{R}^2$ . Therefore, span  $\{\mathbf{u}_1, \mathbf{u}_2\} = \mathbb{R}^2$ , and  $\{\mathbf{u}_1, \mathbf{u}_2\}$  form a basis for  $\mathbb{R}^2$ . To find the coordinate vector for  $\mathbf{x}$ , we need find the solution to  $A\mathbf{y} = \mathbf{x}$ . Replacing  $b_1, b_2$  by 2, -6 respectively in the echelon form we obtained above, we can write out the solution  $\mathbf{y} = (6, 4)^T$ . This is to say,  $\mathbf{x} = 6\mathbf{u}_1 + 4\mathbf{u}_2$ , so the coordinate vector for  $\mathbf{x}$  w.r.t  $\{\mathbf{u}_1, \mathbf{u}_2\}$  is  $(6, 4)^T$ .

6. Let V be an inner product space.

(a) Prove that  $\langle \mathbf{x}, \mathbf{v} \rangle = 0$  for all  $\mathbf{v} \in V$  if and only if  $\mathbf{x} = \mathbf{0}$ .

- (b) Prove that  $\langle \mathbf{x}, \mathbf{v} \rangle = \langle \mathbf{y}, \mathbf{v} \rangle$  for all  $\mathbf{v} \in V$  if and only if  $\mathbf{x} = \mathbf{y}$ .
- (c) Let  $\mathbf{v}_1, \dots, \mathbf{v}_n$  be a basis for V. Prove that  $\langle \mathbf{x}, \mathbf{v}_i \rangle = \langle \mathbf{y}, \mathbf{v}_i \rangle$ ,  $i = 1, \dots, n$ , if and only if  $\mathbf{x} = \mathbf{y}$ .

Proof.

(a) Suppose  $\langle \mathbf{x}, \mathbf{v} \rangle = 0$  for all  $\mathbf{v} \in V$ . Simply let  $\mathbf{v} = \mathbf{x}$  and we get  $\langle \mathbf{x}, \mathbf{x} \rangle = 0$ , which implies  $\mathbf{x} = 0$ .

(b) We reduce the equivalence as follows:

$$\langle \mathbf{x}, \mathbf{v} \rangle = \langle \mathbf{y}, \mathbf{v} \rangle, \forall \mathbf{v} \in V$$

$$\Leftrightarrow \langle \mathbf{x}, \mathbf{v} \rangle - \langle \mathbf{y}, \mathbf{v} \rangle = 0, \forall \mathbf{v} \in V$$

$$\Leftrightarrow \langle \mathbf{x} - \mathbf{y}, \mathbf{v} \rangle = 0, \forall \mathbf{v} \in V$$

$$\Leftrightarrow \mathbf{x} - \mathbf{y} = \mathbf{0}$$

$$(by part (a))$$

$$\Leftrightarrow \mathbf{x} = \mathbf{y}$$

(c) If  $\mathbf{x} = \mathbf{y}$ , of course we have  $\langle \mathbf{x}, \mathbf{v}_i \rangle = \langle \mathbf{y}, \mathbf{v}_i \rangle$ ,  $i = 1, 2, \dots, n$ . Reversely, suppose  $\langle \mathbf{x}, \mathbf{v}_i \rangle = \langle \mathbf{y}, \mathbf{v}_i \rangle$ ,  $i = 1, \dots, n$ . Since  $\mathbf{v}_1, \dots, \mathbf{v}_n$  is a basis for V, any vector  $\mathbf{v} \in V$  can be written as a linear combination of the *n* vectors, say,  $\mathbf{v} = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n$ . Linearity of inner product implies

$$\langle \mathbf{x}, \mathbf{v} \rangle = \langle \mathbf{x}, c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n \rangle$$
  

$$= c_1 \langle \mathbf{x}, \mathbf{v}_1 \rangle + c_2 \langle \mathbf{x}, \mathbf{v}_2 \rangle + \dots + c_n \langle \mathbf{x}, \mathbf{v}_n \rangle$$
 (by linearity)  

$$= c_1 \langle \mathbf{y}, \mathbf{v}_1 \rangle + c_2 \langle \mathbf{y}, \mathbf{v}_2 \rangle + \dots + c_n \langle \mathbf{y}, \mathbf{v}_n \rangle$$
 (since  $\langle \mathbf{x}, \mathbf{v}_i \rangle = \langle \mathbf{y}, \mathbf{v}_i \rangle$ )  

$$= \langle \mathbf{y}, c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n \rangle$$
 (by linearity)  

$$= \langle \mathbf{y}, \mathbf{v} \rangle$$

Since this equality holds for all  $\mathbf{v} \in V$ , part (b) tells us that  $\mathbf{x} = \mathbf{y}$ .

7. Prove that

$$(a_1 + a_2 + \dots + a_n)^2 \leq n(a_1^2 + a_2^2 + \dots + a_n^2)$$

for any real numbers  $a_1, \dots, a_n$ . When does equality hold? PROOF. Let  $\mathbf{u} = (a_1, a_2, \dots, a_n)^T$ ,  $\mathbf{v} = (1, 1, \dots, 1)^T$ . Then,

$$\mathbf{u} \cdot \mathbf{v} = a_1 + a_2 + \dots + a_n, \quad \|\mathbf{u}\|^2 = a_1^2 + a_2^2 + \dots + a_n^2, \quad \|\mathbf{v}\|^2 = n.$$

By Cauchy-Schwarz inequality,  $|\mathbf{u} \cdot \mathbf{v}| \leq ||\mathbf{u}|| \, ||\mathbf{v}||$ . Squaring both sides, we obtain

$$(a_1 + a_2 + \dots + a_n)^2 \leq n(a_1^2 + a_2^2 + \dots + a_n^2).$$

This completes the proof.

