

بِسْمِ اللَّهِ الرَّحْمَنِ الرَّحِيمِ



# شبکه‌های عصبی مصنوعی

درس ۶

# تبدیل‌های خطی برای شبکه‌های عصبی

## Linear Transformations for Neural Networks

کاظم فولادی قلعه

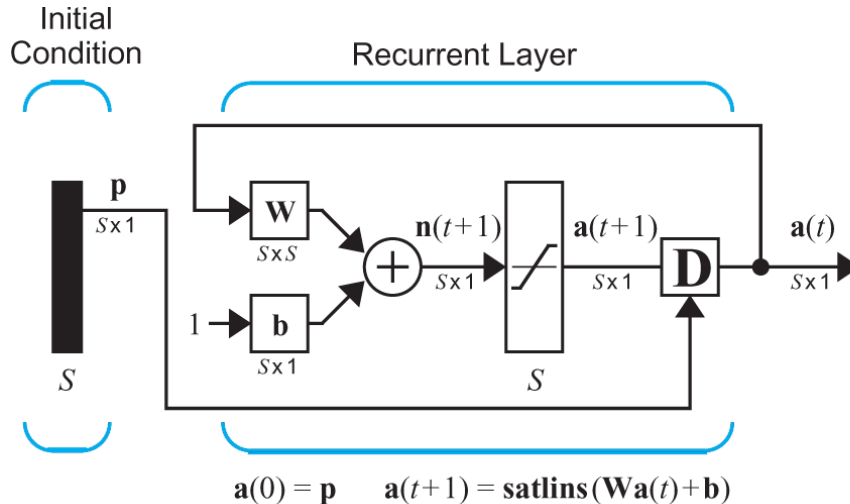
دانشکده مهندسی، پردیس فارابی

دانشگاه تهران

<http://courses.fouladi.ir/nn>



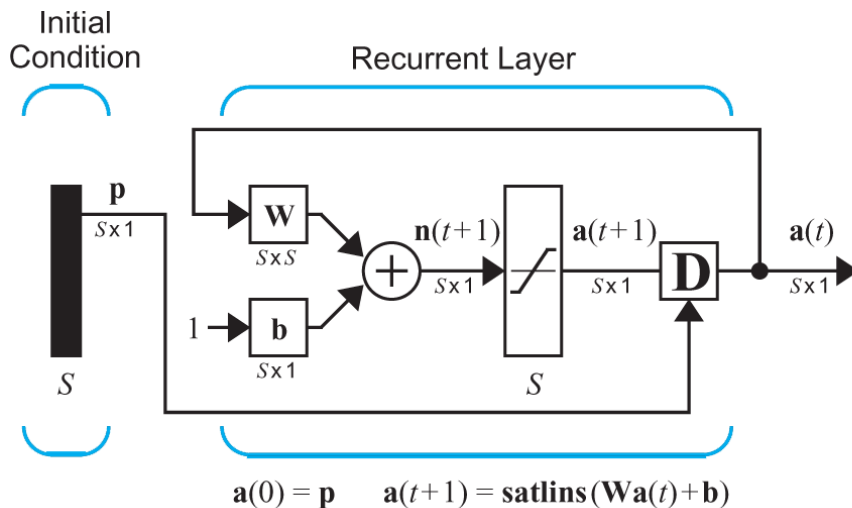
# Linear Transformations



- 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.

## پرسش‌هایی درباره‌ی شبکه‌ی هاپفیلد

ضرورت مطالعه‌ی تبدیل‌های خطی

HOPFIELD NETWORK QUESTIONS

○ خروجی شبکه به طور مکرر در ماتریس وزن  $\mathbf{W}$  ضرب می‌شود.

○ اثر این عمل تکراری چیست؟

○ آیا خروجی همگرا می‌شود؟ به سمت بی‌نهایت می‌رود؟ یا نوسان می‌کند؟

○ ضرب یک ماتریس وزن  $\mathbf{W}$  در بردار ورودی  $\mathbf{p}$  یکی از اعمال کلیدی است که توسط شبکه‌های عصبی انجام می‌شود.

○ این ضرب نمونه‌ای از یک تبدیل خطی است  $\Leftarrow$  نیاز به بررسی تبدیل خطی عمومی و تعیین مشخصات بنیادی آن.

تبدیل های خطی برای شبکه های عصبی

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# تبدیل های خطی



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  $x \in X$ ,  $a \in \mathfrak{R}$ ,  $A(ax) = aA(x)$ .

