

Similarity of Matrices

Prerequisites:

- Matrix Representations of Linear Transformations

Definition 1: *Similarity, (Conjugacy)*

Suppose that \boxed{A} and $\widetilde{\boxed{A}}$ are two $N \times N$ matrices. We say that \boxed{A} and $\widetilde{\boxed{A}}$ are **similar** (or **conjugate**), and write

$$\boxed{A} \sim \widetilde{\boxed{A}}$$

if there is an invertible $N \times N$ matrix \boxed{B} so that

$$\widetilde{\boxed{A}} = \boxed{B}^{-1} \cdot \boxed{A} \cdot \boxed{B}$$

Example 2: $\boxed{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ is similar to $\widetilde{\boxed{A}} = \begin{bmatrix} 5 & 1 \\ 2 & 0 \end{bmatrix}$. To see this, let $\boxed{B} = \begin{bmatrix} \frac{1}{2} & \frac{-1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$. Then $\boxed{B}^{-1} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$. Thus,

$$\begin{aligned} \boxed{B}^{-1} \cdot \boxed{A} \cdot \boxed{B} &= \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \cdot \begin{bmatrix} \frac{1}{2} & \frac{-1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \\ &= \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \cdot \begin{bmatrix} \frac{3}{2} & \frac{1}{2} \\ \frac{7}{2} & \frac{1}{2} \end{bmatrix} \\ &= \begin{bmatrix} 5 & 1 \\ 2 & 0 \end{bmatrix} \\ &= \widetilde{\boxed{A}} \end{aligned}$$

Remark 3:

1. In other words, $\boxed{A} \sim \widetilde{\boxed{A}}$ if there is an invertible $N \times N$ matrix \boxed{B} so that

$$\boxed{B} \cdot \widetilde{\boxed{A}} = \boxed{A} \cdot \boxed{B}$$

2. Notice that

- Any matrix $\boxed{\mathbf{A}}$ is similar to itself, because

$$\boxed{\mathbf{A}} = \boxed{\mathbf{Id}}^{-1} \cdot \boxed{\mathbf{A}} \cdot \boxed{\mathbf{Id}}$$

- If $\boxed{\mathbf{A}} \sim \boxed{\widetilde{\mathbf{A}}}$, then $\boxed{\widetilde{\mathbf{A}}} \sim \boxed{\mathbf{A}}$, because

$$\begin{aligned} & \text{if } \boxed{\widetilde{\mathbf{A}}} = \boxed{\mathbf{B}}^{-1} \cdot \boxed{\mathbf{A}} \cdot \boxed{\mathbf{B}} \\ \text{then } & \boxed{\mathbf{B}} \cdot \boxed{\widetilde{\mathbf{A}}} \cdot \boxed{\mathbf{B}}^{-1} = \boxed{\mathbf{A}} \end{aligned}$$

- If $\boxed{\mathbf{X}} \sim \boxed{\mathbf{Y}}$ and $\boxed{\mathbf{Y}} \sim \boxed{\mathbf{Z}}$, then $\boxed{\mathbf{X}} \sim \boxed{\mathbf{Z}}$:

$$\begin{aligned} & \text{if } \boxed{\mathbf{Y}} = \boxed{\mathbf{B}}^{-1} \cdot \boxed{\mathbf{X}} \cdot \boxed{\mathbf{B}} \\ \text{and } & \boxed{\mathbf{Z}} = \boxed{\mathbf{C}}^{-1} \cdot \boxed{\mathbf{Y}} \cdot \boxed{\mathbf{C}} \\ \text{then } & \boxed{\mathbf{Z}} = \boxed{\mathbf{C}}^{-1} \cdot \boxed{\mathbf{B}}^{-1} \cdot \boxed{\mathbf{X}} \cdot \boxed{\mathbf{B}} \cdot \boxed{\mathbf{C}} \\ & = (\boxed{\mathbf{B}} \cdot \boxed{\mathbf{C}})^{-1} \cdot \boxed{\mathbf{X}} \cdot (\boxed{\mathbf{B}} \cdot \boxed{\mathbf{C}}). \end{aligned}$$

Matrix similarity is important because of its relationship to the *representation* of **linear transformations**....

Proposition 4: *Similarity of Matrix Representations for Linear Transformations*

Let \mathbb{V} be a finite-dimensional vector spaces, and $f : \mathbb{V} \rightarrow \mathbb{V}$ a linear transformation.

