

National University of Singapore
MA2001 Linear Algebra
MATLAB Worksheet 5
Dot Product, Orthogonal Sets and Least Squares Solutions

A. Dot Product and Norm

Let $\mathbf{u} = (u_1, \dots, u_n)$ and $\mathbf{v} = (v_1, \dots, v_n)$ be vectors in \mathbb{R}^n . Their **dot product** is defined by

$$\mathbf{u} \cdot \mathbf{v} = u_1v_1 + \dots + u_nv_n = \begin{pmatrix} u_1 & \dots & u_n \end{pmatrix} \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}.$$

If both \mathbf{u} and \mathbf{v} are defined as row vectors, then $\mathbf{u} \cdot \mathbf{v} = \mathbf{u}\mathbf{v}^T$.

If both \mathbf{u} and \mathbf{v} are defined as column vectors, then $\mathbf{u} \cdot \mathbf{v} = \mathbf{u}^T\mathbf{v}$.

For example, let $\mathbf{u} = (1, 2, 3, 4, 5)$ and $\mathbf{v} = (1, 0, 1, -1, 2)$.

```
>> u = [1 2 3 4 5]; v = [1 0 1 -1 2];  
>> u * v'  
ans = 10
```

Alternatively, MATLAB provides a command `dot` for dot product, regardless whether the vectors are defined as row or column vectors.

```
>> dot(u, v)  
ans = 10  
>> dot(u', v)  
ans = 10  
>> dot(u, v')  
ans = 10  
>> dot(u', v')  
ans = 10
```

The **norm** of a vector $\mathbf{v} = (v_1, \dots, v_n)$ in \mathbb{R}^n is defined by

$$\|\mathbf{v}\| = \sqrt{v_1^2 + \dots + v_n^2} = \sqrt{\mathbf{v} \cdot \mathbf{v}}.$$

Using the vectors \mathbf{u} and \mathbf{v} above, their norms can be evaluated using the `sqrt` and `dot` commands. We shall use the default format in MATLAB.

```
>> format short  
>> sqrt(dot(u, u))  
ans = 7.4162  
>> sqrt(dot(v, v))  
ans = 2.6458
```

Alternatively, in MATLAB `norm` can be used to find the norm of a vector.

```
>> norm(u)  
ans = 7.4162
```

Note that the norm of a vector is usually irrational (because of the square root), and the output is in floating-point. We can use `sym` to define a vector as **symbolic** object, and use `norm` to get the exact value of the norm. For example,

```
>> u = sym([1 2 3 4 5])
u = [1, 2, 3, 4, 5]
>> norm(u)
ans = 55^(1/2)
```

B. Orthogonal Sets

A set of vectors $S = \{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ in \mathbb{R}^n is said to be an **orthogonal** set if

$$\mathbf{v}_i \cdot \mathbf{v}_j = 0 \quad \text{for all } i \neq j.$$

View each vector as a row vector and consider $\mathbf{A} = \begin{pmatrix} \mathbf{v}_1 \\ \vdots \\ \mathbf{v}_k \end{pmatrix}$. Then $\mathbf{A}^T = (\mathbf{v}_1^T \ \cdots \ \mathbf{v}_k^T)$

and

$$\mathbf{A}\mathbf{A}^T = \begin{pmatrix} \mathbf{v}_1\mathbf{v}_1^T & \cdots & \mathbf{v}_1\mathbf{v}_k^T \\ \vdots & \ddots & \vdots \\ \mathbf{v}_k\mathbf{v}_1^T & \cdots & \mathbf{v}_k\mathbf{v}_k^T \end{pmatrix} = \begin{pmatrix} \mathbf{v}_1 \cdot \mathbf{v}_1 & \cdots & \mathbf{v}_1 \cdot \mathbf{v}_k \\ \vdots & \ddots & \vdots \\ \mathbf{v}_k \cdot \mathbf{v}_1 & \cdots & \mathbf{v}_k \cdot \mathbf{v}_k \end{pmatrix}.$$

Hence, to determine whether the set $S = \{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ is orthogonal, we can use **MATLAB** to check whether $\mathbf{A}\mathbf{A}^T$ is a diagonal matrix.

For example, consider a set of vectors

$$\{(1, 1, 1, 1), (1, 0, -1, 0), (1, -1, 1, -1)\}$$

in \mathbb{R}^4 .

- (i) Define the matrix \mathbf{C} whose rows are the given vectors.

```
>> C = [1 1 1 1; 1 0 -1 0; 1 -1 1 -1];
```

- (ii) Evaluate $\mathbf{C}\mathbf{C}^T$.

```
>> C * C'
ans = 4      0      0
      0      2      0
      0      0      4
```

Since $\mathbf{C}\mathbf{C}^T$ is a diagonal matrix, the set of vectors $\{(1, 1, 1, 1), (1, 0, -1, 0), (1, -1, 1, -1)\}$ is an orthogonal set.

