## Recitation 10

## The matrix of a linear transformation

Definition 0.1. A vector space is a set $V$ together with operations $+: V \times V \rightarrow V$ and $\cdot: \mathbb{R} \times V \rightarrow V$ 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 $c_{1} v_{1}+\ldots+c_{n} v_{n}=0$. 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 $v=c_{1} v_{1}+\ldots+c_{n} v_{n}$ 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 $T(v+c w)=T(v)+c T(w)$ 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 $j \in\{1, \ldots, n\}, T\left(v_{j}\right)$ is a linear combination of the $w_{1}, \ldots, w_{m}$. So there exist

$$
a_{1, j}, \ldots, a_{m, j} \text { with } T v_{j}=a_{1, j} w_{1}+a_{2, j} w_{2}+\ldots+a_{m, j} w_{m}
$$

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:

$$
\left(x_{1}, \ldots, x_{n}\right)+\left(y_{1}, \ldots, y_{n}\right)=\left(x_{1}+y_{1}, \ldots, x_{n}+y_{n}\right) \text { and } c\left(x_{1}, \ldots, x_{n}\right)=\left(c x_{1}, \ldots, c x_{n}\right) .
$$

2. Let $e_{i} \in \mathbb{R}^{n}$ be the vector with 1 in the $i^{\text {th }}$ 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 $m \times n$ matirx $A$ defines a linear transformation $T: \mathbb{R}^{n} \longrightarrow \mathbb{R}^{m}$ by $T(x)=A x$.
5. $\mathbb{R}^{n}$ has basis $e_{1}, \ldots, e_{n}$ and $\mathbb{R}^{m}$ has basis $e_{1}^{\prime}, \ldots, e_{m}^{\prime}$, 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

$$
A e_{j}=a_{1, j} e_{1}^{\prime}+a_{2, j} e_{2}^{\prime}+\ldots+a_{m, j} e_{m}^{\prime}
$$

and so the matrix of the linear transformation $T(x)=A x$ with respect to the bases, $e_{1}, \ldots, e_{n}$ and $e_{1}^{\prime} \ldots, e_{m}^{\prime}$ is $A$.

## Changing basis

Definition 0.6. If $V$ is a vector space and $v_{1}, \ldots, v_{n}$ and $v_{1}^{\prime}, \ldots, v_{n}^{\prime}$ are two bases for $V$. Then there exists an invertible $n \times n$ matrix $B$ with entries $b_{i, j}$ such that

$$
v_{j}^{\prime}=b_{1, j} v_{1}+b_{2, j} v_{2}+\ldots+b_{n, j} v_{n}
$$

$B$ is said to be the basis change matrix from $v_{1}, \ldots, v_{n}$ to $v_{1}^{\prime}, \ldots, v_{n}^{\prime}$.
Suppose $V$ and $W$ are vector spaces. Suppose that $V$ has bases $v_{1}, \ldots, v_{n}$ and $v_{1}^{\prime}, \ldots, v_{n}^{\prime}$, that $W$ has bases $w_{1}, \ldots, w_{m}$ and $w_{1}^{\prime}, \ldots, w_{m}^{\prime}$, 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 $C^{-1} A B$ with respect to the bases $v_{1}^{\prime}, \ldots, v_{n}^{\prime}$ and $w_{1}^{\prime}, \ldots, w_{m}^{\prime}$, 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 $T(v)=A v$ has matrix $A$ with respect to the standard bases $e_{1}, \ldots, e_{n}$ and $e_{1}^{\prime}, \ldots, e_{m}^{\prime}$. 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 $C^{-1} A B$ where

$$
B=\left(v_{1}|\cdots| v_{n}\right) \text { and } C=\left(w_{1}|\cdots| w_{m}\right) .
$$

## 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 $u_{j}=A v_{j} / \sigma_{j}$.
3. Let $u_{i+1}, \ldots, u_{m}$ be an ORTHONORMAL basis of $N\left(A A^{T}\right)$.
4. The SVD is

$$
\left(u_{1}|\cdots| u_{m}\right) \Sigma\left(v_{1}|\cdots| v_{n}\right)^{T}
$$

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 $A A^{T}$ and $\lambda_{i+1}, \ldots, \lambda_{m}$ be the zero eigenvalues of $A A^{T}$. Choose corresponding ORTHONORMAL eigenvectors $u_{1}, \ldots, u_{m}$ for $A A^{T}$.
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 $N\left(A^{T} A\right)$.

## Recitation 10 questions

## Question 1

Let $A=\left(\begin{array}{lll}2 & 3 & 1 \\ 1 & 1 & 5\end{array}\right)$.
(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 $T(x)=A x$ 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 $T(x)=A x$ 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

$$
A(4,5)^{T}=(1,2,3)^{T} \text { and } A(3,4)^{T}=(3,2,1)^{T}
$$

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

$$
(4,5)^{T},(3,4)^{T} \text { and }(1,2,3)^{T},(3,2,1)^{T},(0,1,0)^{T}
$$

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

## Question 3

Let $V$ be the set of cubic polynomials $V=\left\{f(x)=a x^{3}+b x^{2}+c x+d: a, b, c, d \in \mathbb{R}\right\}$.
(a) Recall why $V$ is a vector space; what happens to the coefficients under addition and scalar multiplication?
(b) Define $T_{1}(f(x))=f^{\prime}(x)$. 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 $T_{1}(1), T_{1}(x), T_{1}\left(x^{2}\right), T_{1}\left(x^{3}\right)$ ? 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 $1, x-1,(x-1)(x-2), x\left(x^{2}-\frac{9}{2} x+6\right)$ ?

## Question 4

Let $A=\left(\begin{array}{ll}1 & 2 \\ 3 & 6\end{array}\right)$.

1. Find the eigenvalues $\lambda_{1} \neq 0, \lambda_{2}$ and unit eigenvectors $v_{1}, v_{2}$ of $A^{T} A$.
2. Let $\sigma_{1}=\sqrt{\lambda_{1}}, u_{1}=A v_{1} / \sigma_{1}$. Verify that $u_{1}$ is a unit eigenvector for $A A^{T}$ with eigenvalue $\lambda_{1}$.
3. Extend $u_{1}$ to an orthonormal basis $u_{1}, u_{2}$.
4. Check that

$$
A=\left(u_{1} \mid u_{2}\right)\left(\begin{array}{cc}
\sigma_{1} & 0 \\
0 & 0
\end{array}\right)\left(v_{1} \mid v_{2}\right)^{T} .
$$