1. Suppose that \mathcal{A} and $\widetilde{\mathcal{A}}$ are two bases for \mathbb{V} . Let $\boxed{\mathbf{F}}$ be the matrix representation of f with respect to \mathcal{A} , and $\boxed{\widetilde{\mathbf{F}}}$ be the matrix representation of f with respect to $\widetilde{\mathcal{A}}$. Then $\boxed{\mathbf{F}}$ and $\boxed{\widetilde{\mathbf{F}}}$ are similar matrices.
2. Suppose \mathcal{A} is any basis of \mathbb{V} , and $\boxed{\mathbf{F}}$ is the matrix of f with respect to \mathcal{A} . If $\boxed{\widetilde{\mathbf{F}}}$ is *any* matrix similar to $\boxed{\mathbf{F}}$, then there is a basis $\widetilde{\mathcal{A}}$ for \mathbb{V} so that $\boxed{\widetilde{\mathbf{F}}}$ is the matrix of f with respect to $\widetilde{\mathcal{A}}$.

Proof:

Proof of Part 1: Recall: if $\boxed{\mathbf{B}}$ is the **change-of-basis** matrix from \mathcal{A} to $\widetilde{\mathcal{A}}$, then

$$\widetilde{\mathbf{F}} = \mathbf{B} \cdot \mathbf{F} \cdot \mathbf{B}^{-1}$$

is the matrix representation of f with respect to \mathcal{B} . Thus, $\widetilde{\mathbf{F}}$ and \mathbf{F} are similar

Proof of Part 2: Suppose that $\widetilde{\mathbf{F}} = \mathbf{B} \cdot \mathbf{F} \cdot \mathbf{B}^{-1}$ for some matrix \mathbf{B} . Thus, we want to find a basis $\widetilde{\mathcal{A}}$ so that that \mathbf{B} is the change-of-basis matrix from \mathcal{A} to $\widetilde{\mathcal{A}}$. Thus, \mathbf{B}^{-1} should be the change-of-basis matrix from $\widetilde{\mathcal{A}}$ to \mathcal{A} . Thus means that the column-vectors of \mathbf{B}^{-1} should be the *coordinates* of the elements of $\widetilde{\mathcal{A}}$, relative to \mathcal{A} .

So, suppose $\mathcal{A} = \{\mathbf{a}_1, \dots, \mathbf{a}_N\}$, and suppose $\mathbf{B}^{-1} = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1N} \\ c_{21} & c_{22} & \dots & c_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ c_{N1} & c_{N2} & \dots & c_{NN} \end{bmatrix}$,

then, for all $k \in [1 \dots N]$, simply define

$$\widetilde{\mathbf{a}}_k = \sum_{n=1}^N c_{nk} \mathbf{a}_n$$

and let $\widetilde{\mathcal{A}} = \{\widetilde{\mathbf{a}}_1, \dots, \widetilde{\mathbf{a}}_N\}$. Then $\widetilde{\mathcal{A}}$ is the basis we seek.

□ [Proposition 4]

Similarity Invariants

In general, it is quite difficult to tell if two matrices \mathbf{A} and $\widetilde{\mathbf{A}}$ are similar. We need a matrix \mathbf{B} so that $\mathbf{B} \cdot \widetilde{\mathbf{A}} = \mathbf{A} \cdot \mathbf{B}$. But how can we find such a \mathbf{B} , assuming it exists? We might spend a long time looking, only to realize that the two matrices are *not* similar; thus, there *is* no such \mathbf{B} , and we are wasting our time.

It would be nice if there was a quick way to tell when two matrices are *not* similar. This is the purpose of a *similarity invariant*.

Definition 5: *Similarity Invariant*

A **similarity invariant** is a function $f : \mathcal{M}_{N \times N} \rightarrow \mathbb{S}$ (where \mathbb{S} is some set) so that, for any matrices \mathbf{A} and $\widetilde{\mathbf{A}}$,

$$\left(\mathbf{A} \text{ and } \widetilde{\mathbf{A}} \text{ are similar} \right) \implies \left(f(\mathbf{A}) = f(\widetilde{\mathbf{A}}) \right)$$

Hence, if $f(\boxed{A}) \neq f(\widetilde{\boxed{A}})$ then we know right away that \boxed{A} and $\widetilde{\boxed{A}}$ are *not* similar; there is no point looking for \boxed{B} .

Note: If $f(\boxed{A}) = f(\widetilde{\boxed{A}})$, this does *not* automatically mean that \boxed{A} and $\widetilde{\boxed{A}}$ are similar. It only means that they *might* be similar.

