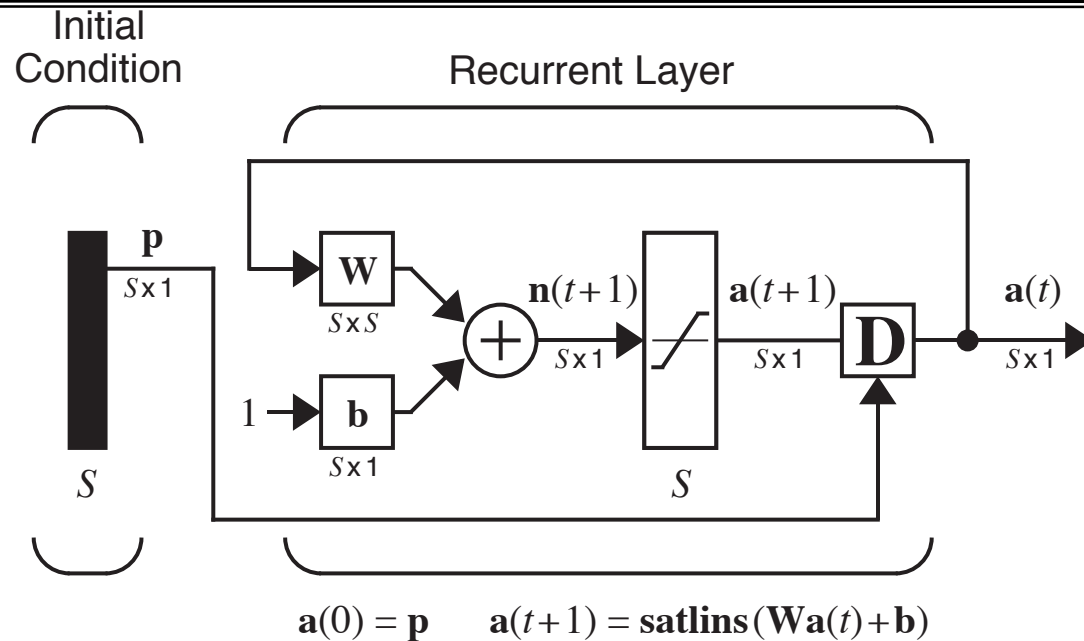




# Linear Transformations

## 6

## Hopfield Network Questions



- The network output is repeatedly multiplied by the weight matrix  $W$ .
- What is the effect of this repeated operation?
- Will the output converge, go to infinity, oscillate?
- In this chapter we want to investigate matrix multiplication, which represents a general linear transformation.



A **transformation** consists of three parts:

1. A set of elements  $X = \{\chi_i\}$ , called the domain,
2. A set of elements  $Y = \{y_i\}$ , called the range, and
3. A rule relating each  $\chi_i \in X$  to an element  $y_i \in Y$ .

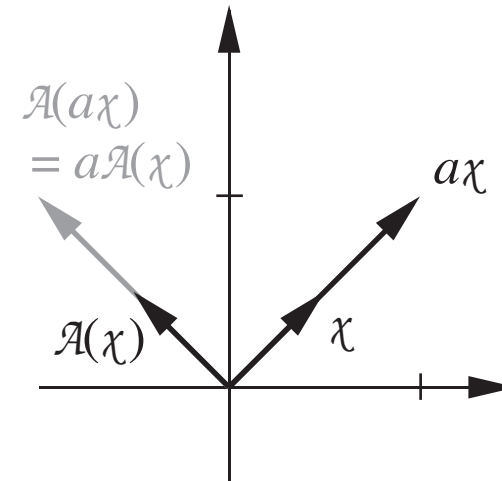
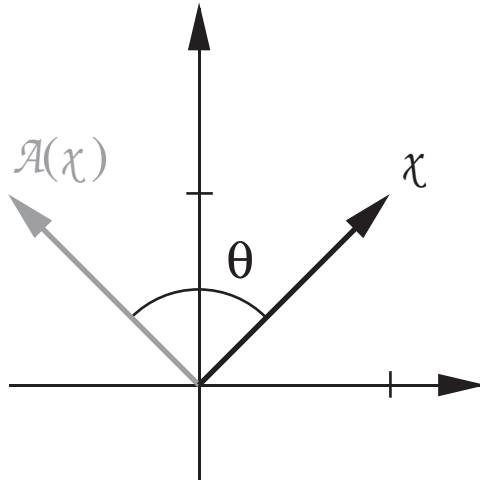
A transformation is **linear** if:

1. For all  $\chi_1, \chi_2 \in X$ ,  $\mathcal{A}(\chi_1 + \chi_2) = \mathcal{A}(\chi_1) + \mathcal{A}(\chi_2)$ ,
2. For all  $\chi \in X$ ,  $a \in \mathfrak{R}$ ,  $\mathcal{A}(a\chi) = a\mathcal{A}(\chi)$ .