## تبدیل‌های خطی

LINEAR TRANSFORMATIONS

یک تبدیل سه جزء دارد:

دامنه

برد

ضابطه

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  $y_i \in Y$ .

خواص یک تبدیل خطی:

(۱) جمع‌پذیری

(۲) همگنی

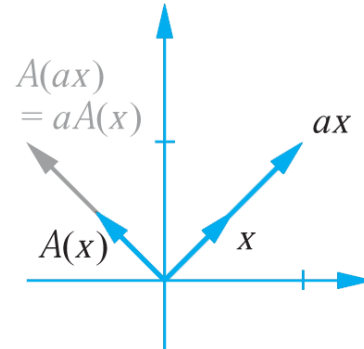
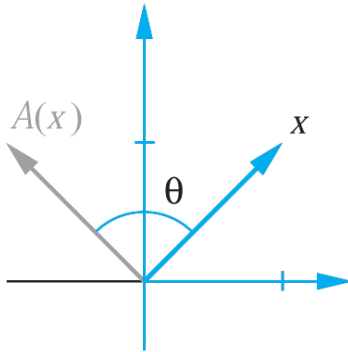
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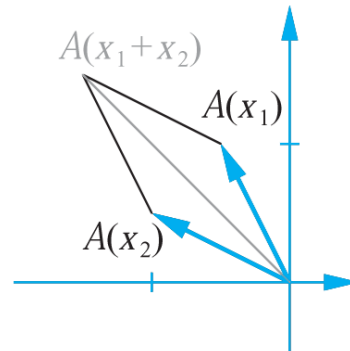
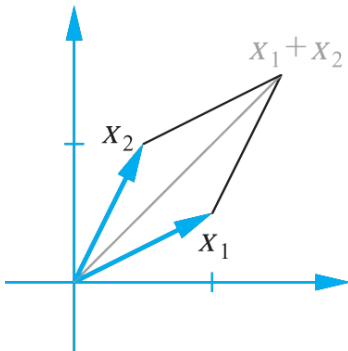


Is rotation linear?

1.



2.



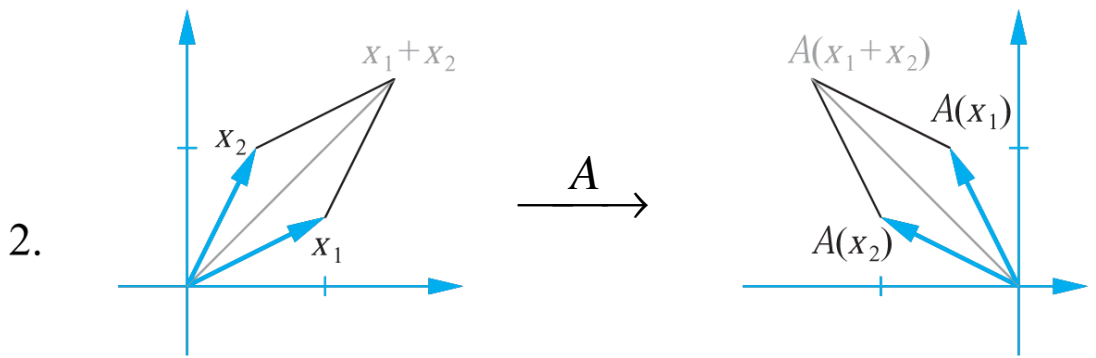
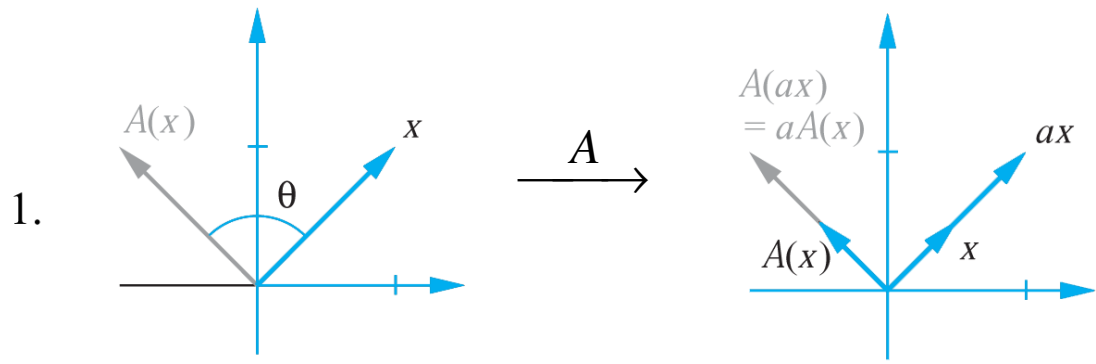


# تبدیل‌های خطی

مثال: تبدیل دوران

## LINEAR TRANSFORMATIONS

تبدیل دوران به اندازه‌ی زاویه‌ی  $\theta$  در  $\mathbb{R}^2$  یک تبدیل خطی است.



