# Principles in Economics and Mathematics: the mathematical part

Bram De Rock

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### Practicalities about me

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# Practicalities about the course

- 12 hours on the mathematical part
- Micael Castanheira: 12 hours on the economics part
- Slides are available at MySBS and on http://mathecosolvay.com/spma/
- Schedule
  - Tuesday 17/9 and 24/9, 18.00-21.00, R42.2.107
  - Wednesday 18/9, 18.00-21.00, R42.2.103
  - Thursday 26/9, 18.00-21.00, R42.2.107
- Course evaluation
  - Written exam in the beginning of November to verify if you can apply the concepts discussed in class
  - Compulsory for students in Financial Markets
  - On a voluntary basis for students in *Quantitative Finance*

## Course objectives and content

- Refresh some useful concepts needed in your other coursework
  - No thorough or coherent study
  - Interested student: see references for relevant material
- Content:
  - Calculus (derivatives, optimization, concavity)
  - Linear algebra (solving system of linear equations, matrices, linear (in)dependence)
  - Fundamentals on probability (probability and cumulative distributions, expectations of a random variable, correlation)

## References

- Chiang, A.C. and K. Wainwright, "Fundamental Methods of Mathematical Economics", Economic series, McGraw-Hill.
- Green, W.H., "*Econometric Analysis, Seventh Edition*", Pearson Education limited.
- Luderer, B., V. Nollau and K. Vetters, "Mathematical Formulas for Economists", Springer, New York. ULB-link
- Simon, C.P. and L. Blume "*Mathematics for Economists*", Norton & Company, New York.
- Sydsaeter, K., A. Strom and P. Berck, "Economists' Mathematical Manual", Springer, New York. ULB-link

Motivation Functions of one variable Functions of more than one variable Optimization

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# Outline



#### Introduction

#### 2 Calculus

- Motivation
- Functions of one variable
- Functions of more than one variable
- Optimization

#### 3 Linear algebra

4 Fundamentals of probability theory

Motivation Functions of one variable Functions of more than one variable Optimization

# Role of functions

- Calculus = "the study of functions"
- Functions allow to exploit mathematical tools in Economics
- E.g. make consumption decisions
  - max  $U(x_1, x_2)$  s.t.  $p_1 x_1 + p_2 x_2 = Y$
  - Characterization:  $x_1 = f(p_1, p_1, Y)$
  - Econometrics: estimate f
  - Allows to model/predict consumption behavior
- Warning about identification
  - Causality: what is driving what?
  - Functional structure: what is driving the result?
  - Does the model allow to identify

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# Derivatives

- Marginal changes are important in Economics
  - The impact of a infinitesimally small change of one of the variables
  - Comparative statistics: what is the impact of a price change?
  - Optimization: what is the optimal consumption bundle?
- Marginal changes are mostly studied by taking derivatives
- Characterizing the impact depends on the function
  - $f: D \subseteq \mathbb{R}^n \to \mathbb{R}^k : (x_1, \ldots, x_n) \mapsto (y_1, \ldots, y_k) = f(x_1, \ldots, x_n)$
  - We will always take k = 1
  - First look at n = 1 and then generalize
  - Note:  $\mathbb{N} \subset \mathbb{Z} \subset \mathbb{Q} \subset \mathbb{R} \subset \mathbb{C}$

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Motivation Functions of one variable Functions of more than one variable Optimization

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# Outline



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Motivation Functions of one variable Functions of more than one variable Optimization

Functions of one variable:  $f : D \subseteq \mathbb{R} \to \mathbb{R}$ 

$$\frac{df}{dx} = f' = \lim_{\Delta x \to 0} \frac{f(x + \Delta x) - f(x)}{\Delta x}$$

- Limit of quotient of differences
- If it exists, then it is called the derivative
- f' is again a function
- E.g.  $f(x) = 3x^2 4$
- E.g. discontinuous functions, border of domain, f(x) = |x|

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## Some important derivatives and rules

Let us abstract from specifying the domain D and assume that  $c, n \in \mathbb{R}_0$ 

• If 
$$f(x) = c$$
, then  $f'(x) = 0$ 

• If 
$$f(x) = cx^n$$
, then  $f'(x) = ncx^{n-1}$ 

• If 
$$f(x) = ce^x$$
, then  $f'(x) = ce^x$ 

• If 
$$f(x) = c \ln(x)$$
, then  $f(x) = c \frac{1}{x}$ 

• 
$$(f(x) \pm g(x))' = f'(x) \pm g'(x)$$

• 
$$(f(x)g(x))' = f'(x)g(x) + f(x)g'(x) \neq f'(x)g'(x)$$

• 
$$(\frac{f(x)}{g(x)})' = \frac{f'(x)g(x) - f(x)g'(x)}{g(x)^2}$$

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# Application: the link with marginal changes

