

# Linear Algebra for Machine Learning in Python

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Introcution

Essential operations

Linear curve fitting

System dynamics, and dimensions

# Introcution

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TODO

$m, n$  is a matrix  $\mathbf{A}$

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}, a_{ij} \in \mathbb{R}. \quad (1)$$

# Essential operations

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TODO

# Multiplication

TODO



# Motivation of the determinant

TODO

## Computing determinants in two or three dimensions

The two dimensional case:

$$\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11} \cdot a_{22} - a_{12} \cdot a_{21} \quad (2)$$

(3)

Computing the determinant of a three dimensional matrix.

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11} \cdot \begin{vmatrix} a_{21} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \cdot \begin{vmatrix} a_{21} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} + a_{13} \cdot \begin{vmatrix} a_{21} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \quad (4)$$

# Linear Algebra for Machine Learning in Python

## └ Essential operations

## └ Computing determinants in two or three dimensions

The two dimensional case:

$$\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11} \cdot a_{22} - a_{12} \cdot a_{21} \quad (2)$$

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Computing the determinant of a three dimensional matrix:

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11} \cdot \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \cdot \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \cdot \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} \quad (4)$$

Draw the sign pattern on the board:

$$\begin{vmatrix} + & - & + & \dots \\ - & + & - & \dots \\ + & - & + & \dots \\ \vdots & \vdots & \vdots & \ddots \end{vmatrix} \quad (5)$$

The determinant can be expanded along any column as long as the sign pattern is respected.

# Determinants in n-dimensions

$$\begin{vmatrix} a_{11} & a_{21} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & \dots & a_{2n} \\ \vdots & & \vdots \\ a_{m2} & \dots & a_{mn} \end{vmatrix} + a_{21} \begin{vmatrix} a_{21} & \dots & a_{2n} \\ \vdots & & \vdots \\ a_{m2} & \dots & a_{mn} \end{vmatrix} \\
 - a_{m1} \begin{vmatrix} a_{11} & \dots & a_{1n} \\ a_{21} & \dots & a_{2n} \\ \vdots & & \vdots \end{vmatrix}$$

TODO

TODO

# Linear curve fitting

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**What is the best line connecting measurements?**



$$\mathbf{A}^\dagger = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \quad (6)$$

## Linear Algebra for Machine Learning in Python

## └ Linear curve fitting

## └ The Pseudoinverse

$$\mathbf{A}^{\dagger} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T$$

(6)

Sometimes solving  $\mathbf{Ax} + \mathbf{b} = 0$  is impossible. One the board, derive:

$$\min_x \frac{1}{2} |\mathbf{Ax} - \mathbf{b}|^2 \quad (7)$$

$$(8)$$

At the optimum we expect,

$$0 = \nabla_x \frac{1}{2} |\mathbf{Ax} - \mathbf{b}|^2 \quad (9)$$

$$= \nabla_x \frac{1}{2} (\mathbf{Ax} - \mathbf{b})^T (\mathbf{Ax} - \mathbf{b}) \quad (10)$$

$$= (\mathbf{Ax} - \mathbf{b}) \mathbf{A}^T \quad (11)$$

$$= \mathbf{A}^T \mathbf{Ax} - \mathbf{A}^T \mathbf{b} \quad (12)$$

$$\mathbf{A}^T \mathbf{b} = \mathbf{A}^T \mathbf{Ax} \quad (13)$$

$$(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b} = \mathbf{x} \quad (14)$$

# **System dynamics, and dimensions**

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TODO