

### Partial Solutions.

Where solutions are given, please either substitute your own solution or rewrite the given solution yourself with some change of notation.

**Exercise 1.** Let  $V = \{v_1, \dots, v_m\}$  be some arbitrary subset of vectors taken from a vector space  $\mathcal{W}$ .

- (a) Show explicitly that a scalar multiple of a linear combination of the vectors in  $V$  is also a linear combination of vectors in  $V$ .

**Solution.** By linearity of scalar-vector multiplication, for every scalar  $\beta$ ,

$$\beta \cdot \sum_{i=1}^m \alpha_i v_i = \sum_{i=1}^m \beta \alpha_i v_i = \sum_{i=1}^m (\beta \cdot \alpha_i) v_i,$$

a linear combination with coefficients  $\beta \alpha_i$ .

- (b) Show explicitly that a linear combination of linear combinations of vectors in  $V$  also a linear combination of vectors in  $V$ .

**Solution.**

$$\sum_j \beta_j \left( \sum_{i=1}^m \alpha_i v_i \right) = \sum_j \sum_{i=1}^m \beta_j \alpha_i v_i = \sum_{i=1}^m \alpha_i \left( \sum_j \beta_j \right) v_i,$$

which is a linear combination of the  $v_i$ , with  $i$ th coefficient  $\alpha_i \sum_j \beta_j$ .

- (c) Use (a) and (b) above to show that  $\text{span}(V)$  is a vector subspace of  $\mathcal{W}$ .

**Solution.** As always, it suffices to show that  $\text{span}(V)$  is closed under vector addition (shown in Part (b) above) and scalar multiplication (shown in Part (a) above).

- (d) Now suppose  $\mathcal{W} = \mathbf{R}^n$ ; hence,  $v_i \in \mathbf{R}^n$  for each  $v_i \in V$ . Define a matrix  $A$  with the property that  $\text{range}(A) = \text{span}(V)$ . How are the dimensions and rank of  $A$  related to  $V$ ,  $\text{span}(V)$ , and  $\mathcal{W}$ ?

### Exercise 2.

- (a) What is the *inverse* of a matrix?  
 (b) What shape of matrix can have an inverse?  
 (c) Prove that the inverse of a matrix is unique, if it exists.

**Solution.** Suppose  $B$  and  $C$  are both inverses of  $A$ . Then  $I = AB = AC$ . Subtracting  $AC$  from both sides of the last equation gives  $A(B - C) = 0$ , since matrix-matrix multiplication is linear. Because  $A$  has independent columns, every column of the matrix  $B - C$  must be zero. Hence,  $B = C$ .

- (d) Define the *transpose*  $A^T$  of a matrix  $A \in \mathbf{R}^{m \times n}$ .

**Solution.**  $A_{ij}^T \equiv A_{ji}$  for all  $i \in \{1, \dots, n\}$ ,  $j \in \{1, \dots, m\}$ .

- (e) Prove that if the matrix-matrix product  $AB$  is conformable (i.e., the number of columns of  $A$  equals the number of rows of  $B$ ), then  $(AB)^T = B^T A^T$ . Verify this statement in Matlab for  $m \times p$  matrix  $A$  and  $p \times n$  matrix  $B$ , for a few choices of  $A$  and  $B$ , both square ( $m = p = n$ ) and non-square.

**Solution.** The  $ij$ th component of  $(AB)^T$  equals the  $j$ th component of  $AB$  equals the inner product of the  $j$ th row of  $A$  and the  $i$ th column of  $B$ . This is the same as the  $ij$ th component of  $B^T A^T$ .

- (f) Prove that if  $A$  is invertible, then so is  $A^T$ , and  $(A^{-1})^T = (A^T)^{-1}$ . Explain why this fact justifies the use of the notation  $A^{-T}$ . Verify this statement in Matlab for a few nontrivial choices of  $A$ .

**Solution Hint.** Use the uniqueness of inverses already shown above.

**Exercise 3.**

- (a) Show explicitly that matrix-vector multiplication is a linear operation. For scalar  $\alpha$ , matrix  $A \in \mathbf{R}^{m \times n}$ , and vectors  $x, y \in \mathbf{R}^n$ ,  $A(\alpha x) = \alpha Ax$  and  $A(x + y) = Ax + Ay$ .