تبدیل‌های خطی برای شبکه‌های عصبی

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بازنمایی‌های  
ماتریسی  
برای  
تبدیل‌های  
خطی



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  $A: X \rightarrow Y$

$$A(x) = y$$

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

## بازنمایی ماتریسی

(۱ از ۳)

### MATRIX REPRESENTATION

هر تبدیل خطی بین دو فضای برداری متناهی-ابعاد می‌تواند توسط ضرب ماتریسی بازنمایی شود.

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.)

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

## بازنمایی ماتریسی

(۲ از ۳)

MATRIX REPRESENTATION

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$$A(v_j) = \sum_{i=1}^m a_{ij} u_i \quad (\text{The coefficients } a_{ij} \text{ will make up the matrix representation of the transformation.})$$

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

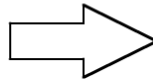


$$\sum_{i=1}^m u_i \sum_{j=1}^n a_{ij} x_j = \sum_{i=1}^m y_i u_i$$

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

Because the  $u_i$  are independent,

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



This is equivalent to  
matrix multiplication.

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

## بازنمایی ماتریسی

(۳ از ۳)

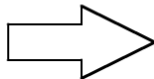
MATRIX REPRESENTATION

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

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- 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.

$$\mathbf{A}(\mathbf{v}_j) = \sum_{i=1}^m a_{ij} \mathbf{u}_i$$

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

## بازنمایی ماتریسی

خلاصه

### MATRIX REPRESENTATION

هر تبدیل خطی می‌تواند توسط ضرب ماتریسی بازنمایی شود.

برای یافتن ماتریس تبدیل باید هر یک از بردارهای پایه‌ی دامنه را تبدیل کنیم و سپس حاصل را بر حسب بردارهای پایه‌ی برد بسط بدهیم.

$$\mathbf{A}(\mathbf{v}_j) = \sum_{i=1}^m a_{ij} \mathbf{u}_i$$

هر معادله به فرم فوق، یک ستون از ماتریس را به ما می‌دهد.

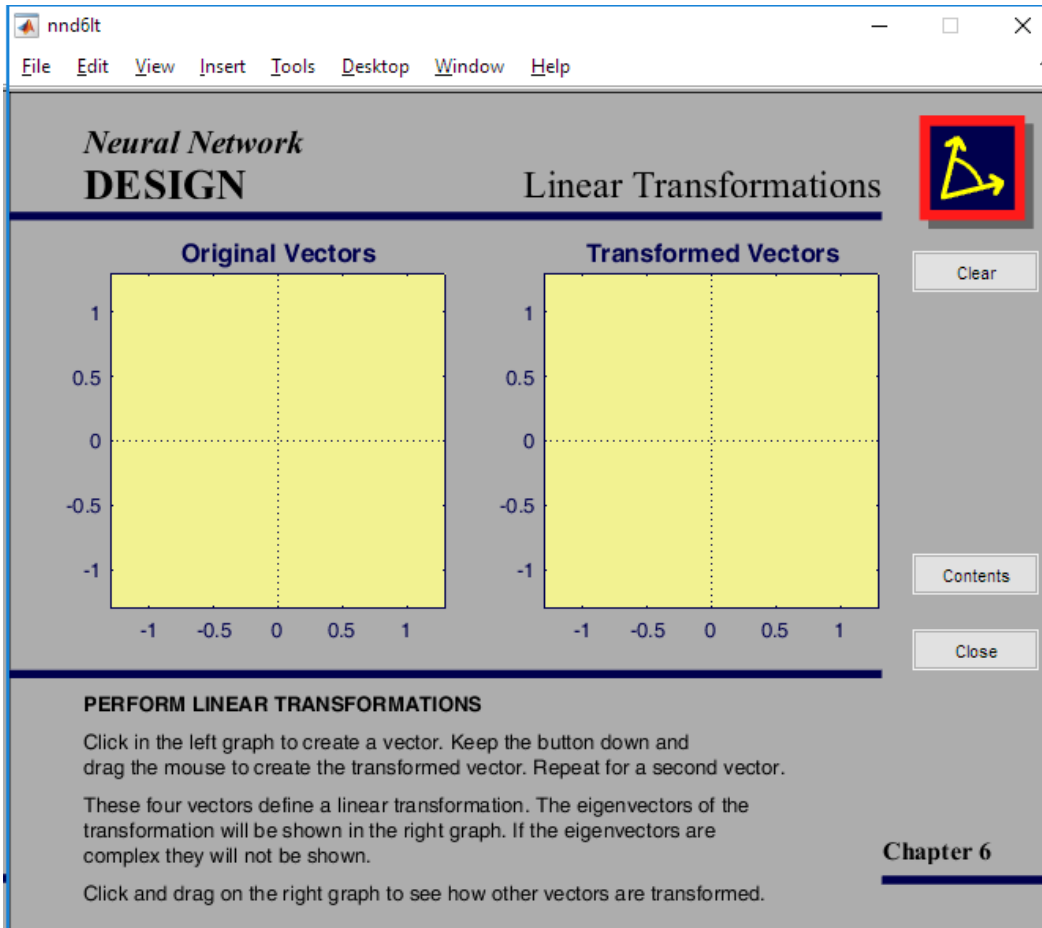
نکته: اگر مجموعه‌ی پایه برای دامنه یا برد را عوض کنیم، بازنمایی ماتریسی نیز تغییر می‌کند.