- By definition it is the limit of changes
- Slope of the tangent line
  - Increasing or decreasing function (and thus impact)
  - Does the inverse function exist?
- First order approximation in some point c
  - Based on expression for the tangent line in c
  - $f(c + \Delta x) \approx f(c) + f'(c)(\Delta x)$
  - More general approximation: Taylor expansion

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# Application: elasticities

• The elasticity of f in x:  $\frac{f'(x)x}{f(x)}$ 

- The limit of the quotient of changes in terms of percentage
  - Percentage change of the function:  $\frac{f(x+\Delta x)-f(x)}{f(x)}$
  - Percentage change of the variable:  $\frac{\Delta x}{x}$

• Quotient: 
$$\frac{f(x+\Delta x)-f(x)}{\Delta x} \frac{x}{f(x)}$$

- Is a unit independent informative number
  - E.g. the (price) elasticity of demand

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# Application: comparative statics for a simple market model

- Demand: *Q* = 10 − 4*P*
- Supply:  $Q = 2 + \alpha P$

• 
$$P^* = \frac{8}{4+\alpha}$$
 and  $Q^* = \frac{8+10\alpha}{4+\alpha}$ 

• 
$$\frac{dP^*}{d\alpha} = \frac{-8}{(4+\alpha)^2}$$
 and  $\frac{dQ^*}{d\alpha} = \frac{32}{(4+\alpha)^2}$ 

• The elasticity of demand is  $\frac{-4P}{10-4P}$ 

Introduction Motivation Calculus Functions of one variable Linear algebra Functions of more than one variable Fundamentals of probability theory Optimization

## Some exercises

- Compute the derivative of the following functions (defined on  $\mathbb{R}^+)$ 
  - $f(x) = 17x^2 + 5x + 7$
  - $f(x) = -\sqrt{x} + 3$
  - $f(x) = \frac{1}{x^2}$
  - $f(x) = 17x^2e^x$
  - $f(x) = \frac{x \ln(x)}{x^2 4}$
- Let  $f(x) : \mathbb{R} \to \mathbb{R} : x \mapsto x^2 + 5x$ .
  - Determine on which region f is increasing
  - Is f invertible?
  - Approximate f in 1 and derive an expression for the approximation error
  - Compute the elasticity in 3 and 5

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# The chain rule

Often we have to combine functions

• If z = f(y) and y = g(x), then z = h(x) = f(g(x))

- We have to be careful with the derivative
- A small change in x causes a chain reaction
  - It changes y and this in turn changes z
- That is why  $\frac{dz}{dx} = \frac{dz}{dy}\frac{dy}{dx} = f'(y)g'(x)$ 
  - Can easily be generalized to compositions of more than two functions

• 
$$\frac{dz}{dx} = \frac{dz}{dy}\frac{dy}{du}\cdots\frac{dy}{dx}$$

• E.g. if 
$$h(x) = e^{x^2}$$
, then  $h'(x) = e^{x^2} 2x$ 

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Motivation Functions of one variable Functions of more than one variable Optimization

# Higher order derivatives

- The derivative is again a function of which we can take derivatives
- Higher order derivatives describe the changes of the changes
- Notation
  - f''(x) or more generally  $f^{(n)}(x)$
  - $\frac{d}{dx}(\frac{df}{dx})$  or more generally  $\frac{d^n}{dx^n}f(x)$
- E.g. if  $f(x) = 5x^3 + 2x$ , then  $f'''(x) = f^{(3)}(x) = 30$

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#### Application: concave and convex functions

$$f: D \subset \mathbb{R}^n \to \mathbb{R}$$

#### • f is concave

- $\forall x, y \in D, \forall \lambda \in [0, 1] : f(\lambda x + (1 \lambda)y) \ge \lambda f(x) + (1 \lambda)f(y)$
- If  $n = 1, \forall x \in D : f''(x) \le 0$
- f is convex
  - $\forall x, y \in D, \forall \lambda \in [0, 1] : f(\lambda x + (1 \lambda)y) \le \lambda f(x) + (1 \lambda)f(y)$
  - If  $n = 1, \forall x \in D : f''(x) \ge 0$

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### Application: concave and convex functions

- Very popular and convenient assumptions in Economics
  - E.g. optimization
- Sometimes intuitive interpretation
  - E.g. risk-neutral, -loving, -averse
- Don't be confused with a convex set
  - S is a set  $\Leftrightarrow \forall x, y \in S, \forall \lambda \in [0, 1] : \lambda x + (1 \lambda)y \in S$