8. Verify the formula  $\|\mathbf{v}\| = \max\{|v_1 + v_2|, |v_1 - v_2|\}$  defines a norm on  $\mathbb{R}^2$ . Establish the equivalence between this norm and the usual Euclidean norm  $\|\cdot\|_2$ .

PROOF. We need verify positivity, homogeneity and triangle inequality one by one.

**Positivity:** Since  $|v_1 + v_2| \ge 0$ ,  $|v_1 - v_2| \ge 0$ , it's clear that  $\|\mathbf{v}\| \ge 0$ . Moreover,  $\|\mathbf{v}\| = 0$  if and only if  $|v_1 + v_2| = |v_1 - v_2| = 0$ , that is,  $v_1 = v_2 = 0$ , namely  $\mathbf{v} = \mathbf{0}$ .

Homogeneity:  $||c\mathbf{v}|| = \max\{|cv_1 + cv_2|, |cv_1 - cv_2|\} = |c| \max\{|v_1 + v_2|, |v_1 - v_2|\} = |c| ||\mathbf{v}||.$ 

Triangle Inequality: We need use triangle inequality for absolute value

$$|a+b| \leqslant |a| + |b|,$$

and the fact (denoted by Fact 3) that "the maximum of sums is less than or equal to the sum of maximums",

$$\max\{a_1 + a_2, b_1 + b_2\} \leqslant \max\{a_1, b_1\} + \max\{a_2, b_2\}$$

Both inequalities are usual. Having them, we can obtain the triangle inequality for norm  $\|\cdot\|$ :

$$\begin{aligned} \|\mathbf{u} + \mathbf{v}\| &= \max\{|(u_1 + v_1) + (u_2 + v_2)|, |(u_1 + v_1) - (u_2 + v_2)|\} & \text{(by definition)} \\ &\leqslant \max\{|u_1 + u_2| + |v_1 + v_2|, |u_1 - u_2| + |v_1 - v_2|\} & \text{(by triangle inequality)} \\ &\leqslant \max\{|u_1 + u_2|, |u_1 - u_2|\} + \max\{|v_1 + v_2|, |v_1 - v_2|\} & \text{(by Fact 3)} \\ &= \|\mathbf{u}\| + \|\mathbf{v}\|. & \text{(by definition)} \end{aligned}$$

We just directly proved  $\|\cdot\|$  defines a norm. If we think in another way, it can be verified that

$$\|\mathbf{v}\| = \max\{|v_1 + v_2|, |v_1 - v_2|\} = |v_1| + |v_2| = \|\mathbf{v}\|_1,$$

namely,  $\|\cdot\|$  is actually the 1-norm  $\|\cdot\|_1$ . The proof is not difficult and left to you. To show the equivalence of two norms, we need to find two POSITIVE constants m, M, such that

$$m \|\mathbf{v}\|_2 \leq \|\mathbf{v}\| \leq M \|\mathbf{v}\|_2$$
, for all  $\mathbf{v} \in \mathbb{R}^2$ .

You may already find it's convenient to compare squares of norms when Euclidean norm is involved. So, let's square:

$$\|\mathbf{v}\|^{2} = (\max\{|v_{1}+v_{2}|, |v_{1}-v_{2}|\})^{2} = \max\{|v_{1}+v_{2}|^{2}, |v_{1}-v_{2}|^{2}\}$$
  
=  $\max\{v_{1}^{2}+v_{2}^{2}+2v_{1}v_{2}, v_{1}^{2}+v_{2}^{2}-2v_{1}v_{2}\} = v_{1}^{2}+v_{2}^{2}+\max\{2v_{1}v_{2}, -2v_{1}v_{2}\}$   
=  $v_{1}^{2}+v_{2}^{2}+|2v_{1}v_{2}| \ge v_{1}^{2}+v_{2}^{2} = \|\mathbf{v}\|_{2}^{2}.$ 

This allows us letting m = 1. On the other hand, since  $|2v_1v_2| \leq v_1^2 + v_2^2$ ,

$$\|\mathbf{v}\|^2 = v_1^2 + v_2^2 + |2v_1v_2| \leq 2(v_1^2 + v_2^2) = 2\|\mathbf{v}\|_2^2.$$

Therefore, we can choose  $M = \sqrt{2}$ . Thus we complete the proof.

9. Prove the matrix  $A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & -2 \\ 1 & -2 & 14 \end{pmatrix}$  is positive definite. Find its Cholesky factorization.

PROOF. We apply Gaussian to show the matrix has all positive pivots and find the  $LDL^{T}$  factorization.

$$\begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & -2 \\ 1 & -2 & 14 \end{pmatrix} \xrightarrow{R_2 - R_1} \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & -3 \\ 0 & -3 & 13 \end{pmatrix} \xrightarrow{R_3 + 3R_2} \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & -3 \\ 0 & 0 & 4 \end{pmatrix}$$
$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \xrightarrow{R_2 + R_1} \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \xrightarrow{R_3 - 3R_2} \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & -3 & 1 \end{pmatrix} = L$$

Now, we see the matrix is regular and has all positive pivots 1, 1, 4, thus is positive definite. Let D = diag(1, 1, 4), S = diag(1, 1, 2), then we obtain the Cholesky factorization

$$A = LDL^T = LS^2L^T = LSS^TL^T = MM^T,$$

where  $M = LS = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & -3 & 2 \end{pmatrix}$ .