## **Recitation** 10

## The matrix of a linear transformation

**Definition 0.1.** A vector space is a set V together with operations  $+: V \times V \to V$  and  $\cdot: \mathbb{R} \times V \to V$  called "addition" and "scalar multiplication," respectively, satisfying various axioms, which should be intuitive by now.

**Definition 0.2.** We say that vectors  $v_1, \ldots, v_n \in V$  are *linearly dependent* if there exists  $c_1, \ldots, c_n \in \mathbb{R}$  not all zero, with  $c_1v_1 + \ldots + c_nv_n = 0$ . If  $v_1, \ldots, v_n$  are not linearly dependent then we say they are *linearly independent*.

**Definition 0.3.**  $v_1, \ldots, v_n \in V$  are said to be a *basis* for V if  $v_1, \ldots, v_n$  are linearly independent and any  $v \in V$  can be written as a linear combination of  $v_1, \ldots, v_n$ : that is  $v = c_1v_1 + \ldots + c_nv_n$ for come  $c_1, \ldots, c_n \in \mathbb{R}$ .

**Definition 0.4.** If V and W are vector spaces, a *linear transformation*  $T: V \longrightarrow W$  is a function such that T(v + cw) = T(v) + cT(w) for any  $v, w \in V$  and  $c \in \mathbb{R}$ .

**Definition 0.5.** Suppose V and W are vector spaces with bases  $v_1, \ldots, v_n$  and  $w_1, \ldots, w_m$ , respectively. Then for each  $j \in \{1, \ldots, n\}, T(v_j)$  is a linear combination of the  $w_1, \ldots, w_m$ . So there exist

 $a_{1,j}, \ldots, a_{m,j}$  with  $Tv_j = a_{1,j}w_1 + a_{2,j}w_2 + \ldots + a_{m,j}w_m$ .

The matrix of T with respect to the bases  $v_1, \ldots, v_n$  and  $w_1, \ldots, w_m$  is the  $m \times n$  matrix A, with entries  $a_{i,j}$ .

### The most basic example of the matrix of a linear transformation

- 1.  $\mathbb{R}^n$  is a vector space:  $(x_1, \dots, x_n) + (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n)$  and  $c(x_1, \dots, x_n) = (cx_1, \dots, cx_n)$ .
- 2. Let  $e_i \in \mathbb{R}^n$  be the vector with 1 in the *i*<sup>th</sup> entry and 0 in the other entries. Then  $e_1, \ldots, e_n$  are linearly independent in  $\mathbb{R}^n$ .
- 3.  $e_1, \ldots, e_n$  is a basis for  $\mathbb{R}^n$ .
- 4. An  $m \times n$  matrix A defines a linear transformation  $T : \mathbb{R}^n \longrightarrow \mathbb{R}^m$  by T(x) = Ax.
- 5.  $\mathbb{R}^n$  has basis  $e_1, \ldots, e_n$  and  $\mathbb{R}^m$  has basis  $e'_1, \ldots, e'_m$ , where we use the prime to highlight the vectors are in  $\mathbb{R}^m$  as opposed to  $\mathbb{R}^n$ , but they have an identical definition. We have

$$Ae_j = a_{1,j}e'_1 + a_{2,j}e'_2 + \ldots + a_{m,j}e'_m$$

and so the matrix of the linear transformation T(x) = Ax with respect to the bases,  $e_1, \ldots, e_n$ and  $e'_1 \ldots, e'_m$  is A.

## Changing basis

**Definition 0.6.** If V is a vector space and  $v_1, \ldots, v_n$  and  $v'_1, \ldots, v'_n$  are two bases for V. Then there exists an invertible  $n \times n$  matrix B with entries  $b_{i,j}$  such that

$$v'_{j} = b_{1,j}v_1 + b_{2,j}v_2 + \ldots + b_{n,j}v_n$$

B is said to be the *basis change matrix* from  $v_1, \ldots, v_n$  to  $v'_1, \ldots, v'_n$ .

Suppose V and W are vector spaces. Suppose that V has bases  $v_1, \ldots, v_n$  and  $v'_1, \ldots, v'_n$ , that W has bases  $w_1, \ldots, w_m$  and  $w'_1, \ldots, w'_m$ , and that  $T: V \longrightarrow W$  is a linear transformation.

If T has matrix A with respect to the bases  $v_1, \ldots, v_n$  and  $w_1, \ldots, w_m$ , then T has matrix  $C^{-1}AB$  with respect to the bases  $v'_1, \ldots, v'_n$  and  $w'_1, \ldots, w'_m$ , where B and C denote the basis change matrices from the unprimed bases to the primed bases of V and W, respectively.

#### The most basic example of changing basis

Suppose A is an  $m \times n$  matrix. Then we have seen that the linear transformation T(v) = Av has matrix A with respect to the standard bases  $e_1, \ldots, e_n$  and  $e'_1, \ldots, e'_m$ . The matrix of T with respect to another pair of bases  $v_1, \ldots, v_n$  and  $w_1, \ldots, w_m$  is given by  $C^{-1}AB$  where

$$B = \left(v_1 \middle| \cdots \middle| v_n\right)$$
 and  $C = \left(w_1 \middle| \cdots \middle| w_m\right)$ .

#### SVD

Here is an algorithm that will always work for the SVD. Suppose you are given an  $m \times n$  matrix A.

