

MATH 221 Solutions to the sample final

- 2 (a). True. $c_i \vec{v}_i \cdot c_j \vec{v}_j = 0$ iff $i \neq j$.
- (b). True. This is the Pythagorean theorem. For an argument, note that $\|\vec{v} - \vec{u}\|^2 = \|\vec{v}\|^2 + \|\vec{u}\|^2 - 2\vec{u} \cdot \vec{v}$. Hence the assumption forces $\vec{u} \cdot \vec{v} = 0$ so that \vec{u} and \vec{v} are orthogonal.
- (c). False. This is not true since the $\vec{0}$ vector is orthogonal to any vector. For an example, note that $\{(1, 0), (0, 0)\}$ is an orthogonal subset of \mathbb{R}^2 . Otherwise this is true if we consider non-zero orthogonal vectors; this is a theorem from class.
- (d). False. Let $A = \begin{bmatrix} 1 & 0 \end{bmatrix}$ so A is a 1×2 matrix. The Nullspace of A is spanned by $(0, 1)$ while the column space is a subspace of \mathbb{R}^1 . What is true is that the orthogonal complement of the row space of A is the nullspace of A .
- (e). True. A theorem from class.
- (f). True. If \vec{x} is in W^\perp then \vec{x} is orthogonal to every vector in W and hence the \vec{v}_j 's. If \vec{x} is orthogonal to the \vec{v}_j 's then \vec{x} is orthogonal to linear combinations of the \vec{v}_j 's and hence every vector in W .
- (g). True. Suppose that a vector \vec{w} is in W and W^\perp . Then $\vec{w} \cdot \vec{w} = 0$ implies $\|\vec{w}\| = 0$ which implies $\vec{w} = \vec{0}$ (since vectors of zero length are zero).
- (h). True by uniqueness of orthogonal projection (a theorem from class).
- (i). True. If $(A^T)A = I$, then A is invertible and $A^{-1} = A^T$. Thus $AA^{-1} = AA^T = I$, which means that the columns of A^T are orthonormal and hence the rows of A are orthonormal.
- (j). False. $P = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$ has orthogonal columns, but P^{-1} is not equal to P^T .
- (k). True. A theorem from class.
- (l). False. Let $\vec{y} = (1, 1)$ and $\vec{u} = (0, 1)$. Then $\text{proj}_{\text{span}(\vec{u})}(\vec{y}) = \vec{u}$ and \vec{u} is not a multiple of \vec{y} .
- (m). True by uniqueness of orthogonal decomposition.

3) If $i \neq j$ then $\vec{v}_i \cdot \vec{v}_j = 0$. Since U is orthogonal, the corresponding linear transformation preserves the dot product (angles), thus $(U\vec{v}_i) \cdot (U\vec{v}_j) = \vec{v}_i \cdot \vec{v}_j = 0$. Thus the set $\{U\vec{v}_1, U\vec{v}_2, \dots, U\vec{v}_n\}$ is orthogonal. To see that this set is orthonormal, the same argument as above shows that $\|U\vec{v}_i\|^2 = (U\vec{v}_i) \cdot (U\vec{v}_i) = \vec{v}_i \cdot \vec{v}_i = \|\vec{v}_i\|^2 = 1$ (or just notice that the linear transformation associated with U also preserves lengths).

4) Take $\vec{x} = \vec{e}_i$. Then $U\vec{x}$ is the i^{th} column of U . Take $\vec{y} = \vec{e}_j$, so that $U\vec{y}$ is the j^{th} column of U . Write \vec{u}_i for the i^{th} column of U for $i = 1, \dots, n$. Then $\vec{u}_i \cdot \vec{u}_j = (U\vec{x}) \cdot (U\vec{y}) = \vec{x} \cdot \vec{y} = \vec{e}_i \cdot \vec{e}_j$. Thus $\vec{u}_i \cdot \vec{u}_j$ is zero if $i \neq j$ and is 1 if $i = j$. This means the columns of U are orthonormal, and hence that $(U^T)U = I$ so that $U^{-1} = U^T$.

5) Suppose that \vec{x} and \vec{y} are each orthogonal to the given vector \vec{v} , and suppose that c and d are scalars. To show that $c\vec{x} + d\vec{y}$ is in the subspace, one needs to see that $c\vec{x} + d\vec{y}$ is orthogonal to \vec{v} . Well, $(c\vec{x} + d\vec{y}) \cdot \vec{v} = c(\vec{x} \cdot \vec{v}) + d(\vec{y} \cdot \vec{v}) = c \cdot 0 + d \cdot 0 = 0$ since $\vec{x} \cdot \vec{v} = \vec{y} \cdot \vec{v} = 0$. This shows that $\{\vec{x} : \vec{x} \cdot \vec{v} = 0\}$ is a subspace.