Motivation Functions of one variable Functions of more than one variable Optimization

## Some exercises

 Compute the first and second order derivative of the following functions (defined on ℝ<sup>+</sup>)

• 
$$f(x) = -\pi$$

• 
$$f(x) = -\sqrt{5x} + 3$$

• 
$$f(x) = e^{-3x}$$

• 
$$f(x) = \ln(5x)$$

• 
$$f(x) = x^3 - 6x^2 + 17$$

Determine which of these functions are concave or convex

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Motivation Functions of one variable Functions of more than one variable Optimization

# Outline



#### 2 Calculus

- Motivation
- Functions of one variable
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#### 3 Linear algebra

#### 4 Fundamentals of probability theory

Functions of more than one variable Functions of more than one variable:  $f: D \subseteq \mathbb{R}^n \to \mathbb{R}$ 

- Same applications in mind but now several variables
  - E.g. what is the marginal impact of changing  $x_1$ , while controlling for other variables?
- Look at the partial impact: partial derivatives

• 
$$\frac{\partial}{\partial x_i} f(x_1, \dots, x_n) = f_{x_i} = \lim_{\Delta x_i \to 0} \frac{f(x_1, \dots, x_i + \Delta x_i, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{\Delta x_i}$$

 Same interpretation as before, but now fixing remaining variables

• E.g. 
$$f(x_1, x_2, x_3) = 2x_1^2x_2 - 5x_3$$
  
•  $\frac{\partial}{\partial x_1} f(x_1, x_2, x_3) = 4x_1x_2$   
•  $\frac{\partial}{\partial x_2} f(x_1, x_2, x_3) = 2x_1^2$   
•  $\frac{\partial}{\partial x_3} f(x_1, x_2, x_3) = -5$ 

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## Partial derivative

- Geometric interpretation: slope of tangent line in the x<sub>i</sub> direction
- Same rules hold
- Higher order derivatives

• 
$$\frac{\partial^2}{\partial x_i^2} f(x_1, \dots, x_n)$$
  
• 
$$\frac{\partial^2}{\partial x_i x_j} f(x_1, \dots, x_n) = \frac{\partial^2}{\partial x_j x_i} f(x_1, \dots, x_n)$$
  
• E.g. 
$$\frac{\partial^2}{\partial x_i^2} f(x_1, x_2, x_3) = 4x_2 \text{ and } \frac{\partial^2}{\partial x_1 x_3} f(x_1, x_2, x_3) = 0$$

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#### Some remarks

• Gradient: 
$$\nabla f(x_1, \ldots, x_n) = (\frac{\partial f}{\partial x_1}, \ldots, \frac{\partial f}{\partial x_n})$$

Chain rule: special case

• 
$$x_1 = g_1(t), \dots, x_n = g_n(t)$$
 and  $f(x_1, \dots, x_n)$   
•  $h(t) = f(x_1, \dots, x_n) = f(g_1(t), \dots, g_n(t))$   
•  $\frac{dh(t)}{dt} = h'(t) = \frac{\partial f(x_1, \dots, x_n)}{\partial x_1} \frac{dx_1}{dt} + \dots + \frac{\partial f(x_1, \dots, x_n)}{\partial x_n} \frac{dx_n}{dt}$   
• E.g.  $f(x_1, x_2) = x_1 x_2$ ,  $g_1(t) = e^t$  and  $g_2(t) = t^2$   
•  $h'(t) = e^t t^2 + e^t 2t$ 

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## Some remarks

- Slope of indifference curve of *f*(*x*<sub>1</sub>, *x*<sub>2</sub>)
  - Indifference curve: all  $(x_1, x_2)$  for which  $f(x_1, x_2) = C$  (with *C* some give number)
  - Implicit function theorem:  $f(x_1, g(x_1)) = C$

• 
$$\frac{\partial}{\partial x_1}f(x_1, x_2) + \frac{\partial}{\partial x_2}f(x_1, x_2)\frac{dg}{dx_1} = 0$$

• Slope = 
$$-\frac{\frac{\partial}{\partial x_1}f(x_1, x_2)}{\frac{\partial}{\partial x_2}f(x_1, x_2)}$$

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### Some exercises

 Compute the gradient and all second order partial derivatives for the following functions (defined on ℝ<sup>+</sup>)

• 
$$f(x_1, x_2) = x_1^2 - 2x_1x_2 + 3x_2^2$$

• 
$$f(x_1, x_2) = \ln(x_1 x_2)$$

• 
$$f(x_1, x_2, x_3) = e^{x_1 + 2x_2} - 3x_1x_3$$

• Compute the marginal rate of substitution for the utility function  $U(x_1, x_2) = x_1^{\alpha} x_2^{\beta}$ 

