# Mathematical Concepts (G6012)

#### Lecture 10

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# **Church/Turing Thesis**

- Every computable function can be computed by a Turing Machine
- I.e.: Turing Machines are universal computing machines
- Every problem that can be solved by an algorithm can be solved by a Turing machine
- Where is the power coming from? The read/write input/output tape !

# More about TM

- The tape can be used to record any data for later access
- There is always space available after last non-blank location
- There is no limit how often the tape is accessed
- Your PC is less powerful than a TM why? Because it has finite memory

# Efficiency

- TM are universal but not efficient
- Progress can be really slow
- Looking up memory involves sequential access – the opposite of efficiency

# Managing complexity

- One can encapsulate useful functionality in "separate" sub-routines
- Collection of states set aside for each subroutine
- (similar to structured programming approach)
- However, TM are mainly useful as a theoretical concept, not for solving real world problems!

# Variations

- There are common variants of TM:
  - Multiple tapes
  - Single-side infinite tape
  - Non-deterministic TM
- It can be shown that these have all equivalent power to the TM discussed here.

# Example: Non-deterministic TM

- To simulate a non-deterministic TM:
  - 3 tapes:
  - One tape for original input
  - One tape for the choice sequence: (2,3,1,2)
  - One tape to run on current choice sequence
- For this to work we need to enumerate all possible sequences of choices (ok, as states are finite)

# Another equivalence

- The "2 pushdown" automaton is equivalent to the Turing Machine:
  - One pushdown holds tape contents to left of tape head
  - One pushdown holds tape contents to the right of tape head
  - As tape head moves, symbols shift across from one pushdown to another

# More generally ...

- Chomsky Hierarchy (for language classes):
  - Type 0: Languages accepted by Turing Machines
  - Type 1: Languages accepted by Turing Machines with linear bounded storage
  - Type 2: Languages accepted by Pushdown Automata
  - Type 3: Languages accepted by Finite State Automata

# **Alternative Characterization**

- Equivalent grammar formalisms:
  - Type 0: Languages generated by unrestricted grammars
  - Type 1: Languages generated by contextsensitive grammars
  - Type 2: Languages generated by context-free grammars
  - Type 3: Languages generated by regular grammars

#### Equivalence and Inclusions



#### **BB**: Full names and acronyms

TM = Turing Machine

LBTM = Linearly bounded Turing Machine

PDA = Pushdown Automaton

FSA = Finite State Automaton

UG = Unrestricted Grammar

CSG = Context Sensitive Grammar

CFG = Context Free Grammar

RG = Regular Languages

#### **VECTORS AND MATRICES**

# Why matrix algebra?

- Multimedia/Design/Art: Computer graphics are 90% vectors and matrices
- AI: Artificial Neural Networks heavily depend on vectors and matrices.
- Music: Discretised sound spectra are vectors; digital filtering & enhancement depend on matrices; modern compression (mp3 etc) is one of the most maths-heavy problems in Informatics

#### **VECTORS AND MATRICES**

# Vectors & Matrices

- A matrix or a vector is simply a way of representing a structured collection of numbers.
- Vectors are order 1 (rows, columns) and can be used to represent
  - sound samples
  - general arrays
  - lists of things
  - datasets ...

- points in space
- directions in space
- velocity

# Vectors & Matrices

- Matrices are order 2 (rectangles) and can be used to represent
  - Images
  - Datasets
  - Transformations of vectors
  - Parameters of Artificial Neural Networks
  - •

## **Vector Notations**



Row vector

### Geometric interpretation

Vectors are (arrows to) points in space:



#### Multiplying numbers with vectors

 $ec{x} \in \mathbb{R}^3$  and  $a \in \mathbb{R}$ then  $a \cdot ec{x} \in \mathbb{R}^3$ 

$$a \cdot \vec{x} = a \cdot \left( egin{array}{c} x_1 \ x_2 \ x_3 \end{array} 
ight) = \left( egin{array}{c} a \cdot x_1 \ a \cdot x_2 \ a \cdot x_3 \end{array} 
ight)$$

Interpretation: **BB** 

## **BB** Interpretation

Multiplying with  $a \in \mathbb{R}$  means stretching (a > 1) or shortening (-1< a < 1) and/or mirroring through the center (a < 0)



## Adding vectors

If  $\vec{x} \in \mathbb{R}^3$  and  $\vec{y} \in \mathbb{R}^3$ then  $\vec{x} + \vec{y} \in \mathbb{R}^3$ 

$$\vec{x} + \vec{y} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} + \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ x_3 + y_3 \end{pmatrix}$$

#### **Example & Interpretation**





#### **Basis vectors**

Every vector can be expressed as a combination of basis vectors

$$\vec{e}_1 = \begin{pmatrix} 1\\0\\0 \end{pmatrix} \quad \vec{e}_2 = \begin{pmatrix} 0\\1\\0 \end{pmatrix} \quad \vec{e}_3 = \begin{pmatrix} 0\\0\\1 \end{pmatrix}$$
$$\begin{pmatrix} x_1\\x_2\\x_3 \end{pmatrix} = x_1 \cdot \vec{e}_1 + x_2 \cdot \vec{e}_2 + x_3 \vec{e}_3$$

### Geometric interpretation

Vectors are (arrows to) points in space:



This just formalises that if going from 0 to a point you can go in x direction first, then in y direction instead of going diagonal ...