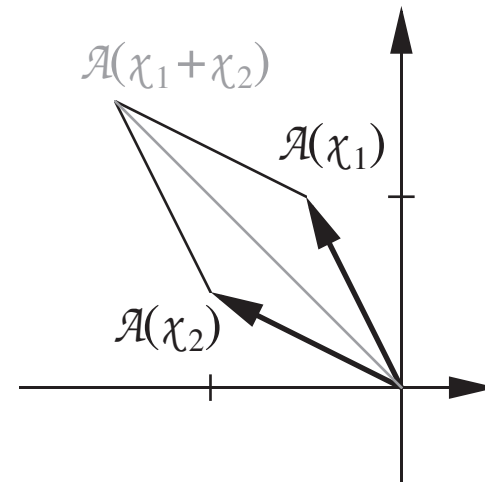
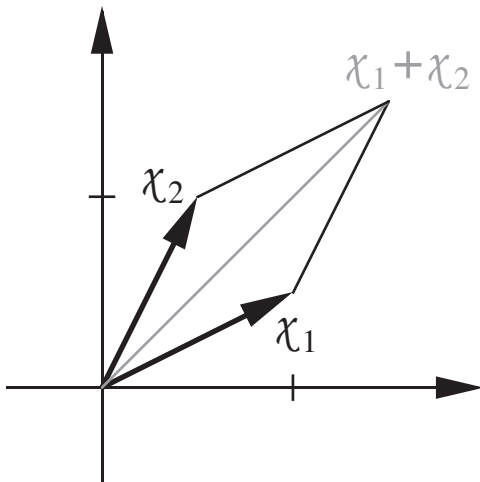


Is rotation linear?

1.



2.



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## Matrix Representation - (1)



Any linear transformation between two finite-dimensional vector spaces can be represented by matrix multiplication.

Let  $\{v_1, v_2, \dots, v_n\}$  be a basis for  $X$ , and let  $\{u_1, u_2, \dots, u_m\}$  be a basis for  $Y$ .

$$\chi = \sum_{i=1}^n x_i v_i \qquad y = \sum_{i=1}^m y_i u_i$$

Let  $\mathcal{A}: X \rightarrow Y$

$$\mathcal{A}(\chi) = y$$

$$\mathcal{A}\left(\sum_{j=1}^n x_j v_j\right) = \sum_{i=1}^m y_i u_i$$



Since  $A$  is a linear operator,

$$\sum_{j=1}^n x_j \mathcal{A}(v_j) = \sum_{i=1}^m y_i u_i$$

Since the  $u_i$  are a basis for  $Y$ ,

$$\mathcal{A}(v_j) = \sum_{i=1}^m a_{ij} u_i$$

(The coefficients  $a_{ij}$  will make up the matrix representation of the transformation.)

$$\sum_{j=1}^n x_j \sum_{i=1}^m a_{ij} u_i = \sum_{i=1}^m y_i u_i$$

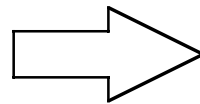


$$\sum_{i=1}^m u_i \sum_{j=1}^n a_{ij} x_j = \sum_{i=1}^m y_i u_i$$

$$\sum_{i=1}^m u_i \left( \sum_{j=1}^n a_{ij} x_j - y_i \right) = 0$$

Because the  $u_i$  are independent,

$$\sum_{j=1}^n a_{ij} x_j = y_i$$



This is equivalent to matrix multiplication.

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}$$



- A linear transformation can be represented by matrix multiplication.
- To find the matrix which represents the transformation we must transform each basis vector for the domain and then expand the result in terms of the basis vectors of the range.

$$\mathcal{A}(v_j) = \sum_{i=1}^m a_{ij} u_i$$

Each of these equations gives us one column of the matrix.