Motivation Functions of one variable Functions of more than one variable **Optimization** 

# Outline



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Motivation Functions of one variable Functions of more than one variable **Optimization** 

## Optimization: important use of derivatives

- Many models in economics entail optimizing behavior
  - Maximize/Minimize objective subject to constraints
- Characterize the points that solve these models
- Note on Mathematics vs Economics
  - Profit = Revenue Cost
  - Marginal revenue = marginal cost
  - Marginal profit = zero

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# Optimization: formal problem

 $\max / \min f(x_1, \dots, x_n)$ s.t.  $g_1(x_1, \dots, x_n) = c_1$  $\dots$  $g_m(x_1, \dots, x_n) = c_m$  $x_1, \dots, x_n \ge 0$ 

- Inequality constraints are also possible
- Kuhn-Tucker conditions

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# Necessity and sufficiency

- Necessary conditions based on first order derivatives
  - Local candidate for an optimum
- Sufficient conditions based on second order derivatives
- Necessary condition is sufficient if
  - The constraints are convex functions
  - E.g. no constraints, linear constraints, ...
  - The objective function is concave: global maximum is obtained
  - The objective function is convex: global minimum is obtained
  - Often the "real" motivation in Economics

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### **Necessary conditions**

#### Free optimization

No constraints

• 
$$f'(x^*) = 0$$
 if  $n = 1$ 

• 
$$\frac{\partial}{\partial x_i} f(x_1^*, \dots, x_n^*) = 0$$
 for  $i = 1, \dots, n$ 

Intuitive given our geometric interpretation

Motivation Functions of one variable Functions of more than one variable **Optimization** 

### **Necessary conditions**

#### Optimization with positivity constraints

- No g<sub>i</sub> constraints
- On the boundary extra optima are possible
- Often ignored: interior solutions

• 
$$x_i^* \ge 0$$
 for  $i = 1, ..., n$ 

• 
$$x_i^* \frac{\partial}{\partial x_i} f(x_1^*, \dots, x_n^*) = 0$$
 for  $i = 1, \dots, n$ 

•  $\frac{\partial}{\partial x_i} f(x_1^*, \dots, x_n^*) \leq 0$  for all  $i = 1, \dots, n$  simultaneously OR  $\frac{\partial}{\partial x_i} f(x_1^*, \dots, x_n^*) \geq 0$  for all  $i = 1, \dots, n$  simultaneously

Optimization

### Necessary conditions

- Constrained optimization without positivity constraints
  - Define Lagrangian:  $L(x_1, \ldots, x_n, \lambda_1, \ldots, \lambda_m) =$  $f(x_1, \dots, x_n) - \lambda_1(g_1(x_1, \dots, x_n)) - \dots - \lambda_m(g_m(x_1, \dots, x_n))$ •  $\frac{\partial}{\partial x_i} L(x_1^*, \dots, x_n^*, \lambda_1^*, \dots, \lambda_m^*) = 0$  for all  $i = 1, \dots, n$

  - $\frac{\partial}{\partial \lambda_i} L(x_1^*, \dots, x_n^*, \lambda_1^*, \dots, \lambda_m^*) = 0$  for all  $j = 1, \dots, m$
  - Alternatively:  $\nabla f(x_1^*, \ldots, x_n^*) =$  $\lambda_1^* \nabla g_1(x_1^*,\ldots,x_n^*) + \cdots + \lambda_m^* \nabla g_m(x_1^*,\ldots,x_n^*)$
  - Some intuition: geometric interpretation
  - Lagrange multiplier = shadow price

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Motivation Functions of one variable Functions of more than one variable Optimization

# Application: utility maximization

$$\max U(x_1, x_2) = x_1^{\alpha} x_2^{1-\alpha} \quad s.t. \quad p_1 x_1 + p_2 x_2 = Y$$

• 
$$L(x_1, x_2, \lambda_1) = x_1^{\alpha} x_2^{1-\alpha} - \lambda_1 (p_1 x_1 + p_2 x_2 - Y)$$
  
•  $\frac{\partial L}{\partial x_1} = \alpha x_1^{\alpha-1} x_2^{1-\alpha} - \lambda_1 p_1 = 0$   
•  $\frac{\partial L}{\partial x_2} = (1-\alpha) x_1^{\alpha} x_2^{-\alpha} - \lambda_1 p_2 = 0$   
•  $\frac{\partial L}{\partial \lambda_1} = p_1 x_1 + p_2 x_2 - Y = 0$   
•  $x_1^* = \frac{\alpha Y}{p_1}, x_2^* = \frac{(1-\alpha)Y}{p_2} \text{ and } \lambda_1^* = (\frac{\alpha}{p_1})^{\alpha} (\frac{(1-\alpha)}{p_2})^{(1-\alpha)}$ 