## بازنمایی ماتریسی

مثال: دوران

ROTATIONماتریس تبدیل برای دوران به اندازه‌ی  $\theta$ 

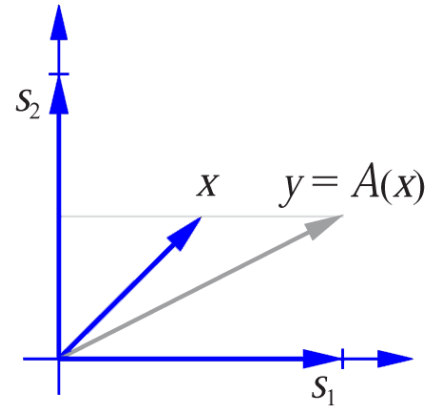
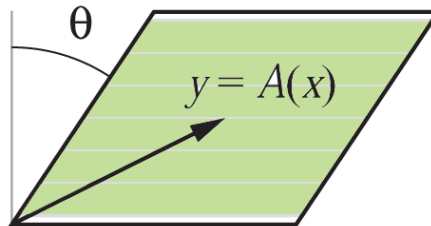
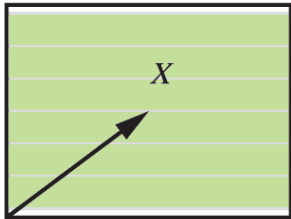
$$A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$



>> nnd61t



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?

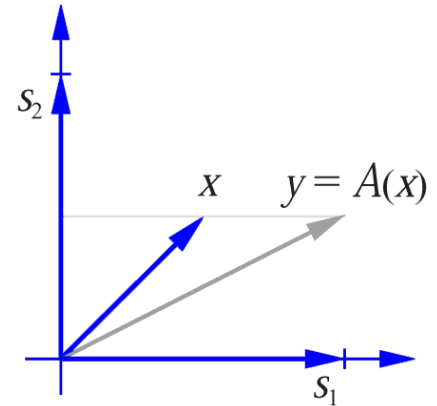
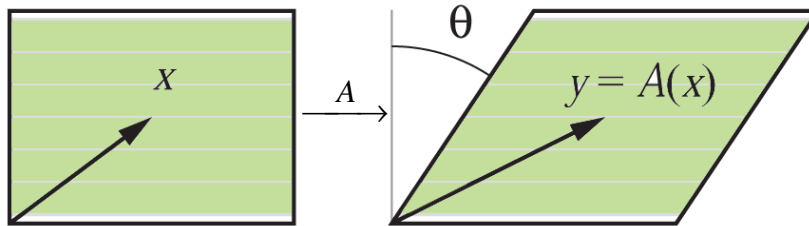


## بازنمایی ماتریسی

مثال: مایل‌سازی (۱ از ۵)

### SKEW

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?





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.

