CX1104: Linear Algebra for Computing

$$\underbrace{\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix}}_{A} \underbrace{\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix}}_{m \times n} = \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}}_{m \times 1}$$

Chap. No: **6.2.4**

Lecture: Orthogonality

Topic: Orthogonality

Concept: Orthonormal Sets & Orthogonal Matrices

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Orthonormal Set and Orthonormal basis

Orthonormal Sets

A set $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ is an **orthonormal set** if it is an orthogonal set of unit vectors. If W is the subspace spanned by such a set, then $\{\mathbf{u}_1,\ldots,\mathbf{u}_p\}$ is an **orthonormal basis** for W, since the set is automatically linearly independent, by Theorem 4.

The simplest example of an orthonormal set is the standard basis $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ for \mathbb{R}^n . Any nonempty subset of $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ is orthonormal, too. Here is a more complicated example.

THEOREM 4

If $S = \{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ is an orthogonal set of nonzero vectors in \mathbb{R}^n , then S is linearly independent and hence is a basis for the subspace spanned by S.

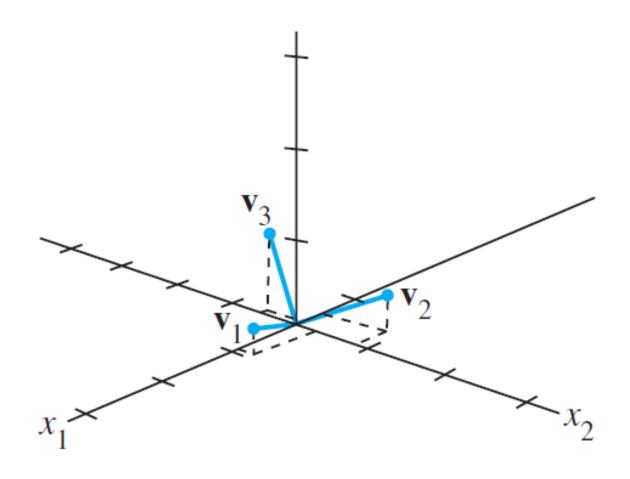


FIGURE 6

Ref: https://www.youtube.com/watch?v=ZJu26chXEiw

Lay's Linear Algebra and Applications

CHAPTER 6 Orthogonality and Least Squares

Linear Algebra: Orthonormal Basis 61,234 views • Jun 28, 2014 **Worldwide Center of Mathematics** 26.5K subscribers

EXAMPLE 5 Show that $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ is an orthonormal basis of \mathbb{R}^3 , where

$$\mathbf{v}_{1} = \begin{bmatrix} 3/\sqrt{11} \\ 1/\sqrt{11} \\ 1/\sqrt{11} \end{bmatrix}, \quad \mathbf{v}_{2} = \begin{bmatrix} -1/\sqrt{6} \\ 2/\sqrt{6} \\ 1/\sqrt{6} \end{bmatrix}, \quad \mathbf{v}_{3} = \begin{bmatrix} -1/\sqrt{66} \\ -4/\sqrt{66} \\ 7/\sqrt{66} \end{bmatrix}$$

SOLUTION Compute

$$\mathbf{v}_1 \cdot \mathbf{v}_2 = -3/\sqrt{66} + 2/\sqrt{66} + 1/\sqrt{66} = 0$$

$$\mathbf{v}_1 \cdot \mathbf{v}_3 = -3/\sqrt{726} - 4/\sqrt{726} + 7/\sqrt{726} = 0$$

$$\mathbf{v}_2 \cdot \mathbf{v}_3 = 1/\sqrt{396} - 8/\sqrt{396} + 7/\sqrt{396} = 0$$

Thus $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ is an orthogonal set. Also,

$$\mathbf{v}_1 \cdot \mathbf{v}_1 = 9/11 + 1/11 + 1/11 = 1$$

 $\mathbf{v}_2 \cdot \mathbf{v}_2 = 1/6 + 4/6 + 1/6 = 1$
 $\mathbf{v}_3 \cdot \mathbf{v}_3 = 1/66 + 16/66 + 49/66 = 1$

which shows that \mathbf{v}_1 , \mathbf{v}_2 , and \mathbf{v}_3 are unit vectors. Thus $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ is an orthonormal set. Since the set is linearly independent, its three vectors form a basis for \mathbb{R}^3 . See Fig. 6.

