

## 30 2025-11-03 | Week 11 | Lecture 30

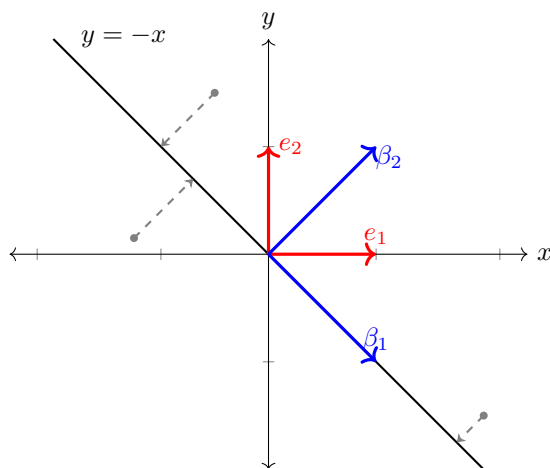
*The nexus question of this lecture: What can we say about two matrices that represent the same linear transformation?*

This lecture draws from G. Strang's textbook "Linear Algebra and its Applications"

### 30.1 Similarity

**Example 136** (Two matrices that represent the same projection). Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the projection onto the line  $y = -x$ . We don't need a basis to describe this transformation. Even without a matrix, we can observe geometric properties of  $T$ :

- the dimension of its range is 1, so it has rank 1
- the dimension of its kernel (the line  $y = x$ ) is 1, so the nullity is 1
- it collapses space, so has determinant 0
- it has two eigenvalues: 1 (with eigenvector  $\beta_1$ ) and 0 (with eigenvector  $\beta_2$ )



Recall we can always represent a linear transformation  $T : V \rightarrow V$  in the form of a matrix, but this representation depends on a choice of basis for  $V$  (see Section 24.2). We'll compare two choices of basis,  $\alpha = \{e_1, e_2\}$  and  $\beta = \{\beta_1, \beta_2\}$  (shown in the picture above).

- **The matrix**  $[T]_{\beta}^{\beta}$ . Note that  $\beta_1$  and  $\beta_2$  are eigenvectors of  $T$ . They are linearly independent and therefore form a basis (this doesn't always happen, but when it does, eigenvectors are the best basis).

Using the notation  $\begin{bmatrix} a \\ b \end{bmatrix}_{\beta} = a\beta_1 + b\beta_2$ , observe from the figure that

$$T(\beta_1) = \beta_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}_{\beta} \quad \text{and} \quad T(\beta_2) = 0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}_{\beta}.$$

Hence

$$[T]_{\beta}^{\beta} = \begin{bmatrix} | & | \\ T(\beta_1) & T(\beta_2) \\ | & | \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}.$$

This is the matrix of  $T$  with respect to the basis  $\beta$ . The first column comes from the first basis vector (projected onto itself). The second column comes from the basis vector that is projected to zero.

- **The matrix  $[T]_\alpha^\alpha$ .** We'll compute this matrix using a change-of-basis matrix.

Let  $P$  be the change-of-basis matrix from  $\alpha$  to  $\beta$ . To compute  $P$ , write  $\beta_1, \beta_2$  in terms of  $\alpha_1, \alpha_2$ :

$$\beta_1 = e_1 - e_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}_\alpha \quad \text{and} \quad \beta_2 = e_1 + e_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}_\alpha$$

Therefore the change of basis matrix from  $\alpha$  to  $\beta$  is

$$P = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

This matrix has inverse

$$P^{-1} = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}.$$

Now we compute  $[T]_\alpha^\alpha$ . By Theorem 116,

$$[T]_\beta^\beta = P^{-1}[T]_\alpha^\alpha P, \tag{25}$$

which implies

$$[T]_\alpha^\alpha = P[T]_\beta^\beta P^{-1}.$$

Therefore

$$\begin{aligned} [T]_\alpha^\alpha &= \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix}. \end{aligned}$$

**Conclusion:** the matrices

$$[T]_\alpha^\alpha = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix} \quad \text{and} \quad [T]_\beta^\beta = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

both represent the same linear transformation  $T$  (a projection), but with different choice of bases. The linear transformation  $T$  has many geometric properties which don't depend on the choice of basis:

- the dimension of its range (the 'rank')
- the dimension of its kernel (the 'nullity')
- the determinant
- its eigenvalues

These properties don't hinge on how we choose to represent  $T$  as a matrix. Therefore although the matrices  $[T]_\alpha^\alpha$  and  $[T]_\beta^\beta$  are different, they share the same rank, determinant, nullity, etc.

End of Example 136.  $\square$

The two matrices  $[T]_\alpha^\alpha, [T]_\beta^\beta$  from Example 136 are an example of *similar matrices*, which we define next.

**Definition 137** (Similar). Let  $A, B \in \mathcal{M}_{n \times n}$ . We say that  $B$  is *similar* to  $A$  if there exists an invertible matrix  $P$  such that

$$B = P^{-1}AP. \tag{26}$$

**Question** What does it *mean* for two matrices to be similar?

**Answer:** It means that they represent the same linear transformation with respect to different bases. This is because the formula Eq. (26) is the change-of-basis formula from Theorem 116 (like we used in Eq. (25)).

To be precise, if  $A, B \in \mathcal{M}_{n \times n}(\mathbb{R})$  are similar, then there is a linear transformation  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  such that

$$A = [T]_\alpha^\alpha \quad \text{and} \quad B = [T]_\beta^\beta$$

for some choice of bases  $\alpha$  and  $\beta$  of  $\mathbb{R}^n$ . Since quantities like rank, nullity, determinant, trace, and eigenvalues all represent geometric properties inherent to the linear transformation  $T$ , and since  $A$  and  $B$  are merely different representations of this transformation, the following theorem holds:

**Theorem 138.** *If  $A$  and  $B$  are similar, then they have the same rank, nullity, determinant, trace, eigenvalues, and characteristic polynomial.*