## Math 2600/5600 - Linear Algebra - Fall 2015

## Extra Problems and Answers/Solutions to Practice Problems for Chapter 3

**Note:** For some problems answers without full explanation are given, while for some problems complete solutions are given. On homework you are always expected to give complete solutions with full details.

**3.1.1.** See book for problem. Answers in book.

**3.1.3.** See book for problem. Incomplete answers in book.

**3.1.4.** See book for problem.

Answer: We provide a 1-1 correspondence between elementary row operations and elementary column operations to create a matrix E from I, based on how E differs from an identity matrix.

(1) Exchanging  $R_i$  and  $R_j$  in I is equivalent to exchanging  $C_i$  and  $C_j$ : these both make  $E_{ii} = E_{jj} = 0$  and  $E_{ij} = E_{ji} = 1$ .

(2) Multiplying  $R_i$  by  $\alpha$  in I is equivalent to multiplying  $C_i$  by  $\alpha$ : these both change  $E_{ii}$  from 1 to  $\alpha$ .

(3) Adding  $\beta R_i$  to  $R_j$  creates an off-diagonal entry  $E_{ji} = \beta$ . The same effect is obtained by adding  $\beta C_j$  to  $C_i$ .

**3.1.5.** See book for problem.

**Answer:** Suppose E is an elementary matrix. If it is type (1) or (2), then  $E^{T} = E$  is also elementary. If it is type (3), obtained by adding  $\beta R_{i}$  to  $R_{j}$ , then it differs from I in the off-diagonal entry  $E_{ji} = \beta$ . Thus,  $E^{T}$  differs from I in the off-diagonal entry  $(E^{T})_{ij} = \beta$ , so  $E^{T}$  can be obtained from I by adding  $\beta R_{j}$  to  $R_{i}$ , so  $E^{T}$  is also elementary.

(Similar arguments apply if we include matrices formed from I using elementary column operations as elementary matrices. Or we can just use problem 3.1.4 to say that we only need to consider elementary row operations.)

**3.2.1.** See book for problem. Answers in book.

**3.2.2.** See book for problem. Answers in book.

**3.2.8.** See book for problem.

**Answer:** Let  $P = cI_m$ , which is invertible with inverse  $(1/c)I_m$ , since  $c \neq 0$ . Hence rank  $cA = \operatorname{rank} cI_mA = \operatorname{rank} PA = \dim \operatorname{colsp} PA = \dim \operatorname{colsp} A = \operatorname{rank} A$ . We used the fact that  $\operatorname{colsp} PA \cong \operatorname{colsp} A$ , so that  $\dim \operatorname{colsp} PA = \dim \operatorname{colsp} A$ . (There are other ways to do this.)

3.2.14. See book for problem.

**Answer:** (a) By X + Y we mean the set  $\{x + y \mid x \in X, y \in Y\}$ .

We want to show that  $R(T+U) \subseteq R(T) + R(U)$ . Let  $w \in R(T+U)$ , then w = (T+U)v = Tv + Uv for some  $v \in V$ . But  $Tv \in R(T)$  and  $Uv \in R(U)$ , so  $w = Tv + Uv \in R(T) + R(U)$ . Hence,  $R(T+U) \subseteq R(T) + R(U)$ .

(b) Note that for any sets X, Y we have span  $X + \text{span } Y \subseteq \text{span } (X \cup Y)$ : a linear combination of things in X plus a linear combination of things in Y is just a linear combination of things in X and Y together.

If W is finite-dimensional then so are its subspaces R(T), R(U), R(T+U), and so T, U, U+T all have finite rank. Let X be a basis for R(T) and Y be a basis for R(U). Then  $R(T+U) \subseteq R(U) + R(T) =$ span X + span Y  $\subseteq$  span  $(X \cup Y)$  and so rank  $(T+U) = \dim R(T+U) \leq \dim \text{span } (X \cup Y) \leq |X \cup Y| \leq |X| + |Y| = \text{rank } T + \text{rank } U$ .

(c) rank  $(A + B) = \operatorname{rank} L_{A+B} = \operatorname{rank} (L_A + L_B) \leq \operatorname{rank} L_A + \operatorname{rank} L_B = \operatorname{rank} A + \operatorname{rank} B$ .

