## MATH355 2017-11-08

## COMPLEX MATRICES ARE SOMETIMES EASIER

For A, B  $\in$  M<sub>2×2</sub>, recall we say A  $\sim$  B (A is similar to B) if there exists an invertible matrix  $P \in GL_n$  such that  $B = PAP^{-1}$ .

We view similarity as follows. A and B are both representing the same linear map but in different bases. The relevant change of basis matrix is given precisely by P.

We call a matrix D diagonal if the only nonzero entries are on the diagonal. I.e., if D =  $(d_{ij})$ then  $d_i j = 0$  for  $i \neq j$ .

We call a matrix A diagonalizable of it is similar to a diagonal matrix. Explicitly, there exists P such that  $PAP^{-1}$  is a diagonal matrix. We say that P diagonalizes A.

Example 1. Consider the matrix  $A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$ . Then det A = 1 but A is not similar to I, the identity matrix. Why? Well, if  $PAP^{-1} = I$  then  $A = P^{-1}IP = P^{-1}P = I$ , but  $A \neq I$ .

Example 2. Consider B = 
$$\begin{pmatrix} 0 & 2 \\ 3 & 0 \end{pmatrix}$$
. Take P =  $\begin{pmatrix} \frac{-\sqrt{\frac{3}{2}}}{2} & \frac{1}{2} \\ \frac{\sqrt{\frac{3}{2}}}{2} & \frac{1}{2} \end{pmatrix}$ . Verify that  $P^{-1} = \begin{pmatrix} -\sqrt{\frac{3}{2}} & \sqrt{\frac{2}{3}} \\ 1 & 1 \end{pmatrix}$ .

A quick calculation reveals that

$$D := PBP^{-1} = \begin{pmatrix} -\sqrt{6} & 0\\ 0 & \sqrt{6} \end{pmatrix}$$

which is a diagonal matrix!

Why is this so great? Well, remember we interpret similarity as a change of basis. Let us do it explicitly here. Write  $q: \mathbb{R}^2 \to \mathbb{R}^2$  for the linear map defined as

$$g\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2y \\ 3x \end{pmatrix}$$

so that  $\operatorname{Rep}_{\operatorname{std}} g = B$ . Now, consider the basis  $\mathbb{B} = (\vec{v}_1, \vec{v}_2)$  where

$$\vec{v}_1 = \begin{pmatrix} -\sqrt{\frac{2}{3}} \\ 1 \end{pmatrix}$$

$$\vec{v}_2 = \begin{pmatrix} \sqrt{\frac{2}{3}} \\ \end{pmatrix}$$

$$\vec{v}_2 = \left( \sqrt{\frac{2}{3}} \right)$$

so that  $\operatorname{Rep}_{\mathbb{B},\operatorname{std}}\operatorname{id}=\operatorname{P}^{-1}$  and  $\operatorname{Rep}_{\mathbb{B}}g=\operatorname{D}$ .

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Wait, how did we figure out this last part (in the previous example) without having to compute anything? Well, first off,  $P^{-1}$  is an invertible matrix so its columns must form a basis (why?). The vectors  $\vec{v}_1, \vec{v}_2$  are indeed the columns of  $P^{-1}$ .

What is  $Rep_{\mathbb{B},std}$  id? Well, the first column is  $Rep_{std}\vec{v}_1$  and the second column is  $Rep_{std}\vec{v}_2$ . So,  $Rep_{\mathbb{B},std}$  id =  $P^{-1}$ . Also, we showed a long time ago that we always have

$$P = (P^{-1})^{-1} = Rep_{std, \mathbb{R}} id$$
.