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

## بازنمایی ماتریسی

مثال: مایل‌سازی (۲ از ۵)

SKEW

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.

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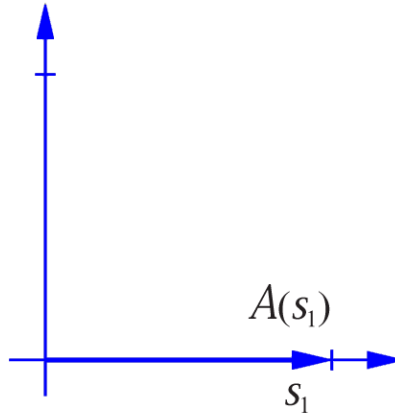


## Example - (3)



We begin with  $s_1$ :

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



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

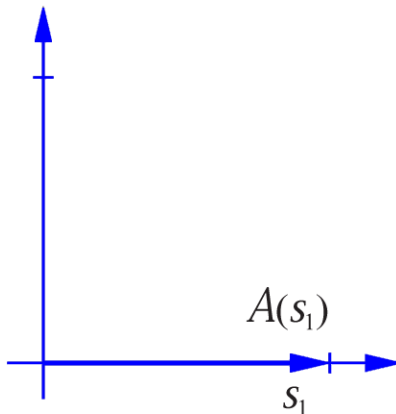
This gives us the first column of the matrix.

## بازنمایی ماتریسی

مثال: مایل‌سازی (۳ از ۵)

SKEWWe begin with  $s_1$ :

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



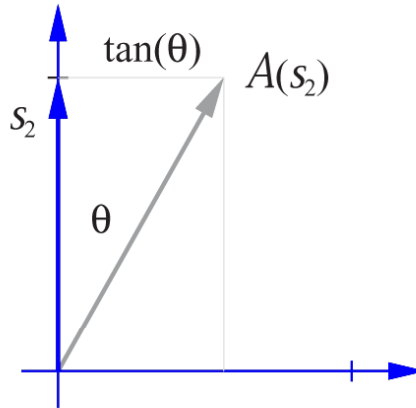
$$A(s_1) = 1s_1 + 0s_2 = \sum_{i=1}^2 a_{i1}s_i = a_{11}s_1 + a_{21}s_2$$

This gives us the first column of the matrix.

# Example - (4)



Next, we skew  $s_2$ :

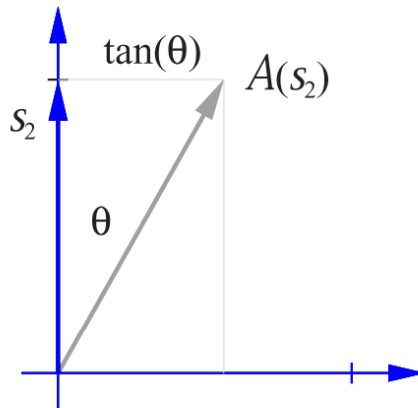


$$A(s_2) = \tan(\theta)s_1 + 1s_2 = \sum_{i=1}^2 a_{i2}s_i = a_{12}s_1 + a_{22}s_2$$

This gives us the second column of the matrix.

## بازنمایی ماتریسی

مثال: مایل‌سازی (۴ از ۵)

SKEWNext, 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:

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

## بازنمایی ماتریسی

مثال: مایل‌سازی (۵ از ۵)

SKEW

The matrix of the transformation is:

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

تبدیل‌های خطی برای شبکه‌های عصبی

۳

تغییر  
پایه‌ها



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:

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

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



## تغییر پایه‌ها

CHANGE OF BASIS

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  $Y$ .

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$$A(x) = y$$

The matrix representation is:

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

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



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:

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

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

## تغییر پایه‌ها

مجموعه‌های پایه‌ی جدید

NEW BASIS SETS

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:

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



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

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

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

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

## تغییر پایه‌ها

ارتباط ماتریس‌های تبدیل  $A$  و  $A'$  (۱ از ۲)NEW BASIS SETSExpand  $t_i$  in terms of the original basis vectors for  $X$ .