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Motivation Functions of one variable Functions of more than one variable **Optimization** 

#### Some exercises

- Find the optima for the following problems
  - $\max / \min x^3 12x^2 + 36x + 8$
  - max / min  $x_1^3 x_2^3 + 9x_1x_2$
  - min  $2x_1^2 + x_1x_2 + 4x_2^2 + x_1x_3 + x_3^2 15x_1$
  - max  $x_1x_2$  s.t.  $x_1 + 4x_2 = 16$
  - max yz + xz s.t.  $y^2 + z^2 = 1$  and xz = 3
- Add positivity constraints to the above unconstrained problems and do the same

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Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Outline



#### Introduction

- 2 Calculus
- 3 Linear algebra
  - Motivation
  - Matrix algebra
  - The link with vector spaces
  - Application: solving a system of linear equations

4 Fundamentals of probability theory

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# Motivation

- Matrices allow to formalize notation
- Useful in solving system of linear equations
- Useful in deriving estimators in econometrics
- Allows us to make the link with vector spaces

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

• • • • • • • • • • • •

# Outline



- 2 Calculus
- 3 Linear algebra
  - Motivation

#### Matrix algebra

- The link with vector spaces
- Application: solving a system of linear equations

#### 4 Fundamentals of probability theory

Introduction Motivation Calculus Matrix algebra Linear algebra The link with vector spaces Fundamentals of probability theory Application: solving a system of linear equations

# Matrices

$$A = (a_{ij})_{i=1,...,n;j=1,...,m} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1m} \\ a_{21} & a_{22} & \cdots & a_{2m} \\ \vdots & \vdots & \cdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nm} \end{pmatrix}$$

• 
$$a_{ij} \in \mathbb{R}$$
 and  $A \in \mathbb{R}^{n \times m}$ 

- *n* rows and *m* columns
- Square matrix if *n* = *m*
- Notable square matrices
  - Symmetric matrix:  $a_{ij} = a_{ji}$  for all i, j = 1, ..., n
  - Diagonal matrix:  $a_{ij} = 0$  for all i, j = 1, ..., n and  $i \neq j$
  - Triangular matrix: only non-zero elements above (or below) the diagonal

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Matrix manipulations

Let  $A, B \in \mathbb{R}^{n \times m}$  and  $k \in \mathbb{R}$ 

- Equality:  $A = B \Leftrightarrow a_{ij} = b_{ij}$  for all  $i, j = 1, \dots, n$
- Scalar multiplication: kA = (ka<sub>ij</sub>)<sub>i=1,...,n;j=1,...,m</sub>
- Addition:  $A \pm B = (a_{ij} \pm b_{ij})_{i=1,...,n;j=1,...,m}$ 
  - Dimensions must be equal
- Transposition:  $A' = A^t = (a_{ji})_{j=1,...,m;i=1,...,n}$

• 
$$A \in \mathbb{R}^{n \times m}$$
 and  $A^t \in \mathbb{R}^{m \times m}$ 

• 
$$(A \pm B)^t = A^t \pm B^t$$

• 
$$(kA)^t = kA^t$$

• 
$$(A^t)^t = A$$

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# Matrix multiplication

Let  $A \in \mathbb{R}^{n \times m}$  and  $B \in \mathbb{R}^{m \times k}$ 

- $AB = (\sum_{h=1}^{m} a_{ih} b_{hj})_{i=1,...,n;j=1,...,k}$
- Multiply the row vector of A with the column vector of B
  - Aside: scalar/inner product and norm of vectors
  - Orthogonal vectors
- Number of columns of *A* must be equal to number of rows of *B*
- $AB \neq BA$ , even if both are square matrices

• 
$$(AB)^t = B^t A^t$$

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Example

Let 
$$A = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}$$
,  $B = \begin{pmatrix} 1 & -1 \\ 0 & 2 \end{pmatrix}$  and  $C = \begin{pmatrix} 1 & 2 & 3 \\ 3 & 2 & 1 \end{pmatrix}$   
•  $A^t = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}$ ,  $B^t \begin{pmatrix} 1 & 0 \\ -1 & 2 \end{pmatrix}$  and  $C = \begin{pmatrix} 1 & 3 \\ 2 & 2 \\ 3 & 1 \end{pmatrix}$   
•  $AB = \begin{pmatrix} 2 & 0 \\ 1 & 1 \end{pmatrix}$  and  $BA = \begin{pmatrix} 1 & 0 \\ 2 & 2 \end{pmatrix}$   
•  $AC = \begin{pmatrix} 5 & 6 & 7 \\ 4 & 4 & 4 \end{pmatrix}$   
•  $(AB)^t = B^t A^t = \begin{pmatrix} 2 & 1 \\ 0 & 1 \end{pmatrix}$ 

