ROB 101 - Computational Linear Algebra Recitation #6

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1 Null Space and Range of a Matrix

For any Matrix $A \in_{m \times n}$, then the following sets (are actually Subspaces!) can be defined:

Def. $\operatorname{null}(A) := \{x \in^m \mid Ax = 0_{n \times 1}\}$ is the **null space** of A.

Def. range $(A) := \{ y \in \mathbb{R} \mid y = Ax \text{ for some } x \in \mathbb{R} \}$ is the **range** of A.

Question? What is a null space of a matrix and why is it important?

Source: FAQ Questions https://umich.instructure.com/courses/475066/files/folder/HW/HW%2006

2 Basis Vectors

Definition: Suppose that V is a subspace of \mathbb{R}^n . Then $\{v_1, v_2, ..., v_k\}$ is a basis for V if

- the set $\{v_1, v_2, ..., v_k\}$ is linearly independent
- span $\{v_1, v_2, ..., v_k\} = \mathbf{V}$

The dimension of V is k, the number of basis vectors.

Remark: Basis vectors provide a simple means to generate all vectors in a vector space or a subspace by forming linear combinations from a finite list of vectors. The three vectors commonly seen in vector calculus and physics $\hat{i}, \hat{j}, \hat{k}$ are orthonormal basis vectors!

3 Span of a Set of Vectors

Definition: The span of a set of vectors $S \in \mathbb{R}^n$ is:

 $\operatorname{span}(S) := \{ \text{all possible linear combinations of elements in S} \}$

The span operation is useful for generating a subspace from an arbitrary set vectors in \mathbb{R}^n by the definition of the span (contains zero vector and is closed under linear combination). That is, the result of span(S) is a subspace of \mathbb{R}^n .

4 Column Span of a Matrix

Definition: The column span of an $n \times m$ matrix A is:

$${\rm col}(A) := {\rm span}(\{a_1^{col}, a_2^{col}, ..., a_m^{col}\})$$

i.e., take the columns of matrix A and form a set S containing m vectors in \mathbb{R}^n ($\{a_1^{col}, a_2^{col}, ..., a_m^{col}\}$) to perform the span(S) operation.

Remark: Ax = b has a solution if, and only if, b is a linear combination of the columns of A. A more elegant way to write this is Ax = b has a solution if, and only if,

$$b \in col(A)$$

We can also discuss, rank and nullity of A here as:

Def. $rank(A) := dim col span{A}.$

Def. $\operatorname{nullity}(A) := \dim \operatorname{null}(A)$.

5 Eigen values and Eigen Vectors

Let A be an $n \times n$ matrix. A scalar $\lambda \in \mathbb{R}$ is an **eigenvalue** (e-value) of A, if there exists a non-zero vector $v \in \mathbb{R}^n$ such that $Av = \lambda v$. Any such vector v is called an **eigenvector** (e-vector) associated with λ .

Eigenvectors are not unique.

• To find e-values, we solve $det(\lambda I - A) = 0$ because

$$A \cdot v = \lambda v \iff (\lambda I - A) \cdot v = 0 \stackrel{v \neq 0}{\iff} \det(\lambda I - A) = 0.$$
 (1)

• To find e-vectors, we find any non-zero $v \in \mathbb{R}^n$ such that

$$(\lambda I - A) \cdot v = 0. \tag{2}$$

Of course, if you prefer, you can solve $(A - \lambda I)v = 0$ when seeking e-vectors.