A set of vectors $S = \{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ in \mathbb{R}^n is said to be an **orthonormal** set if

$$\mathbf{v}_i \cdot \mathbf{v}_j = \begin{cases} 1 & \text{if } i = j, \\ 0 & \text{if } i \neq j. \end{cases}$$

View each vector as a row vector and consider $\mathbf{A} = \begin{pmatrix} \mathbf{v}_1 \\ \vdots \\ \mathbf{v}_k \end{pmatrix}$.

Hence, to determine whether the set $S = \{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ is orthonormal, we can use MATLAB to check whether $\mathbf{A}\mathbf{A}^T = \mathbf{I}_k$, the identity matrix of order k .

For example, consider a set of vectors

$$\left\{ \left(\cos \frac{\pi}{3}, 0, \sin \frac{\pi}{3}, 0 \right), (0, 1, 0, 0), \left(\sin \frac{\pi}{3}, 0, -\cos \frac{\pi}{3}, 0 \right) \right\}$$

in \mathbb{R}^4 .

(i) Define the matrix \mathbf{D} whose rows are the given vectors.

```
>> D = [cos(pi/3) 0 sin(pi/3) 0; 0 1 0 0; sin(pi/3) 0 -cos(pi/3) 0];
```

(ii) Evaluate $\mathbf{D}\mathbf{D}^T$.

```
>> D * D'
ans = 1.0000      0      0.0000
      0      1.0000      0
      0.0000      0      1.0000
```

Since $\mathbf{D}\mathbf{D}^T = \mathbf{I}_3$, the given set of vectors $\left\{ \left(\cos \frac{\pi}{3}, 0, \sin \frac{\pi}{3}, 0 \right), (0, 1, 0, 0), \left(\sin \frac{\pi}{3}, 0, -\cos \frac{\pi}{3}, 0 \right) \right\}$ is an orthonormal set.

C. Orthonormal Bases

If $V = \text{span}(S)$, and S is an orthonormal set, then S is linearly independent, and S is called an **orthonormal basis** for V .

In MATLAB, `orth` can be used to get an **orthonormal basis** for the column space of a matrix.

Suppose $V = \text{span}(S)$, where $S = \{(1, 1, 1, 1), (1, 1, 0, 0), (0, 1, 1, 0)\}$.

(i) Define matrix $\mathbf{E} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}$ whose columns are the vectors in the spanning set S . By definition, $V = \text{span}(S)$ is the column space of the matrix \mathbf{E} .

```
>> E = [1 1 0; 1 1 1; 1 0 1; 1 0 0];
```

It might be easier to input in MATLAB the vectors of S as rows of a matrix and then take the transpose, as shown below.

```
>> E = [1 1 1 1; 1 1 0 0; 0 1 1 0]';
```

- (ii) Use the command `orth` to get an orthonormal basis for the column space of \mathbf{E} , i.e., for V .

```
>> orth(E)
ans =  -0.4835    0.7071   -0.1273
        -0.6635    0.0000    0.5565
        -0.4835   -0.7071   -0.1273
        -0.3035   -0.0000   -0.8111
```

The columns of the resulting matrix give an orthonormal basis for the vector space V :

$\{(-0.4835, -0.6635, -0.4835, -0.3035), (0.7071, 0, -0.7071, 0), (-0.1273, 0.5565, -0.1273, -0.8111)\}$

Important note:

The (classical) Gram-Schmidt process in the textbook (Theorem 5.2.19) is numerically unstable. The MATLAB command `orth` above uses a **modified Gram-Schmidt process**¹ to generate an orthonormal basis which is generally different from the classical process.

In order to get the exact form of the orthonormal basis, we can use `sym` to define the matrix as **symbolic** object. By dealing with symbolic objects, there is no longer numerical error. In this case, `orth` uses the classical Gram-Schmidt process to generate the orthonormal basis.

Again, suppose $V = \text{span}(S)$, where $S = \{(1, 1, 1, 1), (1, 1, 0, 0), (0, 1, 1, 0)\}$.

- (i) Define matrix \mathbf{E} whose columns are the vectors in S as an symbolic object.

```
>> E = sym([1 1 0; 1 1 1; 1 0 1; 1 0 0]);
E =  [1, 1, 0]
      [1, 1, 1]
      [1, 0, 1]
      [1, 0, 0]
```

- (ii) Use `orth` to get an orthonormal basis for the column space of \mathbf{E} , i.e., for V .

```
>> orth(E)
ans =  [1/2, 1/2, -1/2]
        [1/2, 1/2, 1/2]
        [1/2, -1/2, 1/2]
        [1/2, -1/2, -1/2]
```

¹Your may refer to the session **Numerical Stability** on Gram-Schmidt process in WIKIPEDIA: https://en.wikipedia.org/wiki/Gram-Schmidt_process.