OK, now we nailed down our change of basis matrix. How do we compute  $Rep_{\mathbb{B}}$  g? Once again, we know how to do this already (we've seen this formula many times!):

$$\operatorname{Rep}_{\mathbb{B}} g = \operatorname{Rep}_{\mathbb{B},\mathbb{B}} g = \operatorname{Rep}_{\operatorname{std},\mathbb{B}} \operatorname{Rep}_{\operatorname{std}} g \operatorname{Rep}_{\mathbb{B},\operatorname{std}} = \operatorname{PBP}^{-1}.$$

So the basis  $\mathbb B$  is the best possible basis for g, as  $Rep_{\mathbb B} g$  is diagonal! Indeed, you can easily calculate that

$$g(\vec{v}_1) = -\sqrt{6}\vec{v}_1$$
$$g(\vec{v}_2) = \sqrt{6}\vec{v}_2$$

The vectors  $\vec{v}_1$  and  $\vec{v}_2$  are called *eigenvectors* (for g). The scalars  $-\sqrt{6}$ ,  $\sqrt{6}$  are called *eigenvalues*.

Example 3. What about our  $A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$ ? Can A be diagonalized? In other words, can we find P such that PAP<sup>-1</sup> is diagonal? More explicitly, does there exist P and scalars  $\lambda, \mu \in \mathbf{R}$  such that

$$PAP^{-1} = \begin{pmatrix} \lambda & 0 \\ 0 & \mu \end{pmatrix}?$$

Let  $f: \mathbf{R}^2 \to \mathbf{R}^2$  be the linear map  $f\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -y \\ x \end{pmatrix}$ . So that  $A = \operatorname{Rep}_{std} f$ . If A were diagonalizable, then (by proceeding in the same way as the previous example) there would exist a basis  $\vec{v}_1, \vec{v}_2$  such that  $f(\vec{v}_1) = \lambda \vec{v}_1, f(\vec{v}_2) = \mu \vec{v}_2$ .

Suppose now we can find a vector  $\vec{v} \in \mathbb{R}^2$  such that  $f(\vec{v}) = \lambda \vec{v}$  (for whatever value of  $\lambda \in \mathbb{R}$ ). Say  $\vec{v} = \begin{pmatrix} a \\ b \end{pmatrix}$ , then

$$f(\vec{v}) = \begin{pmatrix} -b \\ \alpha \end{pmatrix}$$
$$\lambda \vec{v} = \begin{pmatrix} \lambda \alpha \\ \lambda b \end{pmatrix}$$

so

$$\begin{cases} -b = \lambda a \\ a = \lambda b \end{cases}$$

so that  $-b = \lambda a = \lambda^2 b$  which means  $(\lambda^2 + 1)b = 0$ . Since  $(\lambda^2 + 1)$  is never zero (for *real* values of  $\lambda$ ), we must have b = 0. But then the second equation above implies a = 0. So  $\vec{v} = \vec{0}$ .

This means we can *never* find the basis we want. So A cannot be diagonalized!

Nevertheless, we persist. Indeed, check this out. Consider the matrix  $Q = \begin{pmatrix} 1 & i \\ i & 1 \end{pmatrix}$ , where  $i \in C$  is that funny number such that  $i^2 = -1$ . Consider also  $R = \begin{pmatrix} \frac{1}{2} & -\frac{i}{2} \\ -\frac{i}{2} & \frac{1}{2} \end{pmatrix}$ . A quick calculation will show that QR = I is the identity! So  $Q = R^{-1}$ . Also, compute the following amazing fact

$$RAR^{-1} = \begin{pmatrix} -i & 0 \\ 0 & i \end{pmatrix}$$

WAIT A MINUTE! The matrix on the RHS is diagonal! I thought we couldn't diagonalize A??? The point is, we lacked imagination.

In the example above we used that  $(\lambda^2 + 1)$  is never zero. But if  $\lambda$  is allowed to be complex number, then that quantity *can* be zero:  $(\pm i)^2 + 1 = 0$ ! So, we should have enlarged the domain of definition of A (and viewed it as a linear map of  $\mathbb{C}^2$ , rather than just  $\mathbb{R}^2$ ).