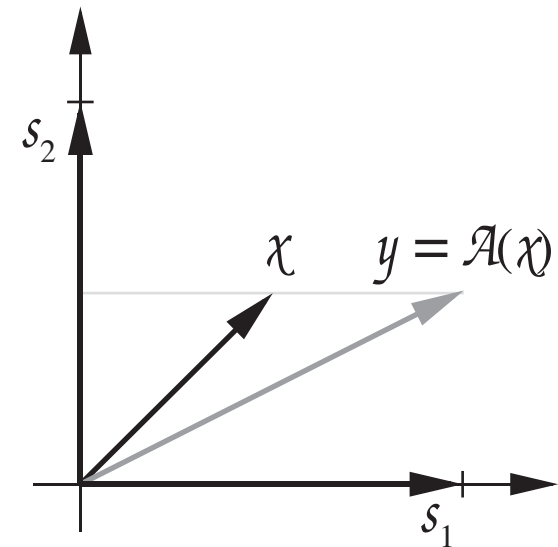
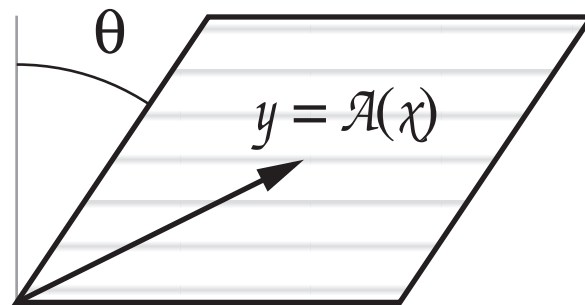
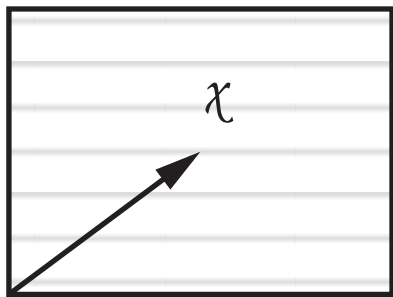


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## Example - (1)



Stand a deck of playing cards on edge so that you are looking at the deck sideways. Draw a vector  $x$  on the edge of the deck. Now “skew” the deck by an angle  $\theta$ , as shown below, and note the new vector  $y = A(x)$ . What is the matrix of this transformation in terms of the standard basis set?



## Example - (2)



To find the matrix we need to transform each of the basis vectors.

$$\mathcal{A}(v_j) = \sum_{i=1}^m a_{ij} u_i$$

We will use the standard basis vectors for both the domain and the range.

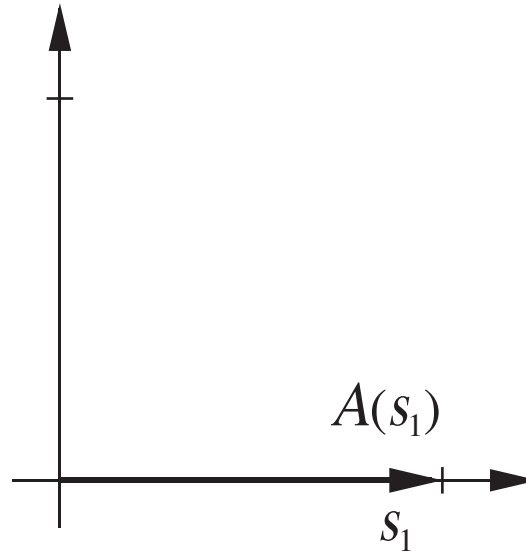
$$A(s_j) = \sum_{i=1}^2 a_{ij} s_i = a_{1j} s_1 + a_{2j} s_2$$

## Example - (3)



We begin with  $s_1$ :

If we draw a line on the bottom card and then skew the deck, the line will not change.



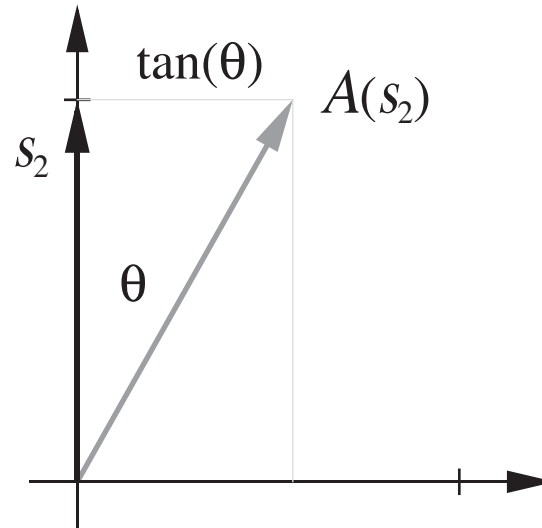
$$A(s_1) = 1s_1 + 0s_2 = \sum_{i=1}^2 a_{i1}s_i = a_{11}s_1 + a_{21}s_2$$

This gives us the first column of the matrix.