**Solution.** Use the definition of matrix multiplication: if  $y = Ax$ , then  $y_i = \sum_{j=1}^n a_{ij}x_j$  for each  $i = 1, \dots, m$ .

$$(A(\alpha x))_i = \sum_{j=1}^n \alpha a_{ij}x_j = \alpha \sum_{j=1}^n a_{ij}x_j = \alpha (Ax)_i$$

Hence,  $A(\alpha x) = \alpha Ax$ .

$$(A(x + y))_i = \sum_{j=1}^n a_{ij}(x_j + y_j) = \sum_{j=1}^n a_{ij}x_j + \sum_{j=1}^n a_{ij}y_j = (Ax + Ay)_i.$$

Hence,  $A(x + y) = Ax + Ay$ .

- (b) Show that if  $F : \mathbf{R}^n \rightarrow \mathbf{R}^m$  is a linear function and  $G : \mathbf{R}^m \rightarrow \mathbf{R}^p$  is a linear function, then the composition function  $H : \mathbf{R}^n \rightarrow \mathbf{R}^p$  defined by  $H(x) = G(F(x))$  is also a linear function. **Remark.** This result says that any sequence of linear transformations (2 or more) may be viewed as a single linear transformation.
- (c) If, in the preceding question,  $m = n = p$  and  $F$  and  $G$  are invertible, then show that  $H$  is invertible, and  $H^{-1}(y) = F^{-1}(G^{-1}(y))$ . Hint: to show that functions  $S$  and  $T$  are inverses of each other, it suffices to show that (i)  $T(S(x)) = x$  for all  $x$  in the domain of  $S$ , and (ii)  $S(T(y)) = y$  for all  $y$  in the domain of  $T$ .

Verify the following statement in Matlab for a few nontrivial choices of  $A$  and  $B$ .

**Corollary.** If  $A$  and  $B$  are invertible  $n \times n$  matrices, then  $(AB)^{-1} = B^{-1}A^{-1}$ .

**Exercise 4.** Prove the following and illustrate in Matlab for a few nontrivial choices of the given matrices.

- (a) If  $D$  and  $E$  are diagonal, then  $DE = ED$ .  
Because  $D$  is diagonal,  $d_{ij} = 0$  for  $i \neq j$ . Show  $DE$  and  $ED$  are also diagonal.
- (b) Premultiplication of  $A \in \mathbf{R}^{m \times n}$  by a diagonal matrix  $D \in \mathbf{R}^m$  scales the  $i$ th row of  $A$  by  $d_{ii}$ .  
Use the row-oriented view of matrix multiplication from the lecture notes (or don't).
- (c) Postmultiplication of  $A \in \mathbf{R}^{m \times n}$  by a diagonal matrix  $D \in \mathbf{R}^n$  scales the  $j$ th column of  $A$  by  $d_{jj}$ .  
Use the column-oriented view of matrix multiplication from the lecture notes (or don't).

**Exercise 5.** Let  $V = \{v_1, v_2, \dots, v_m\}$  be a finite set of vectors taken from a vector space  $\mathcal{W}$ . Let  $\bar{V}$  be the same as  $V$ , except that, in  $\bar{V}$ , some element  $v_i \in V$  is replaced by  $v_i + \kappa_j v_j$ , with  $i \neq j$  and  $\kappa_j \neq 0$ . I.e.,

$$\bar{V} = \{v_1, \dots, v_{i-1}, v_i + \kappa_j v_j, v_{i+1}, \dots, v_m\}.$$

Show that  $\text{span}(V) = \text{span}(\bar{V})$ .

**Exercise 6. Basic Theory.**

Hints for this exercise are posted separately.

Now let  $\{v_1, v_2, \dots, v_m\}$  be the rows of the matrix  $A$ . (Note that  $A$  has  $m$  rows.) Gaussian elimination with row interchanges (GEPP — “pp” stands for “partial pivoting”) can be used to reduce  $A$  to a matrix  $E$  in row-echelon form, defined as follows.

- (i) All zero-rows of  $E$  are at the bottom. I.e., if the  $i$ th row of  $E$  is zero for  $i < m$ , then each row  $j$  below it, i.e., with  $j > i$ , is also zero.

