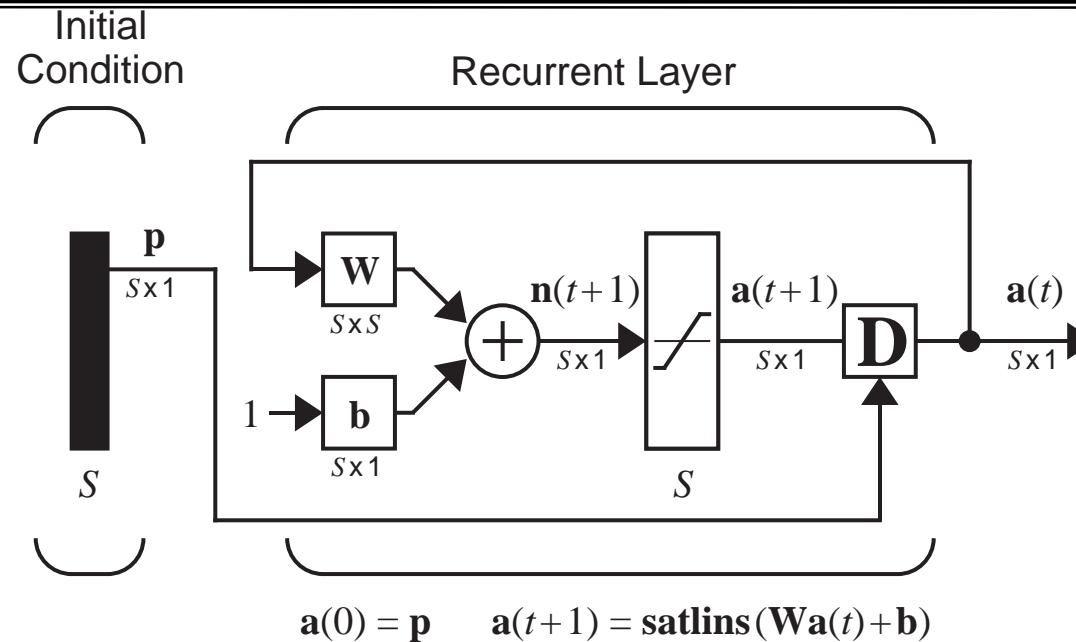




Linear Transformations

6

Hopfield Network Questions



- The network output is repeatedly multiplied by the weight matrix \mathbf{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 $X = \{x_i\}$, called the domain,
2. A set of elements $Y = \{y_i\}$, called the range, and
3. A rule relating each $x_i \in X$ to an element $y_i \in Y$.

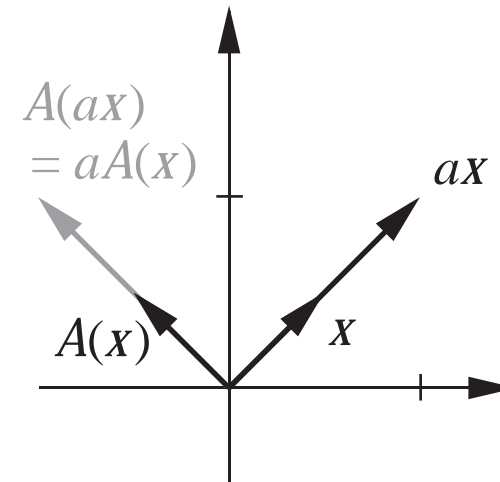
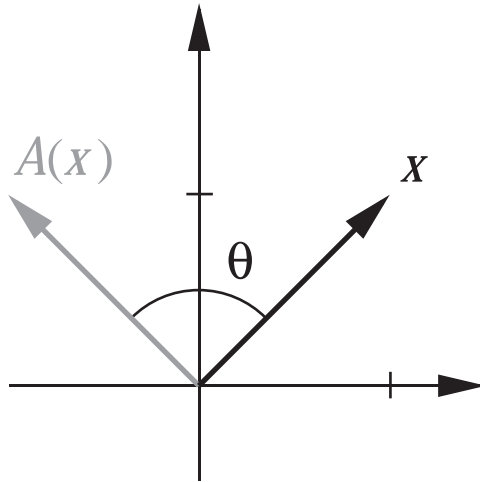
A transformation is **linear** if:

1. For all $x_1, x_2 \in X$, $A(x_1 + x_2) = A(x_1) + A(x_2)$,
2. For all $x \in X$, $a \in \mathfrak{R}$, $A(ax) = aA(x)$.

