

Math Bootcamp AMPBA

Lecture 1

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Today's agenda

- About this course.
- How to get most out of this course?
- Student interaction – to set expectations and understanding starting knowledge.
- Vectors and Matrices.

Introduction

- This course introduces you to linear algebra and matrix algebra.
- This will help you to understand the math behind most of the econometrics, machine and deep learning models.
- Ideas/Concepts covered in these series of lectures are mathematical in nature.
- I will provide some applications/real world connections of these ideas.

How to get most out of this course?

- Ideas grasping through writing.
- Practice.
- Asking questions on LMS/email.
- Knowing that, this is a starting point. You might not be able to appreciate all the ideas now. It's okay, it takes time.

Student Interaction

- 20 mins session.
- What do students know? Covering extremes. (10-15 mins)
- R session. (5-10 mins)

What is a vector?

- A vector is collection of numbers.
- A vector \mathbf{x} of order p (or dimension p) is a column of p numbers:

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_p \end{pmatrix}.$$

Vector

- We can refer \mathbf{x} as column vector.
- The transpose of \mathbf{x} , $\mathbf{x}^T = (x_1, x_2, \dots, x_p)$ is a row vector.
- Addition and subtraction of vectors of the same order is performed element by element

$$\mathbf{x} + \mathbf{y} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ \vdots \\ x_p + y_p \end{pmatrix}.$$

Vector

- It is not possible to add or subtract vectors which are not conformable, i.e., which do not have the same dimension or order.
- Scalar multiplication of a vector is element by element:

$$\lambda \mathbf{x} = \begin{pmatrix} \lambda x_1 \\ \lambda x_2 \\ \vdots \\ \lambda x_p \end{pmatrix}$$

Vector

- Two vectors x and y are equal if they are of the same order and each pair of corresponding elements is equal, i.e., $x_i = y_i$
- A vector with all elements 0 is denoted by $\mathbf{0}$
- A vector e_i with i^{th} element 1 and all others 0 is the i^{th} unit vector.
- A vector with all elements 1 is and is denoted by $\mathbf{1}$

Vector

- Sum vector. $X^T \mathbf{1}_p = \sum_{i=1}^p x_i$
- $x+y = y+x$ (commutativity) and $(x+y)+z = x+(y+z)$ (associativity).
- Inner Product $x^T y$
- Example: $x=(1,2,3)^T$ and $y=(4,5,6)^T$. Find $x^T y$

Vector

- Calculate $x^T x$, where x is a p dimensional vector?

Summary

- Vector definition.
- Row v/s column vector.
- Vector dimension.
- Vector Operations.
- Types of vectors.
- Inner product.

Matrices

- An $m \times n$ matrix X is a rectangular array of scalar numbers:

$$X = \begin{pmatrix} x_{11} & x_{12} & \dots & x_{1n} \\ x_{21} & x_{22} & \dots & x_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ x_{m1} & x_{m2} & \dots & x_{mn} \end{pmatrix}.$$

- This matrix has m rows and n columns; it is an $m \times n$ matrix (m by n matrix), it is a matrix of order mn . X has dimensions m and n .

Matrices

- The numbers x_{ij} are the components (or elements) of X .
- A matrix is a square matrix if $m = n$, i.e., if it has the same number of rows and columns, i.e., a $n \times n$ matrix is a square matrix.
- The transpose of the $m \times n$ matrix X is the $n \times m$ matrix X^T

$$X = \begin{pmatrix} x_{11} & x_{12} & \dots & x_{1n} \\ x_{21} & x_{22} & \dots & x_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ x_{m1} & x_{m2} & \dots & x_{mn} \end{pmatrix}.$$

Matrices

- A square matrix X is **symmetric** if $X^T = X$. Two matrices X and Y are **equal** if they are of the same order and each pair of corresponding elements are equal, i.e., $x_{ij} = y_{ij}$, for $i = 1; 2; \dots ; m$ and $j = 1; 2; \dots ; n$.