- (ii) If  $e_{ij}$  is the first nonzero element in the  $i$ th row of  $E$ , then all elements below it and to its left (i.e. all  $e_{pq}$  such that  $p > i$  and  $q \leq j$ ) are zero.

The first nonzero entry in each nonzero row of  $E$  is called a *pivot*. A column or row containing a pivot is called a *pivot column* or row, respectively. Here is an example of a matrix in row echelon form. The symbol  $\times$  stands for any value, zero or nonzero. The symbol  $*$  stands for a nonzero pivot value.

$$E = \begin{pmatrix} * & \times & \times & \times & \times & \times & \times & \times & \times & \times \\ 0 & 0 & * & \times & \times & \times & \times & \times & \times & \times \\ 0 & 0 & 0 & * & \times & \times & \times & \times & \times & \times \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & * & \times & \times \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}. \tag{1}$$

To obtain *reduced row-echelon form* from row-echelon form, (i) use row operations to eliminate all elements in each pivot column except the pivot itself, then (ii) scale the rows (a linear operation — use a diagonal matrix) to make the pivots all equal one. After reducing  $E$  above to reduced row-echelon form, we obtain the following matrix  $\bar{E}$ .

$$\bar{E} = \begin{pmatrix} 1 & \times & 0 & 0 & \times & \times & \times & 0 & \times & 0 \\ 0 & 0 & 1 & 0 & \times & \times & \times & 0 & \times & 0 \\ 0 & 0 & 0 & 1 & \times & \times & \times & 0 & \times & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & \times & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}. \tag{2}$$

- (a) (i) Prove that the nonzero rows of a matrix in reduced row-echelon form are linearly independent.  
 (ii) Prove that the pivot columns of a matrix in reduced row-echelon form are linearly independent.
- (b) Prove that interchanging 2 rows of any  $m \times n$  matrix  $A$  is an invertible linear operation. Give the form of the matrix  $P$  that exchanges rows  $i$  and  $j$  of  $A$  in the product  $PA$  (hint: interchange rows  $i$  and  $j$  of the identity matrix).
- (c) Each step of the GEPP reduction is either an interchange of two rows or the replacement of some row  $v_i$  by the vector  $v_i + \kappa_j v_j$  for some other row  $v_j$  and some nonzero scalar  $\kappa_j$ , as in the previous exercise. Show that each such row replacement is an invertible linear operation.
- (d) By a previous exercise, the entire sequence of row replacements and row interchanges can be viewed as a single, invertible, linear transformation,  $R$ , applied to  $A$  on its left. Explain why the row space of  $E = RA$  is the same as the row space of  $A$ .
- (e) Prove that the nonzero rows of the (either reduced or unreduced) row-echelon form of any matrix  $A$  are a *basis* for the row space of  $A$ .
- (f) Prove that the pivot positions in the row-echelon form of  $A$  are uniquely determined. I.e., no matter what row ordering of  $A$  you use, the pivots in its row-echelon form must always take the same positions (row and column indices). (Hint: show that changing pivot positions changes the row space.)
- (g) Explain how the above results show that
  - (i) The (maximum possible) number of linearly independent rows of a matrix (i.e., *row rank*) equals the (maximum possible) number of linearly independent columns (i.e., the *column rank*) of the matrix. Both these numbers are equal to the number of vectors in a basis for the range of the matrix (i.e., the *rank* of the matrix).
  - (ii) Any 2 bases for the same subspace  $S$  have the same number of elements (hint: let  $S$  be the row space of a matrix).
  - (iii) The (unmodified) rows in  $A$  corresponding to nonzero rows in the row-echelon form of  $A$  are also a basis for the row space of  $A$ .

- (h) Show by example that, in general,  $\text{range}(E) \neq \text{range}(A)$ .
- (i) Show that the columns of  $A$  corresponding to pivot columns of  $E$  define a basis for  $\text{range}(A)$ .

**Exercise 7.** Given fixed  $A \in \mathbf{R}^{m \times n}$ ,  $b \in \mathbf{R}^m$ , what is the significance each of the following facts in relation to the linear system of equations  $Ax = b$ ?