3.2.17. See book for problem.

**Answer:** We see that rank  $A = \dim \operatorname{colsp} A$  is the maximum number of linearly independent columns of A. We have

$$BC = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \begin{bmatrix} c_1 & c_2 & c_3 \end{bmatrix} = \begin{bmatrix} b_1c_1 & b_1c_2 & b_1c_3 \\ b_2c_1 & b_2c_2 & b_2c_3 \\ b_3c_1 & b_3c_2 & b_3c_3 \end{bmatrix} = \begin{bmatrix} c_1B & c_2B & c_3B \end{bmatrix}$$

Since the columns of BC are all multiples of B, at most one is linearly independent (maybe none, if they are all 0). So rank  $BC \leq 1$ .

Conversely, suppose A is  $3 \times 3$  and rank A = 1. That means there is at most one linearly independent column. So either A is 0, or all columns of A are multiples of one nonzero column. In either case, there is a column vector B so that the columns of A can be regarded as  $c_1B, c_2B, c_3B$  for some scalars  $c_1, c_2, c_3$  (take  $c_i = 1$  if the *i*th column is the one that the other columns are multiples of). But from above this means A = BC where  $C = [c_1 \quad c_2 \quad c_3]$ .

**3.2.5.** See book for problem. Answers in book.

**3.2.6.** See book for problem. Answers in book.

3.2.21. See book for problem.

**Answer:** Suppose A is  $m \times n$  with rank m. Then dim colsp A = m, so the columns of A span  $\mathbb{R}^m$ . So we can set up an  $n \times m$  matrix B whose j-th column tells how to combine the columns of A to give the j-th standard basis vector  $e_j \in \mathbb{R}^m$ . Then  $AB = [e_1 \mid e_2 \mid \ldots \mid e_m] = I_m$ .

**3.3.1.** See book for problem. Answers in book.

**3.3.2.** See book for problem. Answers in book.

**3.3.3.** See book for problem. Answers in book. (Note that answers can also be written in other ways.)

**3.3.6.** See book for problem. Answer in book. (Note that answer can also be written in other ways.)

**3.3.10.** See book for problem.

**Answer:** This is true. If rank A = m then  $m = \operatorname{rank} A \leq \operatorname{rank} [A \mid b] \leq m$ , so rank  $[A \mid b] = m = \operatorname{rank} A$  and the system is consistent.

Another way to say this is that if rank A = m, then the columns of A form a spanning set for  $F^m$ , so there is certainly some linear combination of columns of A giving b, or in other words, there is some x so that Ax = b.

**3.4.1.** See book for problem. Answers in the book.

**3.4.2.** See book for problem. Answers in the book.

**3.4.4.** See book for problem. Answers in the book.

**3.4.5.** See book for problem. Answer in the book.

**3.4.7.** See book for problem. Answer in the book.

3.4.9. See book for problem.

**Answer:** Represent S in terms of the standard ordered basis of  $\mathbf{R}^{2\times 2}$ ,  $B = (E_{11}, E_{12}, E_{21}, E_{22})$  where  $E_{ij}$  is 0 except for a 1 in row *i*, column *j*. Put the coordinate vectors as the columns of a matrix A and reduce A to reduced row echelon form R:

A =	Γ 0	)	1	2	1	-1 J	$\rightarrow$	R =	Γ1	0	3	0	4 J
	-1		2	1	-2	2			0	1	2	0	1
	-1		2	1	-2	2			0	0	0	1	-2
	L 1		3	9	4	-1			Lo	0	0	0	0

We have an isomorphism from W to colsp A by taking coordinates, and an isomorphism from colsp A to colsp R from the elementary row operations that reduced A to R, so their composition is an isomorphism from W to colsp R, in which the *i*th element of S maps to the *i*th column of R.

The columns corresponding to the leading entries, namely columns 1, 2, 4 form a basis for colsp R. Hence the first, second and fourth elements of S form a basis for W, namely  $\left\{ \begin{bmatrix} 0 & -1 \\ -1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix}, \begin{bmatrix} 1 & -2 \\ -2 & 4 \end{bmatrix} \right\}$ .

**3.4.13.** See book for problem. Answer to (b) in the book.

**Answer:** (a) Substitute vectors into equations to show they are in V; they are not multiples of each other so they are linearly independent.