6) a) Let A be the matrix with columns $\vec{v}_1, \vec{v}_2, \vec{v}_3$. One finds that A has two pivots, in the first two columns. So column space of A has dimension 2.

b) Note that V is spanned by \vec{v}_1 and \vec{v}_2 . A vector $\vec{x} = (a, b, c)$ is orthogonal to V iff \vec{x} is orthogonal to \vec{v}_1 and to \vec{v}_2 . Thus, \vec{x} is in V^\perp iff $\begin{bmatrix} 1 & 1 & 1 \\ 1 & -2 & -2 \end{bmatrix} \vec{x} = \vec{0}$. So we are reduced to finding the Null space of the above 2×3 matrix. The reduced echelon form of this matrix is $\begin{bmatrix} 1 & 0 & \frac{4}{3} \\ 0 & 1 & \frac{-1}{3} \end{bmatrix}$. So V^\perp is spanned by $(-\frac{4}{3}, \frac{1}{3}, 1)$ or by $(-4, 3, 1)$.

c) \vec{c} is in V iff \vec{c} is equal to its orthogonal projection onto V . Since the orthogonal complement of V is spanned by a single vector, \vec{c} is in V iff \vec{c} is orthogonal to $(-4, 3, 1)$. Thus the condition is that $-4c_1 + c_2 + 3c_3 = 0$. If there is a particular solution $\vec{p} = (x, y, z)$, then $\vec{p} + \vec{v}$ is also a solution for any vector \vec{v} in the Nullspace of A . Since the nullspace of A is 1 dimensional there are infinitely many solutions in this case.

(7) (a) The matrix formed by the 3 vectors may be row reduced:

$$\begin{pmatrix} 1 & 1 & 0 \\ 2 & 1 & 2 \\ 1 & 0 & 2 \end{pmatrix} \sim \begin{pmatrix} 1 & 1 & 0 \\ 0 & -1 & 2 \\ 0 & -1 & 2 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & 2 \\ 0 & -1 & 2 \\ 0 & 0 & 0 \end{pmatrix}$$

hence the dimension of the column space is 2 and is actually spanned by the vectors $\mathbf{x}_1 = (1, 2, 1)^T$ and $\mathbf{x}_2 = (1, 1, 0)^T$. (b) To find an orthogonal basis $\mathbf{u}_1, \mathbf{u}_2$ with the same span we may take $\mathbf{u}_1 = \mathbf{x}_1$ and

$$\mathbf{u}_2 = \mathbf{x}_2 - \frac{\mathbf{x}_2 \cdot \mathbf{x}_1}{\mathbf{x}_1 \cdot \mathbf{x}_1} \mathbf{x}_1 = (1, 1, 0)^T - \frac{1}{2}(1, 2, 1)^T = (1/2, 0, -1/2)^T.$$

(8) (a) It is clear that $\mathbf{v}_1 \cdot \mathbf{v}_2 = \mathbf{v}_1 \cdot \mathbf{v}_3 = \mathbf{v}_2 \cdot \mathbf{v}_3 = 0$. (b) an orthogonal set is necessarily linearly independent so the dimension of the span W of these 3 vectors is 3, by the definition of dimension. (c) the projection of a vector \mathbf{x} to the space W is given by

$$\text{proj}_W \mathbf{x} = \frac{\mathbf{x} \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 + \frac{\mathbf{x} \cdot \mathbf{v}_2}{\mathbf{v}_2 \cdot \mathbf{v}_2} \mathbf{v}_2 + \frac{\mathbf{x} \cdot \mathbf{v}_3}{\mathbf{v}_3 \cdot \mathbf{v}_3} \mathbf{v}_3 = -\mathbf{v}_1 + 2\mathbf{v}_3 = (-1, -1, 2, -2)^T.$$

(9) (a) The matrix formed by the 3 vectors (as row vectors) can be row reduced to

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 3 & 3 \end{pmatrix} \sim \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 2 & 2 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix}.$$

The null space V^\perp is spanned by the vector $(0, 1, -1)^T$. (b) The matrix formed by the 3 vectors (as column vectors) can be row reduced to

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{pmatrix} \sim \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix}.$$

the equation $A\mathbf{x} = \mathbf{c} = (c_1, c_2, c_3)^T$ is consistent if and only if \mathbf{c} is a linear combination of the vectors $\mathbf{u}_1 = (1, 1, 1)$, $\mathbf{u}_2 = (1, 2, 2)$, i. e.,

$$c_1 = a + b, c_2 = a + 2b, c_3 = a + 2b, a, b \in \mathbf{R}.$$

(c) infinitely many.