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Introduction Motivatio Calculus Matrix alg Linear algebra The link v Fundamentals of probability theory Applicatit

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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## Exercises

• Let 
$$A = \begin{pmatrix} 3 & 6 \\ -1 & -2 \end{pmatrix}, B = \begin{pmatrix} 1 & 2 \\ 3 & -1 \end{pmatrix}, C = \begin{pmatrix} 1 & 1 \end{pmatrix}$$
 and  $D = \begin{pmatrix} 1 & 2 & 0 & 1 \\ 4 & 0 & -3 & 1 \end{pmatrix}$ 

• Compute 
$$-3C$$
,  $A + B$ ,  $A - D$  and  $D^t$ 

- Compute AB, BA, AC, CA, AD and DA
- Let A be a symmetric matrix, show then that  $A^t = A$
- A square matrix A is called idempotent if  $A^2 = A$ 
  - Verify which of the above matrices are idempotent
  - Find the value of  $\alpha$  that makes the following matrix

idempotent:  $\begin{pmatrix} -1 & 2 \\ \alpha & 2 \end{pmatrix}$ 

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

Two numbers associated to square matrices: trace

#### Let $A, B, C \in \mathbb{R}^{n \times n}$

- Trace(A) =  $tr(A) = \sum_{i=1}^{n} a_{ii}$
- Used in econometrics
- Properties
  - $tr(A^t) = tr(A)$
  - tr(A + B) = tr(A) + tr(B)
  - tr(cA) = ctr(A) for any  $c \in \mathbb{R}$
  - tr(AB) = tr(BA)
  - $tr(ABC) = tr(BCA) = tr(CAB) \neq tr(ACB)(= tr(BAC) = tr(CBA))$

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

Two numbers associated to square matrices: trace

#### Example

• Let 
$$A = \begin{pmatrix} 3 & 6 \\ -1 & -2 \end{pmatrix}$$
 and  $B = \begin{pmatrix} 1 & 2 \\ 3 & -1 \end{pmatrix}$   
• Then  $AB = \begin{pmatrix} 21 & 0 \\ -7 & 0 \end{pmatrix}$ ,  $BA = \begin{pmatrix} 1 & 2 \\ 10 & 20 \end{pmatrix}$  and  $A + B = \begin{pmatrix} 4 & 8 \\ 2 & -3 \end{pmatrix}$   
•  $tr(A) = 1$ ,  $tr(B) = 0$  and  $tr(A + B) = 1$   
•  $tr(AB) = 21 = tr(BA)$ 

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Two numbers associated to square matrices: determinant

Let  $A \in \mathbb{R}^{n \times n}$ 

- If n = 1, then  $det(A) = a_{11}$
- If n = 2, then  $\det(A) = a_{11}a_{22} - a_{12}a_{21} = a_{11}\det(a_{22}) - a_{12}\det(a_{21})$
- If n = 3, then  $\det(A) = a_{11} \det \begin{pmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{pmatrix} a_{12} \det \begin{pmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{pmatrix} + a_{13} \det \begin{pmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{pmatrix}$
- Can be generalized to any n
- Works with columns too

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# Two numbers associated to square matrices: determinant

Let  $A, B \in \mathbb{R}^{n \times n}$ 

- $det(A^t) = det(A)$
- $det(A + B) \neq det(A) + det(B)$
- $\det(cA) = c \det(A)$  for any  $c \in \mathbb{R}$
- det(AB) = det(BA)
- A is non-singular (or regular) if  $A^{-1}$  exists

• I.e. 
$$AA^{-1} = A^{-1}A = I_n$$

- In is a diagonal matrix with 1 on the diagonal
- Does not always exist
- det(A) ≠ 0

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# Two numbers associated to square matrices: determinant

Example

• Let 
$$A = \begin{pmatrix} 3 & 6 \\ -1 & -2 \end{pmatrix}$$
 and  $B = \begin{pmatrix} 1 & 2 \\ 3 & -1 \end{pmatrix}$   
• Then  $AB = \begin{pmatrix} 21 & 0 \\ -7 & 0 \end{pmatrix}$ ,  $BA = \begin{pmatrix} 1 & 2 \\ 10 & 20 \end{pmatrix}$  and  
 $A + B = \begin{pmatrix} 4 & 8 \\ 2 & -3 \end{pmatrix}$   
• det $(A) = 0$ , det $(B) = -7$  and det $(A + B) = -28$ 

