Week 10 Homework Summary

- I. From Monday: 1, 4, 5 on p. 55:
 - 1. Show that $Null(A^T) \perp Col(A)$.

SOLUTION: Let $B = A^T$. Then we prove that $\text{Null}(B) \perp \text{Row}(B)$ which is equivalent.

If we write B in terms of rows, $B = [r_1, r_2, \dots, r_m]^T$, then the i^{th} component of $B\mathbf{x}$ can be written as $\mathbf{r}_i\mathbf{x} = 0$. Thus, if \mathbf{x} is in the nullspace of B, then it is orthogonal to every row of B.

4. Prove the Pythagorean Theorem by induction: Given a set of n orthogonal vectors $\{x_i\}$

$$\|\sum_{i=1}^n m{x}_i\|_2^2 = \sum_{i=1}^n \|m{x}_i\|_2^2$$

The case where n=1 is trivial, so you might look at n=2 first. Try starting with

$$\|\mathbf{x} + \mathbf{y}\|^2 = (\mathbf{x} + \mathbf{y})^T (\mathbf{x} + \mathbf{y}) = \cdots$$

and then simplify to get $\|\mathbf{x}\|^2 + \|\mathbf{y}\|^2$. Now try the induction step on your own. SOLUTION:

The case with n = 1 is trivial. The case with n = 2 is straightforward. Assume true for k = n,

$$\|\sum_{i=1}^n m{x}_i\|^2 = \sum_{i=1}^n \|m{x}_i\|^2$$

Show true for k = n + 1:

$$\|\sum_{i=1}^{n+1} oldsymbol{x}_i\|^2 = \left(\sum_{i=1}^n oldsymbol{x}_i + oldsymbol{x}_{n+1}
ight)^T \left(\sum_{i=1}^n oldsymbol{x}_i + oldsymbol{x}_{n+1}
ight) = \left(\sum_{i=1}^n oldsymbol{x}_i
ight)^T \left(\sum_{i=1}^n oldsymbol{x}_i
ight) + \left(\sum_{i=1}^n oldsymbol{x}_i
ight) oldsymbol{x}_{n+1}^T + oldsymbol{x}_{n+1} \left(\sum_{i=1}^n oldsymbol{x}_i
ight)^T + oldsymbol{x}_{n+1}^T oldsymbol{x}_{n+1}$$

By the induction step, that first term is:

$$\left(\sum_{i=1}^n oldsymbol{x}_i
ight)^T \left(\sum_{i=1}^n oldsymbol{x}_i
ight) = \sum_{i=1}^n \|\mathbf{x}_i\|^2$$

The second and third terms are zero since \mathbf{x}_{n+1} is orthogonal to the first n vectors, and the last term is $\|\mathbf{x}_{n+1}\|^2$. Therefore, the expression simplifies to the desired value,

$$\|\sum_{i=1}^{n+1} m{x}_i\|^2 = \sum_{i=1}^{n+1} \|m{x}_i\|^2$$

5. Let A be an $m \times n$ matrix where m > n, and let A have rank n. Let $\mathbf{y}, \hat{\mathbf{y}} \in \mathbb{R}^m$, such that $\hat{\mathbf{y}}$ is the orthogonal projection of \mathbf{y} onto the column space of A. We want a formula for the matrix $\mathbb{P} : \mathbb{R}^m \to \mathbb{R}^m$ so that $\mathbb{P}\mathbf{y} = \hat{\mathbf{y}}$.

The following image shows the relevant subspaces (See page 56 for the graph)

(a) Why is the projector not $\mathbb{P} = AA^T$? SOLUTION: If the columns of A were orthonormal, it would be. In class, we found the actual projection matrix to be the one below, which is:

$$P = A(A^T A)^{-1} A^T$$

which we are showing now.