The columns of the resulting matrix give an orthonormal basis for the vector space V .

$$\left\{ \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2} \right), \left(\frac{1}{2}, \frac{1}{2}, -\frac{1}{2}, -\frac{1}{2} \right), \left(-\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, -\frac{1}{2} \right) \right\}$$

The command `orth` also allows us to generate an orthogonal basis (i.e. the vectors need not be of norm 1) in symbolic form by “skipping the normalization” in Gram-Schmidt process:

```
>> orth(E, 'skipnormalization')
ans = [1, 1/2, -1/2]
      [1, 1/2, 1/2]
      [1, -1/2, 1/2]
      [1, -1/2, -1/2]
```

D. Least Squares Solutions

When a linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ is inconsistent, we can find its **least squares solutions**, which are the “best approximation” in place of an exact solution for the system.

Recall that the least squares solution can be found by solving the following linear system

$$\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}.$$

For example, consider the linear system

$$\begin{cases} x + 2y + z = 1 \\ x + 2y + 2z = 1 \\ 2x + 4y + z = 3 \end{cases}$$

We input the system to MATLAB as an augmented matrix `[A b]`:

```
>> A = [1 2 1; 1 2 2; 2 4 1];
>> b = [1; 1; 3];
>> rref[A b];
ans = 1 2 0 0
      0 0 1 0
      0 0 0 1
```

The RREF shows the system is inconsistent.

We proceed to solve the system $\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}$:

```
>> format rat
>> rref[A'*A A'*b];
ans = 1 2 0 18/11
      0 0 1 -4/11
      0 0 0 0
```

This system is consistent and has infinitely many solutions.

Observe that the 2nd column of the reduced row-echelon form is a pivot column. Set $x_2 = s$ to be arbitrary parameter, and solve the other variables:

$$x_1 = \frac{18}{11} - 2s, \quad x_3 = \frac{-4}{11}.$$

This gives us the least squares solutions for the original system $\mathbf{A}\mathbf{x} = \mathbf{b}$:

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} \frac{18}{11} - 2s \\ s \\ \frac{-4}{11} \end{pmatrix}.$$

E. Projections

In this section, we show how to use least squares solutions to find the **projection** \mathbf{p} of a given vector \mathbf{v} onto a vector space V . Recall that \mathbf{p} is defined to be the unique vector such that $\mathbf{v} - \mathbf{p}$ is orthogonal to every vector in V .

Let's take $V = \text{span}\{(1, 1, 2), (2, 2, 4), (1, 2, 1)\}$ and $\mathbf{v} = (1, 1, 3)$. To find the projection \mathbf{p} of \mathbf{v} onto V , we use the spanning vectors (written in column form) of V to form the matrix $\mathbf{A} = \begin{pmatrix} 1 & 2 & 1 \\ 1 & 2 & 2 \\ 2 & 4 & 1 \end{pmatrix}$, and let $\mathbf{b} = \begin{pmatrix} 1 \\ 1 \\ 3 \end{pmatrix}$ which is the vector \mathbf{v} in column form. Note that the column space of \mathbf{A} is precisely V .

Recall that the projection of a vector \mathbf{b} onto the column space of a matrix \mathbf{A} is given by $\mathbf{A}\mathbf{x}_0$ where \mathbf{x}_0 is a least squares solution of the system $\mathbf{A}\mathbf{x} = \mathbf{b}$.

Note that our \mathbf{A} and \mathbf{b} are the examples from the previous section, and we have found the least squares solutions of the system $\mathbf{A}\mathbf{x} = \mathbf{b}$, given by

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} \frac{18}{11} - 2s \\ s \\ \frac{-4}{11} \end{pmatrix}.$$

Take \mathbf{x}_0 to be any particular least squares solution (say we let $s = 0$):

$$\mathbf{x}_0 = \begin{pmatrix} \frac{18}{11} \\ 0 \\ \frac{-4}{11} \end{pmatrix}.$$

Input this solution in MATLAB:

```
>> x0=[18/11 0 -4/11];
```

Then the required projection \mathbf{p} can be computed as $\mathbf{A}\mathbf{x}_0$.

```
>> p=A*x0
p = 14/11
    10/11
    32/11
```

F. Practices

Use MATLAB to solve Questions 5.1, 5.2, 5.6, 5.10, 5.11(a), 5.12, 5.13, 5.14, 5.15, 5.17, 5.22, 5.24, 5.25, 5.26, 5.27(a) in the textbook Exercise 5.