Matrices

- A square matrix with all elements not on the diagonal equal to 0 is a diagonal matrix, i.e., if $x_{ij} = 0$ for all $i \neq j$
- A diagonal matrix with all diagonal elements 1 (and all others not on the diagonal 0) is denoted by \mathbf{I}_n , called Identity Matrix.

$$\mathbf{I}_n = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix}.$$

Matrices

- The trace of a square matrix is the sum of all the diagonal elements of the matrix, i.e., $\text{trace}(X) = \text{tr}(X) = \text{tr}(x_{ij}) = \sum_{i=1}^n x_{ii}$.

Note that $\text{tr}(\mathbf{I}_n) = n$.

Matrix Arithmetic

- $\mathbf{X} + \mathbf{Y} = x_{ij} + y_{ij}$
- $\mathbf{X} + \mathbf{Y} = \mathbf{Y} + \mathbf{X}$ (commutativity) and $(\mathbf{X} + \mathbf{Y}) + \mathbf{Z} = \mathbf{X} + (\mathbf{Y} + \mathbf{Z})$ (associativity).
- Scalar multiplication of a matrix is element by element:

$$\lambda \mathbf{X} = \lambda(x_{ij}) = (\lambda x_{ij})$$

Matrix Arithmetic

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Matrix Multiplication

- If A and B are matrices then we can multiply A by B (to get AB) only if the number of columns of A equals the number of rows of B .
- If C is $m \times n$ and D is $p \times q$ then the product CD can only be defined if $n = p$, in which case C and D are said to be **conformable**. If C and D are such that CD is not defined then they are **non-conformable**.

Matrix Multiplication

- Example

If $\mathbf{U} = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$, $\mathbf{V} = \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix}$ then \mathbf{U} is 2×2 and \mathbf{V} is 2×2 so \mathbf{UV} is $2 \times 2 \times 2 \times 2 \equiv 2 \times 2$ and \mathbf{VU} is $2 \times 2 \times 2 \times 2 \equiv 2 \times 2$.

Matrix Multiplication

$$\begin{aligned} \mathbf{UV} &= \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} \\ &= \begin{pmatrix} 1 \times 5 + 2 \times 7 & 1 \times 6 + 2 \times 8 \\ 3 \times 5 + 4 \times 7 & 3 \times 6 + 4 \times 8 \end{pmatrix} \\ &= \begin{pmatrix} 5 + 14 & 6 + 16 \\ 15 + 28 & 18 + 32 \end{pmatrix} \\ &= \begin{pmatrix} 19 & 22 \\ 43 & 50 \end{pmatrix} \end{aligned}$$

Matrix Multiplication

- Now calculate **VU** .
- Is **$UV=VU$** ?

Matrix Multiplication

$$\text{If } \mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \text{ and } \mathbf{B} = \begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix}.$$

Matrix Multiplication

$$\begin{aligned} \mathbf{AB} &= \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix} \\ &= \begin{pmatrix} 1 \times 1 + 2 \times 3 + 3 \times 5 & 1 \times 2 + 2 \times 4 + 3 \times 6 \\ 4 \times 1 + 5 \times 3 + 6 \times 5 & 4 \times 2 + 5 \times 4 + 6 \times 6 \end{pmatrix} = \begin{pmatrix} 22 & 28 \\ 49 & 64 \end{pmatrix}. \end{aligned}$$

$$\begin{aligned} \mathbf{BA} &= \begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \\ &= \begin{pmatrix} 1 \times 1 + 2 \times 4 & 1 \times 2 + 2 \times 5 & 1 \times 3 + 2 \times 6 \\ 3 \times 1 + 4 \times 4 & 3 \times 2 + 4 \times 5 & 3 \times 3 + 4 \times 6 \\ 5 \times 1 + 6 \times 4 & 5 \times 2 + 6 \times 5 & 5 \times 3 + 6 \times 6 \end{pmatrix} = \begin{pmatrix} 9 & 12 & 15 \\ 19 & 26 & 33 \\ 29 & 40 & 51 \end{pmatrix}. \end{aligned}$$

Matrix Multiplication

- Is the following true?
- **$(A+B)^2=A^2+B^2+2AB$**
- If false, when will it be true?