(b) Since $\hat{y} - y$ is orthogonal to the column space of A, show that

$$A^{T}(\hat{\boldsymbol{y}} - \boldsymbol{y}) = \boldsymbol{0} \tag{1}$$

SOLUTION: We know that $Col(A) \perp Null(A^T)$. Therefore, if $\hat{\boldsymbol{y}} - \boldsymbol{y}$ is orthogonal to the column space of A, then it must be in the null space of A^T .

(c) Show that there exists $x \in \mathbb{R}^n$ so that Equation (1) can be written as:

$$A^{T}(A\boldsymbol{x} - \boldsymbol{y}) = 0 \tag{2}$$

SOLUTION: Since \hat{y} is in the columnspace of A, by definition there exists a $x \in \mathbb{R}^n$ such that $Ax = \hat{y}$.

(d) Argue that A^TA (which is $n \times n$) is invertible, so that Equation (2) implies that

$$\boldsymbol{x} = \left(A^T A\right)^{-1} A^T \boldsymbol{y}$$

SOLUTION: The matrix A has rank n, therefore A^TA is an $n \times n$ matrix with rank n as well (the ranks are equal). Therefore, A^TA is invertible, and we get the **pseudoinverse**, $A^{\dagger} = (A^TA)^{-1}A^T$.

(e) Finally, show that this implies that

$$\mathbb{P} = A \left(A^T A \right)^{-1} A^T$$

SOLUTION: Multiply both sides of the previous equation by A.

- II. From Tuesday: 1, 3, 4, 6, 7, 8 on p. 53 (Yes, we were working a little backwards).
 - 1. Let the subspace H be formed by the span of the vectors $\mathbf{v}_1, \mathbf{v}_2$ given below. Given the point $\mathbf{x}_1, \mathbf{x}_2$ below, find which one belongs to H, and if it does, give its coordinates. (NOTE: The basis vectors are NOT orthonormal)

$$oldsymbol{v}_1 = \left[egin{array}{c} 1 \ 2 \ -1 \end{array}
ight] \quad oldsymbol{v}_2 = \left[egin{array}{c} 2 \ -1 \ 1 \end{array}
ight] \quad oldsymbol{x}_1 = \left[egin{array}{c} 7 \ 4 \ 0 \end{array}
ight] \quad oldsymbol{x}_2 = \left[egin{array}{c} 4 \ 3 \ -1 \end{array}
ight]$$

SOLUTION: Rather than row-reduce twice, we'll do it once on the augmented matrix below.

$$\begin{bmatrix} 1 & 2 & 7 & 4 \\ 2 & -1 & 4 & 3 \\ -1 & 1 & 0 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 3 & 2 \\ 0 & 1 & 2 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

How should this be interpreted? The second vector, \mathbf{x}_2 is in H, as it can be expressed as $2\mathbf{v}_1 + \mathbf{v}_2$. Its low dimensional representation is thus $[2, 1]^T$.

The first vector, x_1 , cannot be expressed as a linear combination of v_1 and v_2 , so it does not belong to H.

3. Let the subspace G be the plane defined below, and consider the vector \boldsymbol{x} , where:

$$G = \left\{ \alpha_1 \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix} + \alpha_2 \begin{bmatrix} 3 \\ -1 \\ 0 \end{bmatrix} \text{ such that } \alpha_1, \alpha_2 \in \mathbb{R} \right\} \qquad \boldsymbol{x} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$

(Skip part (a))

b. Find the orthogonal projection of the given \boldsymbol{x} onto the plane G (note that the vectors are orthogonal).

SOLUTION: Let the vectors given (normalize them) be \mathbf{u}_1 and \mathbf{u}_2 . Then we compute:

$$\operatorname{Proj}_{G}(\mathbf{x}) = (\mathbf{x} \cdot \mathbf{u}_{1})\mathbf{u}_{1} + (\mathbf{x} \cdot \mathbf{u}_{2})\mathbf{u}_{2} = \frac{-3}{14} \begin{bmatrix} 1\\3\\2 \end{bmatrix} + \frac{3}{10} \begin{bmatrix} 3\\-1\\0 \end{bmatrix} \approx \begin{bmatrix} 0.686\\-0.943\\0.429 \end{bmatrix}$$

(Added) Show that the formula in 5(e) from the previous page works for this problem as well.

