## EECS 16B Designing Information Devices and Systems II Fall 2021 Discussion Worksheet Discussion 8B

The following notes are useful for this discussion: Note 12.

## 1. Gram-Schmidt Algorithm

Let's apply Gram-Schmidt orthonormalization to a list of three linearly independent vectors  $[\vec{s}_1, \vec{s}_2, \vec{s}_3]$ .

(a) Find unit vector  $\vec{q_1}$  such that  $\text{Span}(\{\vec{q_1}\}) = \text{Span}(\{\vec{s_1}\})$ .

(b) Given  $\vec{q_1}$  from the previous step, find unit vector  $\vec{q_2}$  such that  $\operatorname{Span}(\{\vec{q_1}, \vec{q_2}\}) = \operatorname{Span}(\{\vec{s_1}, \vec{s_2}\})$  and  $\vec{q_2}$  is orthogonal to  $\vec{q_1}$ .

(c) Suppose we want to show that  $\text{Span}(\{\vec{q_1}, \vec{q_2}\}) = \text{Span}(\{\vec{s_1}, \vec{s_2}\})$ . What does this mean mathematically? *Hint: you cannot use the word span, but must capture the same concept in your translation of the statement we want to show.* 

- (d) What would happen if  $\{\vec{s}_1, \vec{s}_2, \vec{s}_3\}$  were *not* linearly independent, but rather  $\vec{s}_1$  were a multiple of  $\vec{s}_2$ ?
- (e) Now given  $\vec{q_1}$  and  $\vec{q_2}$  in parts (a) and (b), find  $\vec{q_3}$  such that  $\text{Span}(\{\vec{q_1}, \vec{q_2}, \vec{q_3}\}) = \text{Span}(\{\vec{s_1}, \vec{s_2}, \vec{s_3}\})$ , and  $\vec{q_3}$  is orthogonal to both  $\vec{q_1}$  and  $\vec{q_2}$ , and finally  $\|\vec{q_3}\| = 1$ .

(f) **[Practice] Confirm that**  $\text{Span}(\{\vec{q_1}, \vec{q_2}, \vec{q_3}\}) = \text{Span}(\{\vec{s_1}, \vec{s_2}, \vec{s_3}\}).$ 

## 2. Orthonormal Matrices and Projections

An orthonormal matrix, A, is a matrix whose columns,  $\vec{a}_i$ , are:

- Orthogonal (ie.  $\langle \vec{a}_i, \vec{a}_j \rangle = \vec{a}_j^\top \vec{a}_i = 0$  when  $i \neq j$ )
- Normalized (ie. vectors with length equal to 1,  $\|\vec{a}_i\| = 1$ ). This implies that  $\|\vec{a}_i\|_2 = \langle \vec{a}_i, \vec{a}_i \rangle = \vec{a}_i^\top \vec{a}_i = 1$ .
- (a) When  $A \in \mathbb{R}^{n \times m}$  and  $n \ge m$  (i.e. for tall matrices), show that if the matrix is orthonormal, then  $A^{\top}A = I_{m \times m}$ .

(b) Again, suppose  $A \in \mathbb{R}^{n \times m}$  where  $n \ge m$  is an orthonormal matrix. Show that the projection of  $\vec{y}$  onto the subspace spanned by the columns of A is now  $AA^{\top}\vec{y}$ .

(c) Show if  $A \in \mathbb{R}^{n \times n}$  is an orthonormal matrix then the columns,  $\vec{a}_i$ , form a basis for  $\mathbb{R}^N$ . *Hint: can you use the result of the previous subpart?* 

## **Contributors:**

- Regina Eckert.
- Druv Pai.
- Neelesh Ramachandran.