Theorem 6: *Some Similarity Invariants*

Suppose \boxed{A} and $\widetilde{\boxed{A}}$ are $N \times N$ matrices. If \boxed{A} is **similar** to $\widetilde{\boxed{A}}$, then:

1. $\text{rank}(\boxed{A}) = \text{rank}(\widetilde{\boxed{A}})$.
2. $\text{nullity}(\boxed{A}) = \text{nullity}(\widetilde{\boxed{A}})$.
3. $\det(\boxed{A}) = \det(\widetilde{\boxed{A}})$.
4. \boxed{A} and $\widetilde{\boxed{A}}$ have the same **characteristic polynomial**.

In other words, the **rank**, **nullity**, **determinant**, and **characteristic polynomial** of a matrix are all *similarity invariants*.

Proof: **Part 1** and **Part 2** are left as exercises.

Proof of Part 3: Suppose \boxed{A} and $\widetilde{\boxed{A}}$ are similar; thus, there is an invertible matrix \boxed{B} so that $\widetilde{\boxed{A}} = \boxed{B}^{-1} \cdot \boxed{A} \cdot \boxed{B}$. Thus,

$$\begin{aligned} \det(\widetilde{\boxed{A}}) &= \det(\boxed{B}^{-1} \cdot \boxed{A} \cdot \boxed{B}) \\ &= \det(\boxed{B}^{-1}) \cdot \det(\boxed{A}) \cdot \det(\boxed{B}) \\ &= \det(\boxed{B})^{-1} \cdot \det(\boxed{B}) \cdot \det(\boxed{A}) \\ &= \det(\boxed{A}) \end{aligned}$$

Proof of Part 4: Suppose \boxed{A} and $\widetilde{\boxed{A}}$ are similar; thus, there is an invertible matrix \boxed{B} so that $\widetilde{\boxed{A}} = \boxed{B}^{-1} \cdot \boxed{A} \cdot \boxed{B}$. But also notice, for any fixed $\lambda \in \mathbb{R}$,

$$\lambda \mathbf{Id} = \mathbf{B}^{-1} \cdot \lambda \mathbf{Id} \cdot \mathbf{B}.$$

Thus, for any fixed λ ,

$$\begin{aligned} \widetilde{\mathbf{A}} - \lambda \mathbf{Id} &= (\mathbf{B}^{-1} \cdot \mathbf{A} \cdot \mathbf{B}) - (\mathbf{B}^{-1} \cdot \lambda \mathbf{Id} \cdot \mathbf{B}) \\ &= \mathbf{B}^{-1} \cdot (\mathbf{A} - \lambda \mathbf{Id}) \cdot \mathbf{B} \end{aligned}$$

Thus, by **Part 3**,

$$\tilde{c}(\lambda) = \det(\widetilde{\mathbf{A}} - \lambda \mathbf{Id}) = \det(\mathbf{A} - \lambda \mathbf{Id}) = c(\lambda)$$

where c and \tilde{c} are the **characteristic polynomials** of \mathbf{A} and $\widetilde{\mathbf{A}}$, respectively.

Since this is true for *all* λ , the functions $c(x)$ and $\tilde{c}(x)$ are equal *everywhere*—they must be the same polynomial:

$$c(x) = \tilde{c}(x).$$

□ [Theorem 6]

Example 7: Recall from example 2 on page 1 that $\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \sim \begin{bmatrix} 5 & 1 \\ 2 & 0 \end{bmatrix} = \widetilde{\mathbf{A}}$. Thus

$$\begin{aligned} \text{nullity} \left[\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \right] &= 0 = \text{nullity} \left[\begin{bmatrix} 5 & 1 \\ 2 & 0 \end{bmatrix} \right] \\ \text{rank} \left[\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \right] &= 2 = \text{rank} \left[\begin{bmatrix} 5 & 1 \\ 2 & 0 \end{bmatrix} \right] \\ \det \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} &= -2 = \det \begin{bmatrix} 5 & 1 \\ 2 & 0 \end{bmatrix} \end{aligned}$$

and both matrices have characteristic polynomial $x^2 - 5x + 2$.

However, if $\mathbf{C} = \begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix}$, then

$$\begin{aligned} \text{nullity} \begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} &= 1 \\ \text{rank} \begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} &= 1 \\ \det \begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix} &= 0 \end{aligned}$$

and the characteristic polynomial of \boxed{C} is $x^2 - 7x$. Any single one of these four facts would be enough to prove that \boxed{C} could not be similar to \boxed{A} .