- (a)  $A$  has full row rank; i.e.,  $\text{rank}(A) = m$ .
- (b)  $A$  has full column rank; i.e.,  $\text{rank}(A) = n$ .

**Solution Hints.**

- (a) System  $Ax = b$  has a solution if and only if  $b \in \text{range}(A)$ . It was shown in lecture that  $\text{range}(A)$  is a subspace of  $\mathbf{R}^m$ . By definition,  $\text{rank}(A) = \dim(\text{range}(A))$ . If  $b \in \mathbf{R}^m$  and  $b \notin \text{range}(A)$ , then  $\dim(\mathbf{R}^m) > \dim(\text{range}(A))$ .
- (b) If  $\text{rank}(A) = n$ , then the columns of  $A$  are linearly independent.

**Exercise 8.** Explain the more efficient way to calculate the given matrix-vector products.

- (a)  $y = (AB)x$  or  $y = A(Bx)$ .
- (b)  $A = (I + uv^T)$  followed by  $y = Ax$  or  $y = x + (v^T x)u$ .

**Solution Hint.** Just count the total number of multiplications and additions in each case.

**Exercise 9.** Let

$$M_1 = \begin{pmatrix} A_1 & B_1 \\ C_1 & D_1 \end{pmatrix} \quad \text{and} \quad M_2 = \begin{pmatrix} A_2 & B_2 \\ C_2 & D_2 \end{pmatrix},$$

where  $A_1 \in \mathbf{R}^{m_{11} \times n_{11}}$ ,  $B_1 \in \mathbf{R}^{m_{11} \times n_{12}}$ ,  $C_1 \in \mathbf{R}^{m_{12} \times n_{11}}$ ,  $D_1 \in \mathbf{R}^{m_{12} \times n_{12}}$ ; and  $A_2 \in \mathbf{R}^{m_{21} \times n_{21}}$ ,  $B_2 \in \mathbf{R}^{m_{21} \times n_{22}}$ ,  $C_2 \in \mathbf{R}^{m_{22} \times n_{21}}$ ,  $D_2 \in \mathbf{R}^{m_{22} \times n_{22}}$ . State the conditions on  $m_{11}, \dots, n_{22}$  under which the expression

$$M = \begin{pmatrix} A_1A_2 + B_1C_2 & A_1B_2 + B_1D_2 \\ C_1A_2 + D_1C_2 & C_1B_2 + D_1D_2 \end{pmatrix}$$

is well defined. Show explicitly that, under these conditions,  $M$  is a valid representation of  $M_1M_2$ . Verify in Matlab for 3 separate particular examples.

**Solution.** Well defined if and only if

$$n_{11} = m_{21} \quad \text{and} \quad n_{12} = m_{22}.$$

Consider  $M_{ij}$  for, e.g.,  $m_{11} < i < m_{12}$  and  $j < n_{11}$ . (This is the  $(2, 1)$  block of  $M$ ). Recall that  $M_{ij}$  is the dot product of the  $i$ th row of  $M_1$  with the  $j$ th column of  $M_2$ . Hence, if  $(\tilde{c}_{i-m_{11}}^{(1)})^T$  denotes the  $i - m_{11}$ th row of  $C_1$  and similarly for  $(\tilde{d}_{i-m_{11}}^{(1)})^T$ , then

$$M_{ij} = ((\tilde{c}_{i-m_{11}}^{(1)})^T, (\tilde{d}_{i-m_{11}}^{(1)})^T) \begin{pmatrix} a_j^{(2)} \\ c_j^{(2)} \end{pmatrix},$$

where  $a_j^{(2)}$  is the  $j$ th column of  $A_2$ , and similarly for  $c_j^{(2)}$ . This expression matches the corresponding entry in  $C_1A_2 + D_1C_2$ . The other cases for  $i$  and  $j$  are similar.

**Exercise 10.** (This exercise is a special case of the preceding one.) Show that a matrix-matrix product can always be viewed as a sum of outer products. Suppose  $A \in \mathbf{R}^{m \times k}$  and  $B \in \mathbf{R}^{k \times n}$ . Write

$$A = \begin{pmatrix} | & & | \\ a_1 & \dots & a_k \\ | & & | \end{pmatrix} \quad \text{and} \quad B = \begin{pmatrix} - & \tilde{b}_1^T & - \\ & \vdots & \\ - & \tilde{b}_k^T & - \end{pmatrix}.$$

That is,  $a_q$  is the  $q$ th column of  $A$ , and  $\tilde{b}_q^T$  is the  $q$ th row of  $B$ . Show that

$$AB = \sum_{q=1}^k a_q \tilde{b}_q^T.$$

Verify in Matlab for a few particular examples. *Hint:* It suffices to consider the  $ij$ th entry of both sides.