SOLUTION: I shouldn't have added this- This was enough computation.

c. Find the distance from the plane G to the vector x.

SOLUTION: The distance is the length of the vector going from \mathbf{x} to the projection of \mathbf{x} in G, or

$$\left\| \begin{array}{c} 1 - 0.686 \\ 0.943 \\ 2 - 0.429 \end{array} \right\| \approx 1.859$$

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4. If the low dimensional representation of a vector \boldsymbol{x} is $[9, -1]^T$ and the basis vectors are $[1, 0, 1]^T$ and $[3, 1, 1]^T$, then what was the original vector \boldsymbol{x} ? (HINT: it is easy to compute it directly)

SOLUTION: $[6, -1, 8]^T$.

6. Let $\boldsymbol{a} = [-1, 3]^T$. Find a square matrix P so that $P\boldsymbol{x}$ is the orthogonal projection of \boldsymbol{x} onto the span of \boldsymbol{a} .

SOLUTION: We're thinking of the outer product divided by the inner product, or

 $P = \frac{1}{10} \left[\begin{array}{cc} 1 & -3 \\ -3 & 9 \end{array} \right]$

- 7. (Reworded from the text) Suppose we have vectors \mathbf{x} , \mathbf{u} , and the projection of \mathbf{x} onto \mathbf{u} (Call it $\hat{\mathbf{x}}$).
 - If the projection is *orthogonal*, then which vectors should be perpendicular? SOLUTION: The vector $\mathbf{x} \hat{\mathbf{x}}$ should be orthogonal to \mathbf{u} .
 - If $\hat{\mathbf{x}} = c\mathbf{u}$, find the constant c from your previous answer. SOLUTION: The constant comes from the orthogonality

$$(\mathbf{x} - c\mathbf{u}) \cdot \mathbf{u} = 0 \quad \Rightarrow \quad \mathbf{x} \cdot \mathbf{u} = c\mathbf{u} \cdot \mathbf{u} \quad \Rightarrow \quad c = \frac{\mathbf{x} \cdot \mathbf{u}}{\mathbf{u} \cdot \mathbf{u}}$$

8. Continuing with the last exercise, show that $UU^T\mathbf{x}$ is the *orthogonal* projection of \mathbf{x} into the space spanned by the columns of U by showing that $(UU^T\mathbf{x} - \mathbf{x})$ is orthogonal to \mathbf{u}_i for any $i = 1, 2, \dots, k$.

SOLUTION: Notice that

$$UU^T = \mathbf{u}_1 \mathbf{u}_1^T + \mathbf{u}_2 \mathbf{u}_2^T + \dots + \mathbf{u}_k \mathbf{u}_k^T$$

so that

$$(UU^T\mathbf{x} - \mathbf{x}) \cdot \mathbf{u}_i = [(0 + 0 + \dots + 0) + (\mathbf{u}_i \cdot \mathbf{u}_i)(\mathbf{u}_i \cdot \mathbf{x}) + (0 + 0 + \dots + 0)] - \mathbf{x} \cdot \mathbf{u}_i = 0$$

III. Homework from Thursday, Nov 6th

 ${\bf 1}$. Assume that A and B are row equivalent, where:

$$A = \begin{bmatrix} 1 & 2 & -2 & 0 & 7 \\ -2 & -3 & 1 & -1 & -5 \\ -3 & -4 & 0 & -2 & -3 \\ 3 & 6 & -6 & 5 & 1 \end{bmatrix}, \qquad B = \begin{bmatrix} 1 & 0 & 4 & 0 & -3 \\ 0 & 1 & -3 & 0 & 5 \\ 0 & 0 & 0 & 1 & -4 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

