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## Thm has a name

### Chapter 7

**Theorem 1.1.1** (Dimension Thm). If  $A$  is an  $m \times n$  matrix, then  $\text{rank}A + \dim \text{Null}(A) = n$ .

### Chapter 8

**Theorem 1.2.1** (Rank-Nullity Thm). Let  $\mathbb{V}$  be an  $n$ -dimensional vector space and let  $\mathbb{W}$  be a vector space. If  $L : \mathbb{V} \rightarrow \mathbb{W}$  is linear, then

$$\text{rank}(L) + \text{nullity}(L) = n$$

### Chapter 9

**Theorem 1.3.1** (Gram-Schmidt Orthogonalization Thm). Let  $\{\vec{w}_1, \dots, \vec{w}_n\}$  be a basis for an inner product space  $\mathbb{W}$ . If we define  $\vec{v}_1, \dots, \vec{v}_n$  successively as follows:

$$\begin{aligned}\vec{v}_1 &= \vec{w}_1 \\ \vec{v}_2 &= \vec{w}_2 - \frac{\langle \vec{w}_2, \vec{v}_1 \rangle}{\|\vec{v}_1\|^2} \vec{v}_1 \\ \vec{v}_i &= \vec{w}_i - \frac{\langle \vec{w}_i, \vec{v}_1 \rangle}{\|\vec{v}_1\|^2} \vec{v}_1 - \dots - \frac{\langle \vec{w}_i, \vec{v}_{i-1} \rangle}{\|\vec{v}_{i-1}\|^2} \vec{v}_{i-1}\end{aligned}$$

**Theorem 1.3.2** (QR-Decomposition). I just type this out since it has a name...

**Theorem 1.3.3** (The Fundamental Theorem of Linear Algebra). If  $A$  is an  $m \times n$  matrix, then  $\text{Col}(A)^\perp = \text{Null}(A^T)$  and  $\text{Row}(A)^\perp = \text{Null}(A)$ . In particular,

$$\mathbb{R}^n = \text{Row}(A) \oplus \text{Null}(A) \quad \text{and} \quad \mathbb{R}^m = \text{Col}(A) \oplus \text{Null}(A^T)$$

**Theorem 1.3.4** (Approximation Thm). Let  $\mathbb{W}$  be a finite dimensional subspace of an inner product space  $\mathbb{V}$ . If  $\vec{v} \in \mathbb{V}$ , then the vector closest to  $\vec{v}$  in  $\mathbb{W}$  is  $\text{proj}_{\mathbb{W}}(\vec{v})$ . That is,

$$\|\vec{v} - \text{proj}_{\mathbb{W}}(\vec{v})\| < \|\vec{v} - \vec{w}\|$$

For all  $\vec{w} \in \mathbb{W}, \vec{w} \neq \text{proj}_{\mathbb{W}}(\vec{v})$ .

### Chapter 10

**Theorem 1.4.1** (Triangularization Thm). If  $A$  is an  $n \times n$  matrix with real eigenvalues, then  $A$  is orthogonally similar to an upper triangular matrix  $T$ .

**Theorem 1.4.2** (Principal Axis Thm). Every symmetric matrix  $A$  is orthogonally diagonalizable.

### Chapter 11

**Theorem 1.5.1** (Schur's Thm). If  $A \in M_{n \times n}(\mathbb{C})$ , then  $A$  is unitarily similar to an upper triangular matrix  $T$ . Moreover, the diagonal entries of  $T$  are the eigenvalues of  $A$ .

**Theorem 1.5.2** (Spectral Theorem for Hermitian Matrices). If  $A$  is Hermitian, then  $A$  is unitarily diagonalizable.

**Theorem 1.5.3** (Spectral Theorem for Normal Matrices). A matrix  $A$  is normal if and only if it is unitarily diagonalizable.

**Theorem 1.5.4** (Cayley-Hamilton Thm). If  $A \in M_{n \times n}(\mathbb{C})$ , then  $A$  is a root of its characteristic polynomial  $C(\lambda)$ .