NOTE: The 0's correspond to dot products of orthogonal vectors. See next slide for explanation of result!

Orthonormal Sets and U^TU

THEOREM 6

An $m \times n$ matrix U has orthonormal columns if and only if $U^TU = I$.

PROOF To simplify notation, we suppose that U has only three columns, each a vector in \mathbb{R}^m . The proof of the general case is essentially the same. Let $U = [\mathbf{u}_1 \ \mathbf{u}_2 \ \mathbf{u}_3]$ and compute

$$U^{T}U = \begin{bmatrix} \mathbf{u}_{1}^{T} \\ \mathbf{u}_{2}^{T} \\ \mathbf{u}_{3}^{T} \end{bmatrix} \begin{bmatrix} \mathbf{u}_{1} & \mathbf{u}_{2} & \mathbf{u}_{3} \end{bmatrix} = \begin{bmatrix} \mathbf{u}_{1}^{T}\mathbf{u}_{1} & \mathbf{u}_{1}^{T}\mathbf{u}_{2} & \mathbf{u}_{1}^{T}\mathbf{u}_{3} \\ \mathbf{u}_{2}^{T}\mathbf{u}_{1} & \mathbf{u}_{2}^{T}\mathbf{u}_{2} & \mathbf{u}_{2}^{T}\mathbf{u}_{3} \\ \mathbf{u}_{3}^{T}\mathbf{u}_{1} & \mathbf{u}_{3}^{T}\mathbf{u}_{2} & \mathbf{u}_{3}^{T}\mathbf{u}_{3} \end{bmatrix}$$
(4)

The entries in the matrix at the right are inner products, using transpose notation. The columns of U are orthogonal if and only if

$$\mathbf{u}_{1}^{T}\mathbf{u}_{2} = \mathbf{u}_{2}^{T}\mathbf{u}_{1} = 0, \quad \mathbf{u}_{1}^{T}\mathbf{u}_{3} = \mathbf{u}_{3}^{T}\mathbf{u}_{1} = 0, \quad \mathbf{u}_{2}^{T}\mathbf{u}_{3} = \mathbf{u}_{3}^{T}\mathbf{u}_{2} = 0$$
 (5)

The columns of U all have unit length if and only if

$$\mathbf{u}_1^T \mathbf{u}_1 = 1, \quad \mathbf{u}_2^T \mathbf{u}_2 = 1, \quad \mathbf{u}_3^T \mathbf{u}_3 = 1 \tag{6}$$

The theorem follows immediately from (4)–(6).

Matlab Example: When *U* is square (3x3)

```
>> U = [3/sqrt(11) -1/sqrt(6) -1/sqrt(66);
1/sqrt(11) 2/sqrt(6) -4/sqrt(66);
1/sqrt(11) 1/sqrt(6) 7/sqrt(66)]
U =
                       -0.1231
                       -0.4924
    0.3015
              0.4082
                       0.8616
>> U'*U
                        0.0000
    0.0000
              0.0000
                        1.0000
>> U*U'
                        0.0000
                        0.0000
              0.0000
                        1.0000
```

vs U is rectangle (3x2)

```
>> U2=[U(:,1) U(:,2)]
U2 =
                                   >> P=U2*U2'
             -0.4082
              0.8165
    0.3015
              0.4082
                                                -0.0606
                                                           0.1061
>> U2'*U2
                                                           0.4242
                                                 0.4242
                                                           0.2576
                                      0.1061
ans =
                                   >> P*P
    1.0000
              0.0000
    0.0000
              1.0000
                                   ans =
>> U2*U2'
                                                -0.0606
                                                           0.1061
                                      -0.0606
                                                 0.7576
                                                           0.4242
ans =
                                       0.1061
                                                 0.4242
                                                           0.2576
                         0.1061
             -0.0606
                                      >> rank(P*P)
                         0.4242
              0.7576
                        0.2576
    0.1061
              0.4242
                                            2
```