Transpose and Trace of Sums and Products

If $\mathbf{A} + \mathbf{B}$ is defined (i.e., \mathbf{A} and \mathbf{B} have the same orders) then $(\mathbf{A} + \mathbf{B})' = \mathbf{A}' + \mathbf{B}'$ and $\text{tr}(\mathbf{A} + \mathbf{B}) = \text{tr}(\mathbf{A}) + \text{tr}(\mathbf{B})$. If \mathbf{A} is $m \times n$ and \mathbf{B} is $n \times p$ then \mathbf{AB} is $m \times n \times n \times p \equiv m \times p$ so $(\mathbf{AB})'$ is $p \times m$. It is easy to show that $(\mathbf{AB})' = \mathbf{B}'\mathbf{A}'$:

$$(\mathbf{AB})' = ((a_{ij})(b_{jk}))' = \left(\sum_{j=1}^n a_{ij}b_{jk} \right)' = \left(\sum_{j=1}^n a_{kj}b_{ji} \right) = (b_{jk})'(a_{ij})' = \mathbf{B}'\mathbf{A}'.$$

Note that \mathbf{A}' is $n \times m$ and \mathbf{B}' is $p \times n$ so the product $\mathbf{A}'\mathbf{B}'$ is not defined (unless $p = m$) but $\mathbf{B}'\mathbf{A}'$ is defined.

Clearly $\text{tr}(\mathbf{A}) = \text{tr}(\mathbf{A}')$ and if \mathbf{A} and \mathbf{B} are $m \times n$ and $n \times m$ matrices (so that both \mathbf{AB} and \mathbf{BA} are defined) then $\text{tr}(\mathbf{AB}) = \text{tr}(\mathbf{BA})$ because $\text{tr}(\mathbf{AB}) = \sum_{i=1}^m \sum_{j=1}^n a_{ij}b_{ij} = \sum_{j=1}^n \sum_{i=1}^m a_{ij}b_{ij} = \text{tr}(\mathbf{BA})$.

If λ is any scalar then $(\lambda\mathbf{A})' = \lambda\mathbf{A}'$.

Symmetric and skew-symmetric matrices

A square $n \times n$ matrix $\mathbf{A} = (a_{ij})$ is ***symmetric*** if $\mathbf{A}' = \mathbf{A}$, i.e., $a_{ij} = a_{ji}$ for all i, j . A square matrix $\mathbf{B} = (b_{ij})$ is ***skew-symmetric*** if $\mathbf{B}' = -\mathbf{B}$, i.e., $b_{ij} = -b_{ji}$ for all i, j . It is easy to see that all skew-symmetric matrices have zero elements on the diagonals.

Symmetric and skew-symmetric matrices

Any square matrix X can be expressed as $X = \frac{1}{2}(X + X') + \frac{1}{2}(X - X')$. Since $(X + X')' = (X + X')$ and $(X - X')' = -(X - X')$ we have that any square matrix can be expressed as the sum of a *symmetric part* and a *skew-symmetric part*.

Symmetric and skew-symmetric matrices

- Are A and B symmetric?

- Is \mathbf{AB} symmetric?

$$\text{Let } \mathbf{A} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$$

Orthogonal

- A $p \times p$ square matrix A is orthogonal if $A'A = AA' = I_p$
- If A and B are both orthogonal and both $p \times p$ then AB is orthogonal. **Why?**
- An orthogonal matrix whose elements are all $+1$ or -1 is known as a **Hadamard matrix**.

Orthogonal

- Are A and B orthogonal matrices?

$$\mathbf{A} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} \frac{1}{2} & -\frac{\sqrt{3}}{2} \\ -\frac{\sqrt{3}}{2} & -\frac{1}{2} \end{pmatrix}$$

Other names

- A $p \times p$ matrix is **normal** if $AA' = A'A$
- A $p \times p$ matrix A is **idempotent** if $A^2 = A$
- A matrix $A \neq 0$ is said to be **nilpotent** if $A^2 = 0$
- A $n \times n$ matrix A such that $A^2 = I_n$ is said to be **unipotent**.
- Example: $A = \begin{bmatrix} 1 & x \\ 0 & -1 \end{bmatrix}$, which of the above is A ?

R exercises