- 1. Let  $\lambda_1, \ldots, \lambda_i$  be the non-zero eigenvalues of  $A^T A$  and  $\lambda_{i+1}, \ldots, \lambda_n$  be the zero eigenvalues of  $A^T A$ . Choose corresponding ORTHONORMAL eigenvectors  $v_1, \ldots, v_n$  for  $A^T A$ .
- 2. Let  $\sigma_j = \sqrt{\lambda_j}$ . Let  $u_1, \ldots, u_i$  be given by  $u_j = Av_j/\sigma_j$ .
- 3. Let  $u_{i+1}, \ldots, u_m$  be an ORTHONORMAL basis of  $N(AA^T)$ .
- 4. The SVD is

$$\left(u_1\bigg|\cdots\bigg|u_m\right)\Sigma\left(v_1\bigg|\cdots\bigg|v_n\right)^T$$

where  $\Sigma$  is an  $m \times n$  matrix with (j, j)-entry given by  $\sigma_j$  and all other entries 0.

If m < n it is a little quicker to do the following.

- 1. Let  $\lambda_1, \ldots, \lambda_i$  be the non-zero eigenvalues of  $AA^T$  and  $\lambda_{i+1}, \ldots, \lambda_m$  be the zero eigenvalues of  $AA^T$ . Choose corresponding ORTHONORMAL eigenvectors  $u_1, \ldots, u_m$  for  $AA^T$ .
- 2. Let  $\sigma_j = \sqrt{\lambda_j}$ . Let  $v_1, \ldots, v_i$  be given by  $v_j = A^T u_j / \sigma_j$ .
- 3. Let  $v_{i+1}, \ldots, v_n$  be an ORTHONORMAL basis of  $N(A^T A)$ .

## **Recitation** 10 questions

## Question 1

Let  $A = \begin{pmatrix} 2 & 3 & 1 \\ 1 & 1 & 5 \end{pmatrix}$ .

- (a) Compute  $A(1,1,1)^T$ ,  $A(1,1,0)^T$ ,  $A(1,-1,0)^T$ . Write each as a linear combination of the vectors  $(1,1)^T$  and  $(0,1)^T$ .
- (b) Find the matrix of the linear transformation T(x) = Ax with respect to the bases  $(1, 1, 1)^T$ ,  $(1, 1, 0)^T$ ,  $(1, -1, 0)^T$  and  $(1, 1)^T$ ,  $(0, 1)^T$  by using the definition of the matrix of a linear transformation.
- (c) Find the matrix of the linear transformation T(x) = Ax with respect to the bases  $(1,1,1)^T$ ,  $(1,1,0)^T$ ,  $(1,-1,0)^T$  and  $(1,1)^T$ ,  $(0,1)^T$  by using basis change matrices.

## Question 2

Suppose A is  $3 \times 2$  matrix with the property that

 $A(4,5)^T = (1,2,3)^T$  and  $A(3,4)^T = (3,2,1)^T$ .

(a) What is the matrix of T(x) = Ax with respect to the bases

 $(4,5)^T$ ,  $(3,4)^T$  and  $(1,2,3)^T$ ,  $(3,2,1)^T$ ,  $(0,1,0)^T$ .

- (b) Use the basis change formula to write the matrix you just calculated as  $C^{-1}AB$ .
- (c) What is A?

## Question 3

Let V be the set of cubic polynomials  $V = \{f(x) = ax^3 + bx^2 + cx + d : a, b, c, d \in \mathbb{R}\}.$ 

- (a) Recall why V is a vector space; what happens to the coefficients under addition and scalar multiplication?
- (b) Define  $T_1(f(x)) = f'(x)$ . Is this a linear transformation  $V \longrightarrow V$ ? Why?
- (c) Define  $T_2(f(x)) = f(x+1)$ . Is this a linear transformation  $V \longrightarrow V$ ? Why?
- (d) Define  $T_3(f(x)) = x^3 f(1/x)$ . Is this a linear transformation  $V \longrightarrow V$ ? Why?
- (e) Recall that  $1, x, x^2, x^3$  is a basis for V. Does this make sense to you?
- (f) What is  $T_1(1), T_1(x), T_1(x^2), T_1(x^3)$ ? What is the matrix of  $T_1$  with respect to the basis  $1, x, x^2, x^3$  (using this as a basis for the domain and codomain)?
- (g) What is the matrix of  $T_2$  with respect to the basis  $1, x, x^2, x^3$ ?
- (h) What is the matrix of  $T_3$  with respect to the basis  $1, x, x^2, x^3$ ?
- (i) What is the matrix of  $T_1$  with respect to the basis 1, x 1, (x 1)(x 2),  $x(x^2 \frac{9}{2}x + 6)$ ?

# Question 4

Let  $A = \begin{pmatrix} 1 & 2 \\ 3 & 6 \end{pmatrix}$ .

- 1. Find the eigenvalues  $\lambda_1 \neq 0$ ,  $\lambda_2$  and unit eigenvectors  $v_1$ ,  $v_2$  of  $A^T A$ .
- 2. Let  $\sigma_1 = \sqrt{\lambda_1}$ ,  $u_1 = Av_1/\sigma_1$ . Verify that  $u_1$  is a unit eigenvector for  $AA^T$  with eigenvalue  $\lambda_1$ .
- 3. Extend  $u_1$  to an orthonormal basis  $u_1, u_2$ .
- 4. Check that

$$A = (u_1|u_2) \begin{pmatrix} \sigma_1 & 0\\ 0 & 0 \end{pmatrix} (v_1|v_2)^T.$$