## Example - (4)



Next, we skew  $s_2$ :



$$A(s_2) = \tan(\theta)s_1 + 1s_2 = \sum_{i=1}^2 a_{i2}s_i = a_{12}s_1 + a_{22}s_2$$

This gives us the second column of the matrix.



The matrix of the transformation is:

$$\mathbf{A} = \begin{bmatrix} 1 & \tan(\theta) \\ 0 & 1 \end{bmatrix}$$



Consider the linear transformation  $\mathcal{A}: X \rightarrow Y$ . Let  $\{v_1, v_2, \dots, v_n\}$  be a basis for  $X$ , and let  $\{u_1, u_2, \dots, u_m\}$  be a basis for  $Y$ .

$$\chi = \sum_{i=1}^n x_i v_i \qquad y = \sum_{i=1}^m y_i u_i$$

$$\mathcal{A}(\chi) = y$$

The matrix representation is:

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}$$

$$\mathbf{Ax} = \mathbf{y}$$



Now let's consider different basis sets. Let  $\{t_1, t_2, \dots, t_n\}$  be a basis for  $X$ , and let  $\{w_1, w_2, \dots, w_m\}$  be a basis for  $Y$ .

$$x = \sum_{i=1}^n x'_i t_i \qquad y = \sum_{i=1}^m y'_i w_i$$

The new matrix representation is:

$$\begin{bmatrix} a'_{11} & a'_{12} & \cdots & a'_{1n} \\ a'_{21} & a'_{22} & \cdots & a'_{2n} \\ \vdots & \vdots & & \vdots \\ a'_{m1} & a'_{m2} & \cdots & a'_{mn} \end{bmatrix} \begin{bmatrix} x'_1 \\ x'_2 \\ \vdots \\ x'_n \end{bmatrix} = \begin{bmatrix} y'_1 \\ y'_2 \\ \vdots \\ y'_m \end{bmatrix}$$

$$\mathbf{A}'\mathbf{x}' = \mathbf{y}'$$

## 6

How are  $\mathbf{A}$  and  $\mathbf{A}'$  related?

Expand  $t_i$  in terms of the original basis vectors for  $X$ .

$$t_i = \sum_{j=1}^n t_{ji} \mathbf{v}_j \qquad \mathbf{t}_i = \begin{bmatrix} t_{1i} \\ t_{2i} \\ \vdots \\ t_{ni} \end{bmatrix}$$

Expand  $w_i$  in terms of the original basis vectors for  $Y$ .

$$w_i = \sum_{j=1}^m w_{ji} \mathbf{u}_j \qquad \mathbf{w}_i = \begin{bmatrix} w_{1i} \\ w_{2i} \\ \vdots \\ w_{mi} \end{bmatrix}$$



# How are $\mathbf{A}$ and $\mathbf{A}'$ related?



$$\mathbf{B}_t = [\mathbf{t}_1 \ \mathbf{t}_2 \ \dots \ \mathbf{t}_n] \quad \mathbf{x} = x'_1 \mathbf{t}_1 + x'_2 \mathbf{t}_2 + \dots + x'_n \mathbf{t}_n = \mathbf{B}_t \mathbf{x}'$$

$$\mathbf{B}_w = [\mathbf{w}_1 \ \mathbf{w}_2 \ \dots \ \mathbf{w}_m] \quad \mathbf{y} = \mathbf{B}_w \mathbf{y}'$$

$$\mathbf{A}\mathbf{x} = \mathbf{y} \quad \Rightarrow \quad \mathbf{A}\mathbf{B}_t \mathbf{x}' = \mathbf{B}_w \mathbf{y}'$$