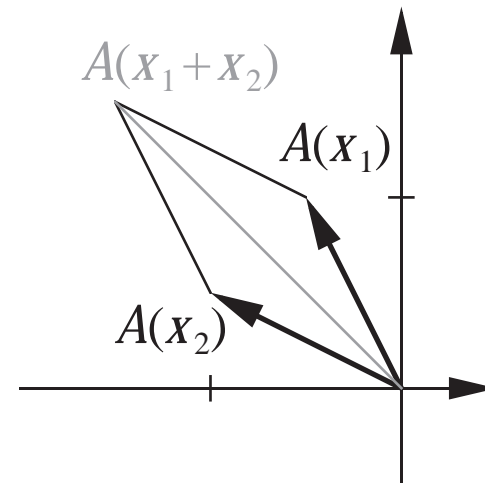
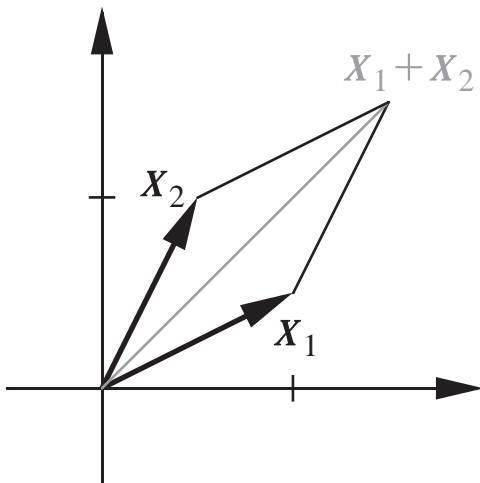


Is rotation linear?

1.



2.





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 .

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

Let $A: X \rightarrow Y$

$$A(x) = y$$

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



Since A is a linear operator,

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

Since the u_i are a basis for Y ,

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

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

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

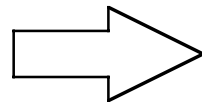


$$\sum_{i=1}^m u_i \sum_{j=1}^n a_{ij} x_j = \sum_{i=1}^m y_i u_i$$

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

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.

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

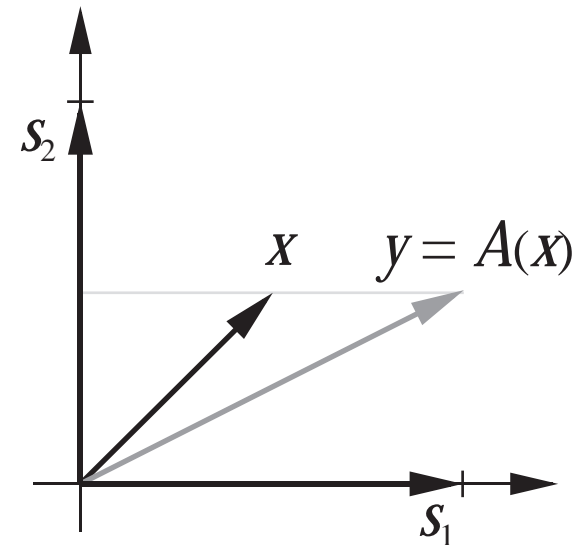
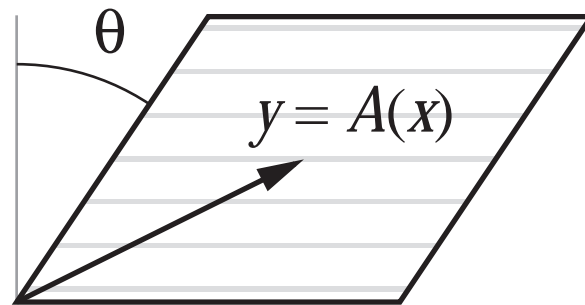
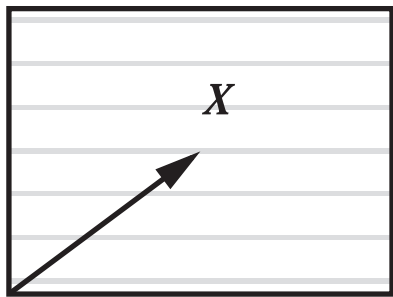
Each of these equations gives us one column of the matrix.

6

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 θ , 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.

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

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

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

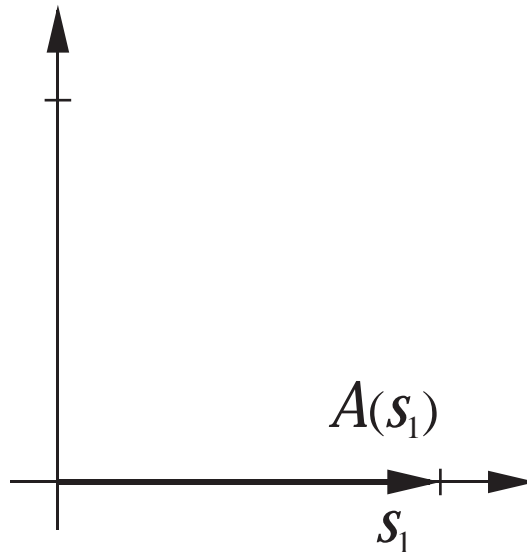
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Example - (3)



We begin with s_1 :

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



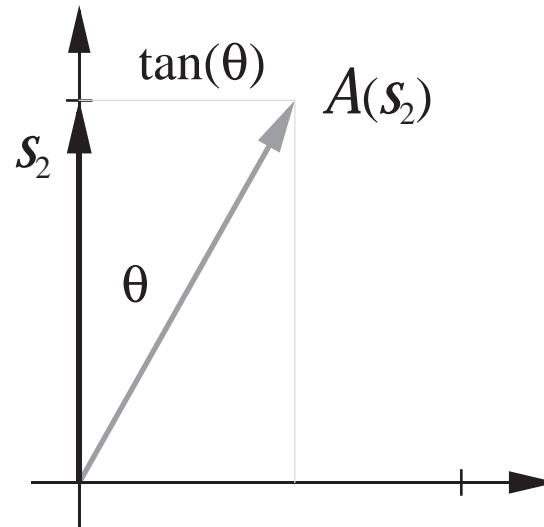
$$A(s_1) = 1s_1 + 0s_2 = \sum_{i=1}^2 a_{i1}s_i = a_{11}s_1 + a_{21}s_2$$

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 $A: X \rightarrow Y$. 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 .

