L-17: Orthogonal Matrices and Gram Schmidt

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1 Orthogonal Matrices

1.1 Orthonormal Vectors

Unit vectors which are prependicular to each other are called orthonormal vectors.

$$q_i^T \cdot q_j = \begin{cases} 0 & if \ i \neq j \\ 1 & if \ i = j \end{cases}$$

The self dot product is 1 as it is a unit vector. So length squared is zero.

1.2 What are orthogonal matrices?

Consider an n-dimensional vector space with n orthogonal basis $q_1, q_2, q_3...q_n$

The orthogonal matrices has these basis vectors in its columns.

$$Q = [q_1 \ q_2 \ \dots \ q_n]$$

$$Q_T.Q = \begin{bmatrix} q_1^T \\ \vdots \\ q_n^T \end{bmatrix} [q_1 \ q_2 \ \dots \ q_n] = I$$

Solve the identity proof yourself.

1.3 Examples of orthogonal matrices

1)
$$Q = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$Q = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

4) Rectangular example.

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

1.4 Some properties

Suppose Q is an orthogonal matrix with orthonormal vectors as its columns. Projection matrix for projecting a vector b onto its column space is given by:

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

$$P = QQ^T$$

Now, suppose Q is an orthogonal matrix. All columns are orthonormal and independent. This means in an n dimensional space, we get n pivots after elimination. So the column space will be the entire n dimensional space.

In this case P=I, as we won't have the need to project a vector as it would be already in the n-dimensional space. We can verify properties like $P^T = P$ and $P^2 = P$.

1.5 Normal Equation

$$\begin{split} A^T A \hat{x} &= A^T b \\ Q^T Q \hat{x} &= Q^T b \end{split}$$

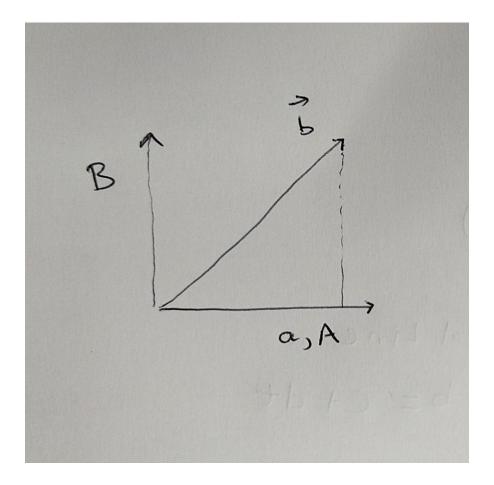
$$\hat{x} = Q^T b$$

 $\hat{x_i} = q_i^T b$ [i=ith component]

2 Gram Schmidt Process

In mathematics, particularly linear algebra and numerical analysis, the Gram–Schmidt process is a method for orthonormalizing a set of vectors in an inner product space, most commonly the Euclidean space \mathbb{R}^n equipped with the standard inner product.

2.1 Two dimensional intuition



We have to make both vectors orthogonal. Lets take a vector as reference. Let's say say vector A is equal to a.

Suppose B is perpendicular to A vector.

B is the error projection of b. Therefore, B=e

B=b-e [parallelogram law]

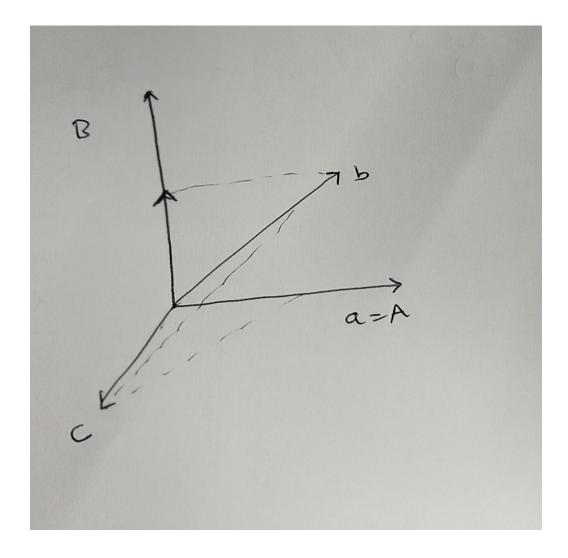
B= $b - \hat{x}a$ [Proved in projection onto subspace lecture]

Or

 $B = b - \hat{x}A$

 $\therefore B = b - \frac{A^Tb}{A^TA}A$ [Gram's formula]

Three dimensional intuition 2.2



Here, $A \perp B \perp C$

Like 2d intuition, A=a and $B=b-\frac{A^Tb}{A^TA}A$

For C, we have to think differently.

Projection of C on A is given by, $p_1 = \frac{A^T c}{A^T A} A$

Projection of C on B is given by, $p_2 = \frac{B^T c}{B^T B} B$ Now, these two vectors are perpendicular as A and B are perpendicular. They will become basis for a plane. A vector line on the plane would be linear combination of the above mentioned vectors, which is:

$$p_p lane = p1 + p2$$

Now, this is equivalent to projection of c on that plane. The perpendicular vector is given by:

$$C = c - \left(\frac{A^T c}{A^T A}A\right) - \left(\frac{B^T c}{B^T B}B\right)$$

3 Extra

3.1 Numerical Example

$$a = \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix}, b = \begin{bmatrix} 1\\0\\2 \end{bmatrix}, S = \begin{bmatrix} 1&1\\1&0\\1&2 \end{bmatrix}$$

$$B = \begin{bmatrix} 1\\0\\2 \end{bmatrix} - \frac{3}{3} \begin{bmatrix} 1\\1\\1 \end{bmatrix}$$

$$B = \begin{bmatrix} 0\\-1\\1 \end{bmatrix}$$

$$Q = \begin{bmatrix} q1, q2 \end{bmatrix}$$

$$Q = \begin{bmatrix} \frac{1}{\sqrt{3}} & 0\\ \frac{1}{\sqrt{3}} & \frac{-1}{\sqrt{2}}\\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} \end{bmatrix}$$

3.2 Relation between C(Q) and C(S)

The column space of both matrices is the same. A and B are perpendicular. They form a 2d column space. It's just that Q matrix is more refined than the above mentioned vectors, as it contains unit vectors in the directions of A and B. Thus, it has basis for C(S).