#### Subtraction





### Matrices

$$\left( egin{array}{cccc} -1 & 5 & -4 \ 9.1 & 3 & -4.5 \ 7 & 0.1 & \sqrt{2} \end{array} 
ight) \in \, \mathcal{M}(3,3) \, \, ext{is a 3x3 matrix.}$$



#### Tensors

Generalization: Vector: 1-tensor Matrix: 2-tensor

#### Example: 3-tensor $A = (a_{ijk})$ $i = 1, \dots, 3$ $j = 1, \dots, 3$ is a 3x3x3 tensor $k = 1, \dots, 3$ with 27 entries

(If you write it out it would be a cube of numbers)

# Adding and subtracting matrices

• Same as for vectors ...

#### BB

Interpretation not so direct: Operations on vectors – next time.

# **BB** Example: Subtractring a matrix from an other matrix

$$\begin{pmatrix} 1 & -2 & 3 \\ 0 & 4 & -2 \\ -2 & 2 & 1 \end{pmatrix} - \begin{pmatrix} 1 & 0 & -1 \\ 0 & 2 & 2 \\ 3 & 1 & -1 \end{pmatrix}$$
$$= \begin{pmatrix} 1-1 & -2-0 & 3-(-1) \\ 0-0 & 4-2 & -2-2 \\ -2-3 & 2-1 & 1-(-1) \end{pmatrix}$$
$$= \begin{pmatrix} 0 & -2 & 4 \\ 0 & 2 & -4 \\ -5 & 1 & 2 \end{pmatrix}$$

#### Properties of +, -

 $A, B, C \in \mathcal{M}(m, n)$ 

Associativity A + B + C = (A + B) + C = A + (B + C)Commutativity A + B = B + A

# Properties of scalar multiplication

 $A, B, C \in \mathcal{M}(m, n)$   $r, s \in \mathbb{R}$ 

Compatibility with scalar operations (multiplying with numbers)

$$r \cdot (A + B) = r \cdot A + r \cdot B$$

$$(r+s)\cdot A = r\cdot A + s\cdot A$$

#### **Matrix-Vector Multiplication**

$$A \in \mathsf{M}(3,3) \quad \text{and} \quad \vec{x} \in \mathbb{R}^{3}$$
$$A \cdot \vec{x} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \\ x_{3} \end{pmatrix} =$$

BB

#### **BB** Matrix-vector multiplication

$$A\vec{x} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 \\ a_{31}x_1 + a_{32}x_2 + a_{33}x_3 \end{pmatrix}$$

#### The result is again a vector!

#### **BB** Matrix-vector multiplication

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#### The result is again a vector!

#### **BB** Matrix-vector multiplication



#### The result is again a vector!

# Properties of Matrix-Vector Multiplication

Linear (both ways)  $A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y}$  $(A + B)\vec{x} = A\vec{x} + B\vec{x}$ 

Associative:

$$(A \cdot B)\vec{x} = A(B\vec{x})$$

### Interpretation

Matrices are transformations (linear functions)

$$A = \underline{A} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \in M(3,3)$$

 $\begin{array}{ccc} A: \mathbb{R}^3 \to \mathbb{R}^3 & A \text{ maps vectors from } \mathbb{R}^3 \text{ to} \\ \vec{x} \mapsto A \cdot \vec{x} & & \\ A \vec{x} & & \\ \end{array} & \begin{array}{c} A \text{ maps vectors in } \mathbb{R}^3 \end{array} \\ \begin{array}{c} A \text{ maps vectors from } \mathbb{R}^3 \text{ to} \\ \text{ vectors in } \mathbb{R}^3 \end{array} \end{array}$ 

### Matrix as a transformation: Can we see what it does?



BB

#### **BB** Matrix as a transformation

$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} a_{11} \\ a_{21} \\ a_{31} \end{pmatrix}$$
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} a_{12} \\ a_{22} \\ a_{32} \end{pmatrix}$$
$$\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 0 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} a_{13} \\ a_{23} \\ a_{33} \end{pmatrix}$$

Column of the matrix are the images of the basis vectors!

## Matrix as a transformation

The columns of the matrix are the vectors the basis vectors

$$\vec{e}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad \vec{e}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \vec{e}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

are mapped to!

Example: **BB** 



# Remember: Basis vectors "span" the space

Every vector can be expressed as the sum of basis vectors:

$$\left(\begin{array}{c} x_1 \\ x_2 \\ x_3 \end{array}\right) = x_1 \left(\begin{array}{c} 1 \\ 0 \\ 0 \end{array}\right) + x_2 \left(\begin{array}{c} 0 \\ 1 \\ 0 \end{array}\right) + x_3 \left(\begin{array}{c} 0 \\ 0 \\ 1 \end{array}\right)$$

 $\vec{x} = x_1 \vec{e}_1 + x_2 \vec{e}_2 + x_3 \vec{e}_3$ 

So we can see how a matrix defines a mapping of the whole space.

## **Reminder: Summation notation**



#### **Matrix-vector Multiplication**

$$(A\vec{x})_i = \sum_{j=1}^3 a_{ij}x_j = a_{i1}x_1 + a_{i2}x_2 + a_{i3}x_3$$

... simplifies many calculations.