**Solution.** By definition,

$$(AB)_{ij} = \sum_{p=1}^k a_{ip} b_{pj}.$$

For each  $p \in \{1, \dots, k\}$ , interpret  $a_{ip}$  as the  $i$ th element of the  $p$ th column of  $A$ , and interpret  $b_{pj}$  as the  $j$ th element of the  $p$ th row of  $B$ :

$$(AB)_{ij} = \sum_{p=1}^k a_{ip} b_{pj} = \sum_{p=1}^k (a_p)_i (\tilde{b}_p^T)_j = \sum_{p=1}^k (a_p \tilde{b}_p^T)_{ij} = \left( \sum_{q=1}^k a_q \tilde{b}_q^T \right)_{ij}.$$

Alternatively, since each product  $a_q \tilde{b}_q^T$  is conformable (well defined dimensionally), the result follows by simple block matrix multiplication, where the blocks  $A_q$  of  $A$  are the columns of  $A$ , and the blocks  $B_q$  of  $B$  are the rows of  $B$ :

$$\begin{pmatrix} A_1 & A_2 & \dots & A_k \end{pmatrix} \begin{pmatrix} B_1 \\ B_2 \\ \vdots \\ B_k \end{pmatrix} = A_1 B_1 + A_2 B_2 + \dots + A_k B_k.$$

In this case, each  $A_q B_q$  happens to be an “outer product”  $a_k \tilde{b}_k^T$ , but the result is true for any partitions of  $A$  and  $B$  such that each product  $A_q B_q$  is conformable.

**Exercise 11.** Given  $A \in \mathbf{R}^{m \times n}$  with  $m < n$ ,  $b \in \mathbf{R}^m$ , and  $x_0 \in \mathbf{R}^n$  such that  $Ax_0 = b$ . Let the columns of  $Z$  form a basis for the null space of  $A$ . Show that

$$\{x \mid Ax = b\} \equiv \{x_0 + Zw \mid w \in \mathbf{R}^{n-r}\},$$

where  $r = \text{rank}(A)$ . *Hint.* To show two sets are equal, you must show that each is a subset of the other.

**Partial Solution.** If  $Ax = Ay = b$ , then  $A(x - y) = 0$ . I.e.,  $x - y \in \text{null}(A)$ , and hence by definition of  $Z$ , there exists some vector  $v$  such that  $y - x = Zv$ . Let  $x = x_0$  and hold it fixed while  $y$  takes on different solutions. Hence,

$$\{x \mid Ax = b\} \subset \{x_0 + Zw \mid w \in \mathbf{R}^{n-r}\}.$$

You must still show that the right-hand expression is a subset of the left-hand one.

**Exercise 12.** Specify whether each of the following operations on a vector  $x \in \mathbf{R}^n$  is linear or nonlinear. Prove your answers.

To show that an operation  $f(x)$  is linear in  $x$ , you have to prove that  $f(\alpha x) = \alpha f(x)$  and  $f(x + y) = f(x) + f(y)$ . But to show non-linearity, you have to only find one counterexample to one of these two properties.

- (a) Find a weighted average of the elements of  $x$  (the weights are constants).

A weighted average has the form  $f(x) = \sum_i \alpha_i x_i$  with  $\sum_i \alpha_i = 1$ . LINEAR

- (b) Find the maximum element of  $x$ .

Consider  $x = (2, 1)$  and  $y = (-2, 0)$ . NONLINEAR

- (c) Reverse the order of the elements of  $x$ .

See the lecture notes for the  $2 \times 2$  case and extend it to  $\mathbf{R}^n$ . **Hint.** Consider the “antidiagonal” matrix

$$m_{ij} = \begin{cases} 1 & \text{if } i + j = n + 1 \\ 0 & \text{otherwise} \end{cases}$$