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

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

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

# How are $A$ and $A'$ related?



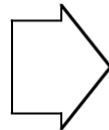
$$\mathbf{B}_t = [\mathbf{t}_1 \ \mathbf{t}_2 \ \dots \ \mathbf{t}_n] \quad \mathbf{x} = x'_1 \mathbf{t}_1 + x'_2 \mathbf{t}_2 + \dots + x'_n \mathbf{t}_n = \mathbf{B}_t \mathbf{x}'$$

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

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

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

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



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

Similarity  
Transform

## تغییر پایه‌ها

ارتباط ماتریس‌های تبدیل  $A$  و  $A'$  (۲ از ۲)NEW BASIS SETS

بردار پایه‌ی متقابل

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

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

هر ستون یک  $\mathbf{w}_i$ 

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

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

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

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

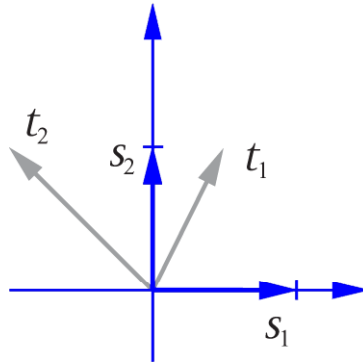
Similarity  
Transform

تبدیل شباهت

# Example - (1)



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

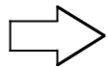


$$t_1 = 0.5s_1 + s_2$$

$$t_2 = -s_1 + s_2$$

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

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



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

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

(Same basis for domain and range.)

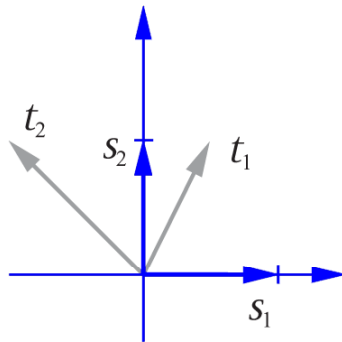


## تغییر پایه‌ها

مثال (۱ از ۳)

NEW BASIS SETS

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

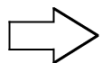


$$t_1 = 0.5s_1 + s_2$$

$$t_2 = -s_1 + s_2$$

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

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



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

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

(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  $\theta = 45^\circ$ :

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

## تغییر پایه‌ها

مثال (۲ از ۳)

NEW BASIS SETS

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

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

For  $\theta = 45^\circ$ :

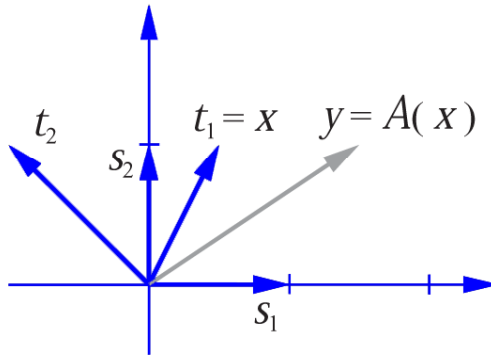
$$\mathbf{A}' = \begin{bmatrix} 5/3 & 2/3 \\ -2/3 & 1/3 \end{bmatrix} \quad \mathbf{A} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

# 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:

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

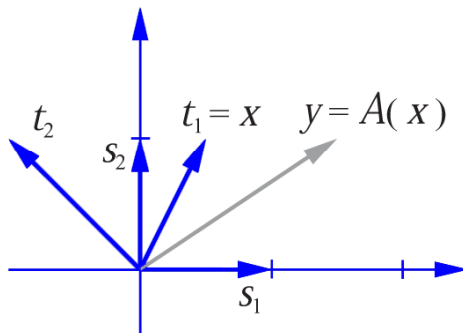
## تغییر پایه‌ها

مثال (۳ از ۳)

NEW BASIS SETS

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:

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

تبدیل‌های خطی برای شبکه‌های عصبی

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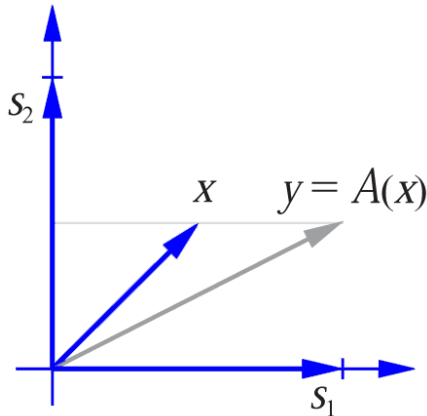
مقدارهای  
ویژه  
و  
بردارهای  
ویژه



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

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

are called eigenvectors and eigenvalues, respectively.



Can you find an eigenvector for this transformation?