$$[\mathbf{B}_w^{-1} \mathbf{A} \mathbf{B}_t] \mathbf{x}' = \mathbf{y}'$$

$$\mathbf{A}' \mathbf{x}' = \mathbf{y}'$$

$$\mathbf{A}' = [\mathbf{B}_w^{-1} \mathbf{A} \mathbf{B}_t]$$

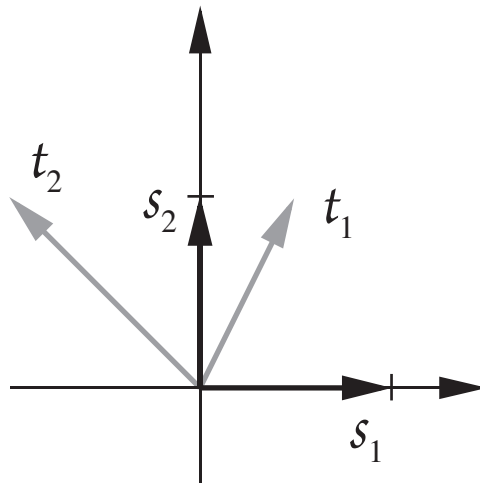
Similarity  
Transform

6

# Example - (1)



Take the skewing problem described previously, and find the new matrix representation using the basis set  $\{s_1, s_2\}$ .

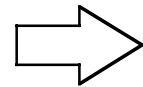


$$t_1 = 0.5s_1 + s_2$$

$$t_2 = -s_1 + s_2$$

$$\mathbf{t}_1 = \begin{bmatrix} 0.5 \\ 1 \end{bmatrix}$$

$$\mathbf{t}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$



$$\mathbf{B}_t = [\mathbf{t}_1 \ \mathbf{t}_2] = \begin{bmatrix} 0.5 & -1 \\ 1 & 1 \end{bmatrix}$$

$$\mathbf{B}_w = \mathbf{B}_t = \begin{bmatrix} 0.5 & -1 \\ 1 & 1 \end{bmatrix}$$

(Same basis for domain and range.)

## Example - (2)



$$\mathbf{A}' = [\mathbf{B}_w^{-1} \mathbf{A} \mathbf{B}_t] = \begin{bmatrix} 2/3 & 2/3 \\ -2/3 & 1/3 \end{bmatrix} \begin{bmatrix} 1 & \tan\theta \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0.5 & -1 \\ 1 & 1 \end{bmatrix}$$

$$\mathbf{A}' = \begin{bmatrix} (2/3)\tan\theta + 1 & (2/3)\tan\theta \\ (-2/3)\tan\theta & (-2/3)\tan\theta + 1 \end{bmatrix}$$

For  $\theta = 45^\circ$ :

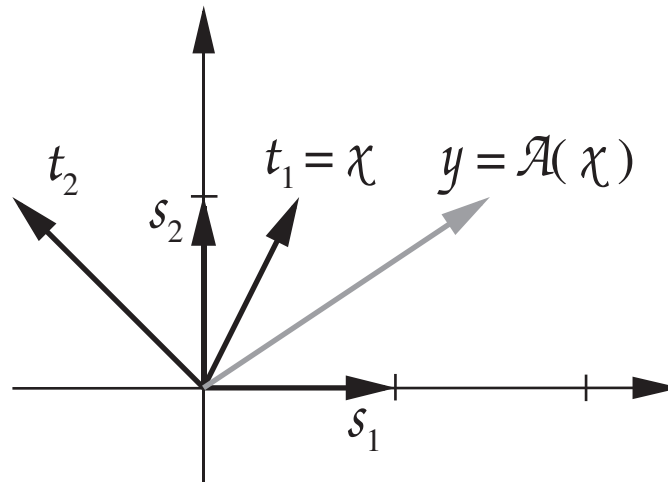
$$\mathbf{A}' = \begin{bmatrix} 5/3 & 2/3 \\ -2/3 & 1/3 \end{bmatrix} \quad \mathbf{A} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

## Example - (3)



Try a test vector:  $\mathbf{x} = \begin{bmatrix} 0.5 \\ 1 \end{bmatrix}$   $\mathbf{x}' = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$\mathbf{y} = \mathbf{A}\mathbf{x} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0.5 \\ 1 \end{bmatrix} = \begin{bmatrix} 1.5 \\ 1 \end{bmatrix} \quad \mathbf{y}' = \mathbf{A}'\mathbf{x}' = \begin{bmatrix} 5/3 & 2/3 \\ -2/3 & 1/3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 5/3 \\ -2/3 \end{bmatrix}$$



