

# CS5014 Machine Learning

## Lecture 2 Maths background review

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of  
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# So why this review session ?

Maths is essential ...

- for CS5014 (all other (data) *science* modules)
- we use a lot of maths
  - rigorous and concise way of communicating results
  - help us understand why *and* why not algorithms work

Review some key concepts

- not complete; just a small subset
- set the level of math maturity we expect

Self-assessment for yourself

- find the weak or rusty area
- do self studies

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# Mathematics for machine learning

## Linear algebra

- leap forward from scalar maths e.g.  $a + b = c$ 
  - $a, b$ , care scalars:  $1 + 2 = 3$
  - $a, b$ , care vectors:  $\begin{bmatrix} 1 \\ 2 \end{bmatrix} + \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}$

## Probability theory and statistics

- study of uncertainty: uncertainty is the norm
  - e.g. rain tomorrow? blood pressure measurement (reading error)?
- how to generalise your results
  - from one sample to the universe: vaccine trial

## Calculus

- study of continuous (real-valued) functions (using approximation *polynomial*)
  - $y = \sin(x)$  is well approximated by  $y = x$  when  $x \approx 0$
- useful when we do optimisation

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# Linear algebra: Basic concepts

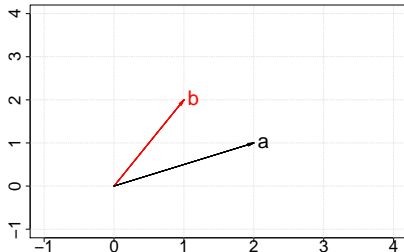
- vectors
- norms and distances
- linear independence
- matrices, linear transformation
- matrix arithmetics
- rank, determinant, trace



# Vector

A vector is a collection of  $n$  scalars

- $\mathbf{a} \in \mathbb{R}^n$ , default option is column vector i.e.  $n \times 1$
- represents a **displacement** from  $\mathbf{0}$  to  $\mathbf{a}$  in  $\mathbb{R}^n$
- e.g.  $\mathbf{a} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ ,  $\mathbf{b} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  (or  $\mathbf{a} = [2, 1]^T$  to save space)



# experiment