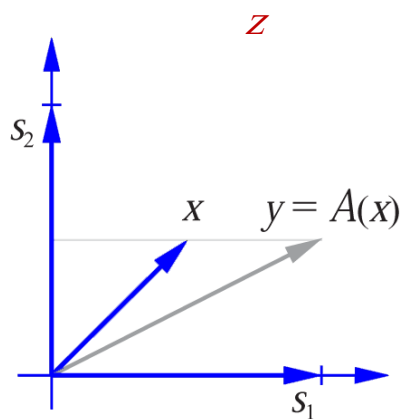
## مقدارهای ویژه و بردارهای ویژه

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  $\lambda$  which satisfy

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

are called eigenvectors and eigenvalues, respectively.



Can you find an eigenvector for this transformation?

 $z$  $\lambda$





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

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

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.

## مقدارهای ویژه و بردارهای ویژه

محاسبه‌ی مقادیر ویژه

COMPUTING THE EIGENVALUES

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

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

Skewing example (45°):

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

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

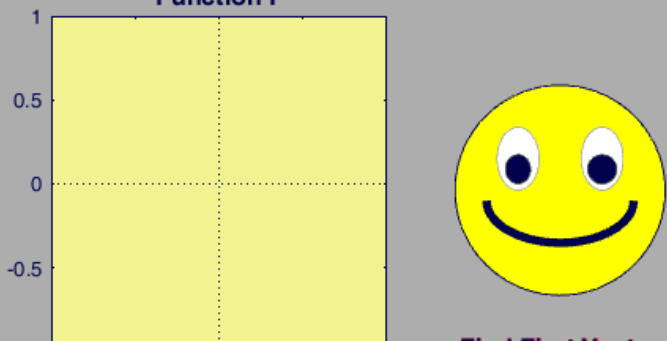
For this transformation there is only one eigenvector.

nnd6eg

File Edit View Insert Tools Desktop Window Help

## Neural Network DESIGN

## Eigenvector Game



Function F

Eigenpoints

New Game

Contents

Close

>Find First Vector<

**FINDING THE EIGENVECTORS**

Your job is to find the two eigenvectors of an unknown transformation

Click on the graph and hold the mouse button down. The vector you have chosen will appear in red. The result of transforming this vector will be black. Release the button and try to click so the red and black vectors point in the same (or exactly opposite) direction. When you find an eigenvector it will be shown in green. Continue searching for the other eigenvector.

You must find both eigenvectors in ten clicks. Happy Eigenhunting!

Chapter 6



>> nnd6eg

تبدیل‌های خطی برای شبکه‌های عصبی

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قطری سازی



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.

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

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

## قطری‌سازی

DIAGONALIZATION

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.

با استفاده از بردارهای ویژه به عنوان بردارهای پایه، تغییر پایه را انجام می‌دهیم (تبدیل شباهت).  
اگر مقادیر ویژه متمایز باشند  $\Leftarrow$  بردارهای ویژه مستقل خطی هستند  $\Leftarrow$  ماتریس جدید قطری خواهد بود.

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

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



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

$$\left| \begin{bmatrix} 1-\lambda & 1 \\ 1 & 1-\lambda \end{bmatrix} \right| = 0 \quad \lambda^2 - 2\lambda = (\lambda)(\lambda - 2) = 0$$

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

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

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

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

## قطری‌سازی

مثال

DIAGONALIZATION

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

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

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

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

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



تبدیل‌های خطی برای شبکه‌های عصبی

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منابع

## منبع اصلی



Martin T. Hagan, Howard B. Demuth, Mark H. Beale, Orlando De Jesus,  
**Neural Network Design,**  
 2<sup>nd</sup> Edition, Martin Hagan, 2014.

### Chapter 6

Online version can be downloaded from: <http://hagan.okstate.edu/nnd.html>

## 6 Linear Transformations for Neural Networks

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### Objectives

This chapter will continue the work of Chapter 5 in laying out the mathematical foundations for our analysis of neural networks. In Chapter 5 we reviewed vector spaces; in this chapter we investigate linear transformations as they apply to neural networks.

As we have seen in previous chapters, the multiplication of an input vector by a weight matrix is one of the key operations that is performed by neural networks. This operation is an example of a linear transformation. We want to investigate general linear transformations and determine their fundamental characteristics. The concepts covered in this chapter, such as eigenvalues, eigenvectors and change of basis, will be critical to our understanding of such key neural network topics as performance learning (including the Widrow-Hoff rule and backpropagation) and Hopfield network convergence.