(10) The points are $(-1, 1), (0, 1), (1, 0), (2, 2)$,

$$A = \begin{pmatrix} 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 2 & 4 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix},$$

$$A^T = \begin{pmatrix} 1 & 1 & 1 & 1 \\ -1 & 0 & 1 & 2 \\ 1 & 0 & 1 & 4 \end{pmatrix}, A^T \mathbf{b} = \begin{pmatrix} 1 & 1 & 1 & 1 \\ -1 & 0 & 1 & 2 \\ 1 & 0 & 1 & 4 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 0 \\ 2 \end{pmatrix} = \begin{pmatrix} 4 \\ 3 \\ 9 \end{pmatrix},$$

$$A^T A = \begin{pmatrix} 1 & 1 & 1 & 1 \\ -1 & 0 & 1 & 2 \\ 1 & 0 & 1 & 4 \end{pmatrix} \begin{pmatrix} 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 2 & 4 \end{pmatrix} = \begin{pmatrix} 4 & 2 & 6 \\ 2 & 6 & 8 \\ 6 & 8 & 18 \end{pmatrix}.$$

To find $\hat{\mathbf{x}}$ row reduce the augmented matrix,

$$\begin{pmatrix} 4 & 2 & 6 & 4 \\ 2 & 6 & 8 & 3 \\ 6 & 8 & 18 & 9 \end{pmatrix}.$$

(11) The vectors are $\mathbf{v}_1 = (-1, 3, 1, 1)$, $\mathbf{v}_2 = (6, -8, -2, -4)$, $\mathbf{v}_3 = (6, 3, 6, -3)$. First find an orthogonal basis $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$ with $\mathbf{u}_1 = \mathbf{v}_1$,

$$\mathbf{u}_2 = (6, -8, -2, -4) + 3(-1, 3, 1, 1) = (3, 1, 1, -1)$$

$$\mathbf{u}_3 = (6, 3, 6, -3) - \frac{1}{2}(-1, 3, 1, 1) - \frac{5}{2}(3, 1, 1, -1) = (-1, -1, 3, -1).$$

Then normalize these vectors:

$$\mathbf{w}_1 = \frac{\mathbf{u}_1}{2\sqrt{3}}, \mathbf{w}_2 = \frac{\mathbf{u}_2}{2\sqrt{3}}, \mathbf{w}_3 = \frac{\mathbf{u}_3}{2\sqrt{3}}.$$

Now form a matrix Q with $\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3$ as column vectors:

$$Q = \frac{1}{2\sqrt{3}} \begin{pmatrix} -1 & 3 & -1 \\ 3 & 1 & -1 \\ 1 & 1 & 3 \\ 1 & -1 & -1 \end{pmatrix}$$

Then $R = Q^T A$ where A is the matrix formed by taking the 3 original vectors $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ as column vectors:

$$A = \begin{pmatrix} -1 & 6 & 6 \\ 3 & -8 & 3 \\ 1 & -2 & 6 \\ 1 & -4 & -3 \end{pmatrix}$$

$$R = \frac{1}{2\sqrt{3}} \begin{pmatrix} -1 & 3 & 1 & 1 \\ 3 & 1 & 1 & -1 \\ -1 & -1 & 3 & -1 \end{pmatrix} \begin{pmatrix} -1 & 6 & 6 \\ 3 & -8 & 3 \\ 1 & -2 & 6 \\ 1 & -4 & -3 \end{pmatrix} = \frac{1}{2\sqrt{3}} \begin{pmatrix} 12 & -36 & 6 \\ 0 & 12 & 30 \\ 0 & 0 & 12 \end{pmatrix}.$$

(12) First compute

$$A^T A = \begin{pmatrix} 1 & 2 & -1 \\ 2 & 4 & -2 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 2 & 4 \\ -1 & -2 \end{pmatrix} = \begin{pmatrix} 6 & 12 \\ 12 & 24 \end{pmatrix}$$

$$A^T \mathbf{b} = \begin{pmatrix} 1 & 2 & -1 \\ 2 & 4 & -2 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 6 \\ 12 \end{pmatrix}$$

then solve the equation $A^T A \mathbf{x} = A^T \mathbf{b}$

$$\begin{pmatrix} 6 & 12 \\ 12 & 24 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 6 \\ 12 \end{pmatrix}$$

hence $x_1 = 1, x_2 = 0$.