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

$$A(x) = y$$

The matrix representation is:

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

$$Ax = y$$



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 .

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

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

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

6

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



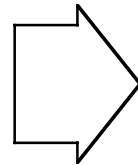
$$\mathbf{B}_t = [\mathbf{t}_1 \ \mathbf{t}_2 \ \dots \ \mathbf{t}_n] \quad \mathbf{x} = x'_1 \mathbf{t}_1 + x'_2 \mathbf{t}_2 + \dots + x'_n \mathbf{t}_n = \mathbf{B}_t \mathbf{x}'$$

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

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

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

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



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

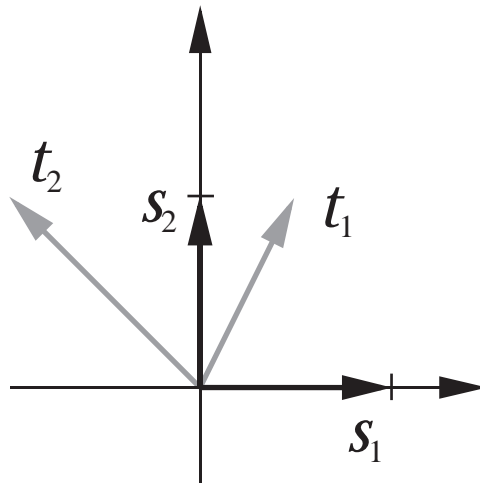
Similarity
Transform

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Example - (1)



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

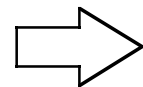


$$t_1 = 0.5s_1 + s_2$$

$$t_2 = -s_1 + s_2$$

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

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



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

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

(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 $\theta = 45^\circ$:

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

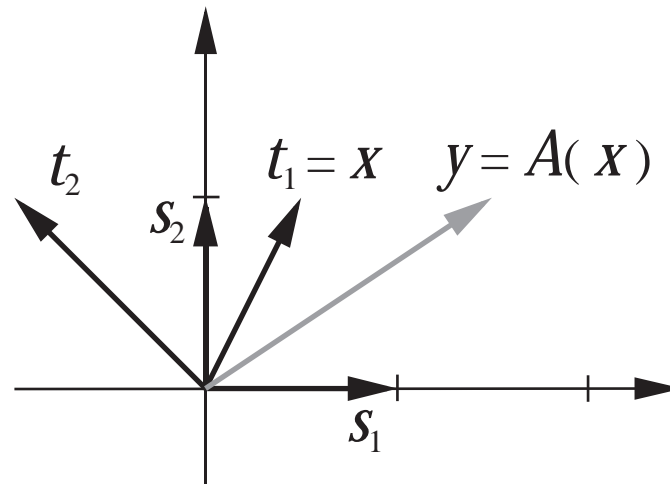
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Example - (3)



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

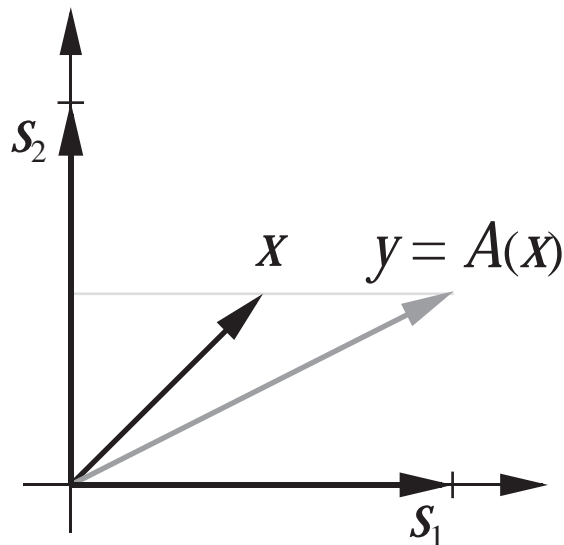
Eigenvalues and Eigenvectors



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

$$A(z) = \lambda z$$

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°):

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

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

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.

$$\mathbf{B} = \begin{bmatrix} \mathbf{z}_1 & \mathbf{z}_2 & \dots & \mathbf{z}_n \end{bmatrix}$$

$\{\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_n\}$ Eigenvectors

$\{\lambda_1, \lambda_2, \dots, \lambda_n\}$ Eigenvalues

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

6

Example



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

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

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

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

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