Question: If U (a mx2 matrix) ONLY has 2 **orthonormal** columns, (m>2), what is the characteristics of matrixes: U^TU and UU^T ?

```
Ans: U^TU = identity (2x2) matrix P = UU^T = (mxm)matrix but not identity (only rank 2) It spans the column space of U, lets call it W. (later we show) it is a projection matrix P P can be used to projecting a given vector Y onto W by \hat{y} = UU^T y = Py (see Ch 6.2.5, pg 7 Theorem 10)
```

Orthonormal Sets and Orthogonal matrix

THEOREM 7

Let U be an $m \times n$ matrix with orthonormal columns, and let \mathbf{x} and \mathbf{y} be in \mathbb{R}^n . Then

a.
$$||U\mathbf{x}|| = ||\mathbf{x}||$$

b.
$$(U\mathbf{x}) \cdot (U\mathbf{y}) = \mathbf{x} \cdot \mathbf{y}$$

c. $(U\mathbf{x}) \cdot (U\mathbf{y}) = 0$ if and only if $\mathbf{x} \cdot \mathbf{y} = 0$

Properties (a) and (c) say that the linear mapping $\mathbf{x} \mapsto U\mathbf{x}$ preserves lengths and orthogonality. These properties are crucial for many computer algorithms.

EXAMPLE 6 Let
$$U = \begin{bmatrix} 1/\sqrt{2} & 2/3 \\ 1/\sqrt{2} & -2/3 \\ 0 & 1/3 \end{bmatrix}$$
 and $\mathbf{x} = \begin{bmatrix} \sqrt{2} \\ 3 \end{bmatrix}$. Notice that U has or-

thonormal columns and

$$U^{T}U = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} & 0\\ 2/3 & -2/3 & 1/3 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 2/3\\ 1/\sqrt{2} & -2/3\\ 0 & 1/3 \end{bmatrix} = \begin{bmatrix} 1 & 0\\ 0 & 1 \end{bmatrix}$$

Verify that $||U\mathbf{x}|| = ||\mathbf{x}||$.

SOLUTION

$$U\mathbf{x} = \begin{bmatrix} 1/\sqrt{2} & 2/3 \\ 1/\sqrt{2} & -2/3 \\ 0 & 1/3 \end{bmatrix} \begin{bmatrix} \sqrt{2} \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ -1 \\ 1 \end{bmatrix}$$
$$\|U\mathbf{x}\| = \sqrt{9+1+1} = \sqrt{11}$$
$$\|\mathbf{x}\| = \sqrt{2+9} = \sqrt{11}$$

Theorems 6 and 7 are particularly useful when applied to *square* matrices. An **orthogonal matrix** is a square invertible matrix U such that $U^{-1} = U^T$. By Theorem 6, such a matrix has orthonormal columns.¹ It is easy to see that any *square* matrix with orthonormal columns is an orthogonal matrix. Surprisingly, such a matrix must have orthonormal *rows*, too.

Orthogonal Matrix definitions

Ref: https://en.wikipedia.org/wiki/Orthogonal_matrix Important!

In linear algebra, an **orthogonal matrix**, or **orthonormal matrix**, is a real square matrix whose columns and rows are orthonormal vectors.

One way to express this is

$$Q^{\mathrm{T}}Q = QQ^{\mathrm{T}} = I,$$

where Q^{T} is the transpose of Q and I is the identity matrix.

This leads to the equivalent characterization: a matrix Q is orthogonal if its transpose is equal to its inverse:

$$Q^{\mathrm{T}}=Q^{-1},$$

where Q^{-1} is the inverse of Q.

Orthogonal Matrix MUST be SQUARE!

Note: confusion

If you have a mxn matrix called *U*with its column ortho-normal,
and m>n (tall matrix)

- 1) IT IS NOT an orthogonal matrix since it satisfy ONLY $U^TU = I$ (nxn)
- 2) its UU^T is mxm matrix BUT it is not equals to I) .Instead UU^T is a projection matrix and has rank n.

Note: There is no standard name for "rectangular matrix with orthonormal columns"