(a) State which vector space (\mathbb{R}^2) contains each of the four subspaces, and state the dimension of each of the four subspaces:

SOLUTION: Here are the spaces and dimensions:

- A. The row space of A is a subspace of \mathbb{R}^5 and has dimension 3 (this is the number of pivot rows).
- B. The column space of A is a subspace of \mathbb{R}^4 and has dimension 3 (this is also known as the rank of A and is the number of pivot columns).
- C. The null space of A is a subspace of \mathbb{R}^5 and has dimension 2 (the number of free variables).
- D. The null space of A^T is a subspace of \mathbb{R}^4 and has dimension 1 (because the dimension of the column space is 3, and they should add to 4).
- (b) Find a basis for Col(A):

SOLUTION: We see from the RREF that the pivot columns are cols 1, 2, and 4. Be sure to use the columns from the original matrix A! The basis is the set

$$\left\{ \begin{bmatrix} 1\\-2\\-3\\3 \end{bmatrix}, \begin{bmatrix} 2\\-3\\-4\\6 \end{bmatrix}, \begin{bmatrix} 0\\-1\\-2\\5 \end{bmatrix} \right\}$$

(c) Find a basis for Row(A):

SOLUTION: We can use the reduced rows (written as columns) for the basis:

$$\left\{ \begin{bmatrix} 1\\0\\4\\0\\-3 \end{bmatrix}, \begin{bmatrix} 0\\1\\-3\\0\\5 \end{bmatrix}, \begin{bmatrix} 0\\0\\0\\1\\-4 \end{bmatrix} \right\}$$

(d) Find a basis for Null(A):

SOLUTION: Solve $A\mathbf{x} = \mathbf{0}$, and the vectors are the basis vectors:

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$$\begin{array}{rcl}
x_1 & = -4x_3 + 3x_5 \\
x_2 & = 3x_3 - 5x_5 \\
x_3 & = x_3 \\
x_4 & = & 4x_5 \\
x_5 & = & x_5
\end{array}
\Rightarrow
\begin{cases}
\begin{bmatrix}
-4 \\
3 \\
1 \\
0 \\
0
\end{bmatrix}, \begin{bmatrix}
3 \\
-5 \\
0 \\
4 \\
1
\end{bmatrix}
\end{cases}$$

2. From Monday's HW, we said that if we're solving for $A\mathbf{x} = \mathbf{y}$ and A has linearly independent columns, then we can solve for \mathbf{x} using the **normal equations** (just as we did in the linear regression problem):

$$A\mathbf{x} = \mathbf{y} \quad \Rightarrow \quad A^T A\mathbf{x} = A^T \mathbf{y}$$

And since A has full rank (it has linearly independent columns), then

$$\mathbf{x} = (A^T A)^{-1} A^T \mathbf{y}$$

The matrix $(A^TA)^{-1}A^T$ is called the *pseudoinverse* of A, and is denoted by A^{\dagger} . Find the pseudoinverse matrix of A, if

$$A = \left[\begin{array}{cc} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{array} \right]$$

SOLUTION: This is just a matter of multiplying it out. See the exam review sheet.

3. Also from Monday's HW, we said that if

$$\mathbf{x} = (A^T A)^{-1} A^T \mathbf{y}$$

then

$$A\mathbf{x} = A(A^T A)^{-1} A^T \mathbf{y}$$

Since $A\mathbf{x}$ is in the column space of A, then the vector on the right side of the equation must be the orthogonal projection of \mathbf{y} into the column space of A. Therefore, the matrix that takes \mathbf{y} and produces the projection of \mathbf{y} is:

$$P = A(A^T A)^{-1} A^T$$

Find the matrix P using the matrix A from the previous exercise.

SOLUTION: It's just a matter of multiplying it out. See the exam review sheet.