Check using reciprocal basis vectors:

$$\mathbf{y}' = \mathbf{B}^{-1}\mathbf{y} = \begin{bmatrix} 0.5 & -1 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1.5 \\ 1 \end{bmatrix} = \begin{bmatrix} 2/3 & 2/3 \\ -2/3 & 1/3 \end{bmatrix} \begin{bmatrix} 1.5 \\ 1 \end{bmatrix} = \begin{bmatrix} 5/3 \\ -2/3 \end{bmatrix}$$

## 6

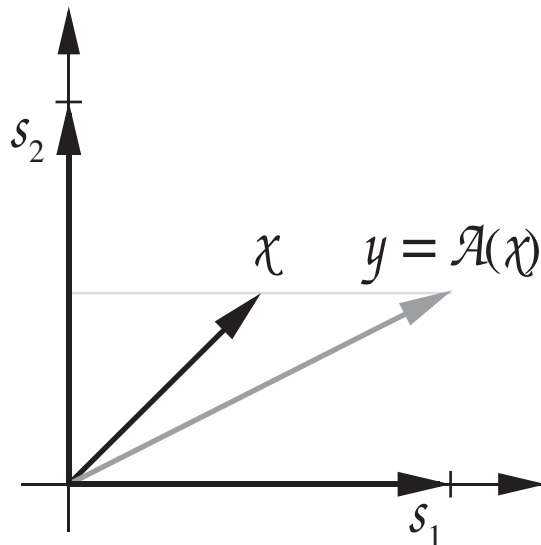
## Eigenvalues and Eigenvectors



Let  $\mathcal{A}:X\rightarrow X$  be a linear transformation. Those vectors  $z\in X$ , which are not equal to zero, and those scalars  $\lambda$  which satisfy

$$\mathcal{A}(z) = \lambda z$$

are called eigenvectors and eigenvalues, respectively.



Can you find an eigenvector for this transformation?



$$\mathbf{A}\mathbf{z} = \lambda\mathbf{z}$$

$$[\mathbf{A} - \lambda\mathbf{I}]\mathbf{z} = \mathbf{0} \quad \Rightarrow \quad |[\mathbf{A} - \lambda\mathbf{I}]| = 0$$

Skewing example ( $45^\circ$ ):

$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad \left| \begin{bmatrix} 1-\lambda & 1 \\ 0 & 1-\lambda \end{bmatrix} \right| = 0 \quad (1-\lambda)^2 = 0 \quad \begin{array}{l} \lambda_1 = 1 \\ \lambda_2 = 1 \end{array}$$

$$\begin{bmatrix} 1-\lambda & 1 \\ 0 & 1-\lambda \end{bmatrix} \mathbf{z} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \mathbf{z}_1 = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} z_{11} \\ z_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad z_{21} = 0 \quad \mathbf{z}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

For this transformation there is only one eigenvector.



Perform a change of basis (similarity transformation) using the eigenvectors as the basis vectors. If the eigenvalues are distinct, the new matrix will be diagonal.

$$\mathbf{B} = \begin{bmatrix} \mathbf{z}_1 & \mathbf{z}_2 & \dots & \mathbf{z}_n \end{bmatrix} \quad \begin{array}{l} \{\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_n\} \text{ Eigenvectors} \\ \{\lambda_1, \lambda_2, \dots, \lambda_n\} \text{ Eigenvalues} \end{array}$$

$$[\mathbf{B}^{-1} \mathbf{A} \mathbf{B}] = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$



$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1-\lambda & 1 \\ 1 & 1-\lambda \end{bmatrix} = 0 \quad \lambda^2 - 2\lambda = (\lambda)(\lambda - 2) = 0 \quad \begin{array}{l} \lambda_1 = 0 \\ \lambda_2 = 2 \end{array} \quad \begin{bmatrix} 1-\lambda & 1 \\ 1 & 1-\lambda \end{bmatrix} \mathbf{z} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\lambda_1 = 0 \Rightarrow \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \mathbf{z}_1 = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} z_{11} \\ z_{21} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad z_{21} = -z_{11} \quad \mathbf{z}_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\lambda_2 = 2 \Rightarrow \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \mathbf{z}_1 = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} z_{12} \\ z_{22} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad z_{22} = z_{12} \quad \mathbf{z}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\text{Diagonal Form:} \quad \mathbf{A}' = [\mathbf{B}^{-1} \mathbf{A} \mathbf{B}] = \begin{bmatrix} 1/2 & -1/2 \\ 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 2 \end{bmatrix}$$