Corollary 8: The **spectrum** of a matrix is also a similarity invariant.

Proof: If \boxed{A} and $\widetilde{\boxed{A}}$ are similar, then they have the same characteristic polynomials. The **spectrum** of \boxed{A} is simply the list of all roots of its characteristic polynomial; thus, \boxed{A} and $\widetilde{\boxed{A}}$ must have the same spectrum.

□ [Corollary 8]

Corollary 9: *The characteristic polynomial of a transformation*

Let \mathbb{V} be a finite-dimensional vector space, and $f : \mathbb{V} \rightarrow \mathbb{V}$ a linear transformation.

If \boxed{F} and $\widetilde{\boxed{F}}$ are two different **matrix representations** of f (with respect to two different bases \mathcal{B} and $\widetilde{\mathcal{B}}$ of \mathbb{V}), then \boxed{F} and $\widetilde{\boxed{F}}$ are *similar*, and therefore have the *same determinant* and the *same characteristic polynomial*.

In other words, the *determinant* and *characteristic polynomial* of a linear transformation on an abstract vector space are well-defined, *independent* of the choice of basis.

Proof: Exercise

□ [Corollary 9]

Another similarity invariant is a function called the **trace**

Definition 10: *Trace*

Let \boxed{A} be an $N \times N$ matrix. The **trace** of \boxed{A} is the *sum of the diagonal*

entries in \boxed{A} . In other words, if $\boxed{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ a_{N1} & a_{N2} & \dots & a_{NN} \end{bmatrix}$, then

$$\text{trace } \boxed{\mathbf{A}} = a_{11} + a_{22} + \dots + a_{nn}$$

Example 11:

- $\text{trace} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = 1 + 4 = 5.$
- The trace of the $N \times N$ identity matrix $\boxed{\mathbf{Id}_N}$ is $N.$

Proposition 12: *Properties of the Trace*

1. $\text{trace} : \mathcal{M}_{N \times N} \rightarrow \mathbb{R}$ is a linear function. In other words, $\text{trace } \boxed{\mathbf{A} + \mathbf{B}} = \text{trace } \boxed{\mathbf{A}} + \text{trace } \boxed{\mathbf{B}},$ and $\text{trace } \boxed{r \cdot \mathbf{A}} = r \cdot \text{trace } \boxed{\mathbf{A}}.$
2. If $\boxed{\mathbf{A}}$ and $\boxed{\mathbf{B}}$ are two $N \times N$ matrices, then

$$\text{trace } \boxed{\mathbf{A} \cdot \mathbf{B}} = \text{trace } \boxed{\mathbf{B} \cdot \mathbf{A}}$$

Proof: **Part 1** is an exercise. To see **Part 2**, suppose that

$$\boxed{\mathbf{A}} \cdot \boxed{\mathbf{B}} = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1N} \\ c_{21} & c_{22} & \dots & c_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ c_{N1} & c_{N2} & \dots & c_{NN} \end{bmatrix},$$

and

$$\boxed{\mathbf{B}} \cdot \boxed{\mathbf{A}} = \begin{bmatrix} \tilde{c}_{11} & \tilde{c}_{12} & \dots & \tilde{c}_{1N} \\ \tilde{c}_{21} & \tilde{c}_{22} & \dots & \tilde{c}_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ \tilde{c}_{N1} & \tilde{c}_{N2} & \dots & \tilde{c}_{NN} \end{bmatrix}.$$

Notice that,

$$\text{for any } n \in [1 \dots N], \quad c_{nn} = \sum_{k=1}^N a_{nk} b_{kn},$$

and, for any $k \in [1 \dots N], \quad \tilde{c}_{kk} = \sum_{n=1}^N b_{kn} a_{nk},$

$$\begin{aligned}
\text{Thus, trace } [\mathbf{A} \cdot \mathbf{B}] &= \sum_{n=1}^N c_{nn} \\
&= \sum_{n=1}^N \sum_{k=1}^N a_{nk} b_{kn} \\
&= \sum_{k=1}^N \sum_{n=1}^N b_{kn} a_{nk} \\
&= \sum_{k=1}^N \tilde{c}_{kk} \\
&= \text{trace } [\mathbf{B} \cdot \mathbf{A}]
\end{aligned}$$

□ [Proposition 12]

Corollary 13: The trace is a **similarity invariant**.

Proof: Exercise. Use the previous theorem

□ [Corollary 13]