The point being:

sometimes we make things bigger to understand them.

For example, say we have a polynomial  $ax^2+bx+c$  and we wish to find the roots. Luckily, we have a formula that tells us the answer! However, even if a, b,  $c \in \mathbf{R}$  the formula (implicitly) goes via the complex numbers. It's a good thing that we have a formula for the roots, even when the roots aren't real. The point being: if we didn't know about the complex numbers we couldn't have a homogeneous formula to calculate the roots of a quadratic polynomial. The same philosophy applies to diagonalizing matrices.

OK, what is a complex vector space anyway? Well, it's a set of vectors V with addition and scalar multiplication, only now the scalars are allowed to be complex numbers. For instance, if  $\vec{v} \in V$  the quantity  $(3+2i)\vec{v}=3\vec{v}+i(2\vec{v})\in V$  makes sense.

Example 4.  $\mathbb{R}^3$  is *not* a complex vector space: the only meaning we give to the quantity  $\begin{pmatrix} 3 \\ 1 \\ 2 \end{pmatrix}$  is

the vector 
$$\begin{pmatrix} 3i\\i\\2i \end{pmatrix}$$
 which lives in  $C^3$  but not  $R^3$ .

On the other hand,  $\mathbb{C}^3$  is a complex vector space.

Now, if  $\vec{v}_1, \dots, \vec{v}_k \in \mathbf{C}^n$  are a bunch of vectors in  $\mathbf{C}^n$  (or any complex vector space V), we should be careful to distinguish between two things.

A C-linear combination is 
$$\sum_j \lambda_j \vec{v}_j$$
 with  $\lambda_j \in C$ 

An **R**-linear combination is 
$$\sum_{i}^{n} \alpha_{i} \vec{v}_{i}$$
 with  $\alpha_{i} \in \mathbf{R}$ .

This is crux the matter: whether we allow arbitrary complex numbers as scalars, or just real numbers.

Using this, we have different notions of span, linear independence, linear dependence, subspace, dimension etc.

Example 5. Take 
$$\vec{v} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$
,  $\vec{w} = \begin{pmatrix} 0 \\ i \end{pmatrix}$  in  $\mathbf{C}^2$ . Then 
$$\operatorname{Span}_{\mathbf{C}} \{ \vec{v}, \vec{w} \} = \mathbf{C}^2$$
 
$$\operatorname{Span}_{\mathbf{R}} \{ \vec{v}, \vec{w} \} \neq \mathbf{C}^2$$

(exercise!) indeed,

$$\operatorname{Span}_{\mathbf{R}}\{\vec{v}, \vec{w}\} = \left\{ \begin{pmatrix} a \\ ib \end{pmatrix} \mid a, b \in \mathbf{R} \right\}$$

and this is only a *real* subspace of  $\mathbb{C}^2$  but not a complex one (why? what does real/complex subspace mean? what's the difference?)

Example 6. The vectors 
$$\begin{pmatrix} 1 \\ 0 \end{pmatrix}$$
,  $\begin{pmatrix} 0 \\ -2 \end{pmatrix}$  are a **C**-basis of  $\mathbf{C}^2$  (why?). While  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ -2 \end{pmatrix}$  are *not* an **R**-basis of  $\mathbf{C}^2$ . However,  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ -2 \end{pmatrix}$ ,  $\begin{pmatrix} i \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ -2i \end{pmatrix}$  is an **R**-basis of  $\mathbf{C}^2$  (why?). So we see that  $\dim_{\mathbf{C}} \mathbf{C}^2 = 2$  while  $\dim_{\mathbf{R}} \mathbf{C}^2 = 4$ .

In any case, don't stress too much about vector spaces over the complex numbers. I promise you that things will become obvious in a few days (you'll realize complex vector spaces aren't at all that complex).

We will follow the following convention:

if we don't say anything, vector spaces, matrices, linear maps, bases, dimension etc. will be over the real numbers.