• det(AB) = 0 = det(BA)

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Exercises

• Let 
$$A = \begin{pmatrix} 2 & 4 & 5 \\ 0 & 3 & 0 \\ 1 & 0 & 1 \end{pmatrix}$$

• Compute tr(A) and det(-2A)

- Show that for any triangular matrix *A*, we have that det(*A*) is equal to the product of the elements on the diagonal
- Let  $A, B \in \mathbb{R}^{n \times n}$  and assume that B is non-singular
  - Show that  $tr(B^{-1}AB) = tr(A)$
  - Show that  $tr(B(B^tB)^{-1}B^t) = n$
- Let  $A, B \in \mathbb{R}^{n \times n}$  be two non-singular matrices
  - Show that AB is then also invertible
  - Give an expression for  $(AB)^{-1}$

Motivation Matrix algebra **The link with vector spaces** Application: solving a system of linear equations

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# Outline



2 Calculus

#### 3 Linear algebra

- Motivation
- Matrix algebra
- The link with vector spaces
- Application: solving a system of linear equations

#### 4 Fundamentals of probability theory

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# A notion of vector spaces

- A set of vectors V is a vector space if
  - Addition of vectors is well-defined
  - $\forall a, b \in V : a + b \in V$
  - Scalar multiplication is well-defined
  - $\forall k \in \mathbb{R}, \forall a \in V : ka \in V$
- We can take linear combinations
  - $\forall k_1, k_2 \in \mathbb{R}, \forall a, b \in V : k_1a + k_2b \in V$
- E.g.  $\mathbb{R}^2$  or more generally  $\mathbb{R}^n$
- Counterexample R<sup>2</sup><sub>+</sub>

Motivation Matrix algebra **The link with vector spaces** Application: solving a system of linear equations

# Linear (in)dependence

#### Let V be a vector space

 A set of vectors v<sub>1</sub>,..., v<sub>n</sub> ∈ V is *linear dependent* if one of the vectors can be written as a linear combination of the others

• 
$$\exists k_1, \ldots, k_{n-1} \in \mathbb{R}$$
 :  $v_n = k_1 v_1 + \cdots + k_{n-1} v_{n-1}$ 

• A set of vectors are *linear independent* if they are not linear dependent

•  $\forall k_1, \ldots, k_n \in \mathbb{R} : k_1 v_1 + \cdots + k_n v_n = 0 \Rightarrow k_1 = \cdots = k_n = 0$ 

 In a vector space of dimension n, the number of linear independent vectors cannot be higher than n

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# Linear (in)dependence

#### Example

•  $\mathbb{R}^2$  is a vector space of dimension 2

• 
$$v_1 = (1,0), v_2 = (1,2), v_3 = (-1,4)$$
 and  $v_4 = (2,4)$ 

- $v_3 = -3v_1 + 2v_2$ , so  $v_1, v_2, v_3$  are linear dependent
- $v_4 = 2v_2$ , so  $v_2$ ,  $v_4$  are linear dependent
- v<sub>1</sub>, v<sub>2</sub> are linear independent
- v<sub>3</sub> is linear independent

Motivation Matrix algebra **The link with vector spaces** Application: solving a system of linear equations

# Link with matrices: rank

Let  $A \in \mathbb{R}^{n \times m}$  and  $B \in \mathbb{R}^{m \times k}$ 

- The row rank of A is the maximal number of linear independent rows of A
- The column rank of A is the maximal number of linear independent columns of A
- Rank of A = column rank of A = row rank of A
- Properties
  - $rank(A) \leq min(n, m)$
  - rank(AB) ≤ min(rank(A), rank(B))
  - $rank(A) = rank(A^{t}A) = rank(AA^{t})$
  - If n = m, then A has maximal rank if and only if det(A)  $\neq 0$

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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## Link with matrices: rank

Example

• Let 
$$A = \begin{pmatrix} 3 & 6 \\ -1 & -2 \end{pmatrix}$$
 and  $B = \begin{pmatrix} 1 & 2 \\ 3 & -1 \end{pmatrix}$   
• Then  $AB = \begin{pmatrix} 21 & 0 \\ -7 & 0 \end{pmatrix}$ ,  $BA = \begin{pmatrix} 1 & 2 \\ 10 & 20 \end{pmatrix}$   
•  $rank(A) = 1$  and  $rank(B) = 2$ 

• 
$$rank(AB) = 1$$

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Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

## Exercises

• Let 
$$A = \begin{pmatrix} 2 & 4 & 5 \\ 0 & 3 & 0 \\ 1 & 0 & 1 \end{pmatrix}$$

• Show in two ways that A has maximal rank

- Let  $A, B \in \mathbb{R}^{n \times m}$ 
  - Show that there need not be any relation between rank(A + B), rank(A) and rank(B)
- Show that if *A* is invertible, then it needs to have a maximal rank

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Outline



- 2 Calculus
- 3 Linear algebra
  - Motivation
  - Matrix algebra
  - The link with vector spaces
  - Application: solving a system of linear equations

#### 4 Fundamentals of probability theory

Introduction Calculus Linear algebra

Fundamentals of probability theory

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

## System of linear equations

- Linear equations in the unknowns *x*<sub>1</sub>,..., *x<sub>m</sub>* 
  - Not  $x_1 x_2, x_m^2, ...$

Constraints hold with equality

• Not 
$$2x_1 + 5x_2 \le 3$$
  
• E.g.  $\begin{cases} 2x_1 + 3x_2 - x_3 = 5 \\ 3x_1 + 3x_2 - x_3 = 5 \end{cases}$ 

$$-x_1 + 4x_2 + x_3 = 0$$

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# Using matrix notation

- *m* unknowns: *x*<sub>1</sub>,...,*x<sub>m</sub>*
- *n* linear constraints:  $a_{i1}x_1 + \cdots + a_{im}x_m = b_i$  with  $a_{i1}, \ldots, a_{im}, b_i \in \mathbb{R}$  and  $i = 1, \ldots, n$
- Ax = b

• 
$$A = (a_{ij})_{i=1,...,n,j=1,...,m}$$
  
•  $x = (x_i)_{i=1,...,n}$   
•  $b = (b_i)_{i=1,...,n}$ 

• Homogeneous if  $b_i = 0$  for all i = 1, ..., n

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Solving these system

- Solve this by logical reasoning
  - Eliminate or substitute variables
  - · Can also be used for non-linear systems of equations
  - Can be cumbersome for larger systems
- Use matrix notation
  - Gaussian elimination of the augmented matrix (A|b)
  - Can be programmed
  - Only for systems of linear equations
  - Theoretical statements are possible

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

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# Example

$$\begin{cases} -x_1 + 4x_2 = 0\\ 2x_1 + 3x_2 = 5 \end{cases}$$

• 
$$x_1 = 4x_2 \Rightarrow 11x_2 = 5 \Rightarrow x_2 = \frac{5}{11}$$
 and  $x_1 = \frac{20}{11}$   
•  $\begin{pmatrix} -1 & 4 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 5 \end{pmatrix}$ 

• Take linear combinations of rows of the augmented matrix • Is the same as taking linear combinations of the equations •  $\begin{pmatrix} -1 & 4 & | 0 \\ 2 & 3 & | 5 \end{pmatrix} \Rightarrow \begin{pmatrix} -1 & 4 & | & 0 \\ 0 & 11 & | & 5 \end{pmatrix} \Rightarrow \begin{pmatrix} -1 & 0 & | & \frac{-20}{11} \\ 0 & 11 & | & 5 \end{pmatrix}$ 

Motivation Matrix algebra The link with vector spaces Application: solving a system of linear equations

# Theoretical results

Consider the linear system Ax = b with  $A \in \mathbb{R}^{n \times m}$ 

- This system has a solution if and only if rank(A) = rank(A|b)
  - $rank(A) \le rank(A|b)$  by definition
  - Is the (column) vector *b* a linear combination of the column vectors of *A*?
  - If rank(A) < rank(A|B), the answer is no
  - If rank(A) = rank(A|B), the answer is yes
- The solution is unique if rank(A) = rank(A|B) = m

n ≥ m

- There are  $\infty$  many solutions if rank(A) = rank(A|B) < m
  - *n* < *m* or too many constraints are 'redundant'

## Exercises

Solve the following systems of linear equations

• 
$$\begin{cases} x_1 + 2x_2 + 3x_3 = 1\\ 3x_1 + 2x_2 + x_3 = 1 \end{cases} \text{ and } \begin{cases} x_1 - x_2 + x_3 = 1\\ 3x_1 + x_2 + x_3 = 0\\ 4x_1 + 2x_3 = -1 \end{cases}$$

• For which values of *k* does the following system of linear equations have a unique solution?

$$\begin{cases} x_1 + x_2 = 1\\ x_1 - kx_2 = 1 \end{cases}$$

- Consider the linear system Ax = b with  $A \in \mathbb{R}^{n \times n}$ 
  - Show that this system has a unique solution if and only if A is invertible
  - Give a formula for this unique solution
- Show that homogeneous systems of linear equations always have a (possibly non-unique) solution

## Outline



## 2 Calculus

#### 3 Linear algebra

4 Fundamentals of probability theory

#### To be continued

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