# Discrete Mathematics and Its Applications 2 (CS147)

Lecture 1: Introduction to the module

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# **Module Organizers**

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- Ramanujan Sridharan (Week 6 to Week 10): R.Maadapuzhi-Sridharan@warwick.ac.uk

## Details can be found in module webpage:

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https://warwick.ac.uk/fac/sci/dcs/teaching/modules/cs147/
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- ▶ google search: Warwick CS -> https://warwick.ac.uk/fac/sci/dcs/
- ► Teaching -> Modules Taught -> CS147

#### What the module is about

- ► The word "application" in the name refers to "applications in theoretical computer science (TCS)" and "applications in machine learning theory"
- ▶ This is a (completely) mathematical module.

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- ► This is a (completely) mathematical module.
- o theoretical computer science: theory of computation, algorithms analysis
- o machine learning theory:
  - machine learning: learn rules from data
  - theory: TCS, statistical principles

# Reasons to analyse algorithms

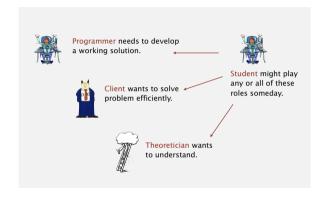


figure credit:

# Reasons to analyse algorithms

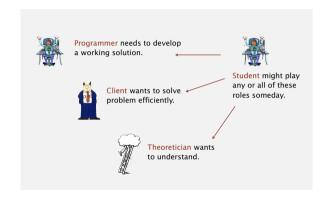


figure credit:

- predict performance
- compare algorithms
- provide guarantees

# **Example:** image recognition

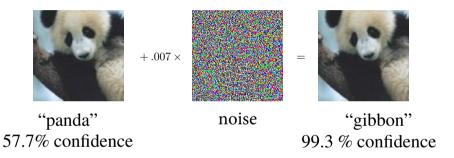


Figure: sensitive output by algorithms. source from [GSS15]

## Contents of the module

o CS146: proofs, sequences, sets, relations...

o CS147: algorithm analysis...

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- o CS146: proofs, sequences, sets, relations...
- o CS147: algorithm analysis...
  - How to analyse runtimes of algorithms
  - discrete probability
  - graph theory and combinatonics
- ⇒ will prepare you for TCS modules in years 2,3 as well as machine learning theory

# \*Examples in TCS, ML theory

**Corollary 1.3.** Let  $\mathcal{D}$  be the distribution over pairs  $(x,y) \in \mathbb{R}^d \times \mathbb{R}$  where  $x \sim \mathcal{N}(0, \mathrm{Id})$  and y = F(x) for a size-S ReLU network F for which the product of the spectral norms of its weight matrices is a constant.

Then there is an algorithm that draws  $N = d\log(1/\delta)\exp(O(k^3/\varepsilon^2 + kS))$  samples, runs in time  $\widetilde{O}(d^2\log(1/\delta))\exp(O(k^3S^2/\varepsilon^2 + kS^3))$ , and outputs a ReLU network  $\widetilde{F}$  such that  $\mathbb{E}[(y-\widetilde{F}(x))^2] \leq \varepsilon$  with probability at least  $1-\delta$ .

Figure: sample complexity and time complexity [CKM22].

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Figure: sample complexity and time complexity [CKM22].

#### It answers the following questions:

- what is the performance of this algorithm?
- guarantees for algorithms (under what conditions will they succeed, how much data and computation time is needed)

## Student Cohort

- Core module for Discrete Maths students
- Optional module for Maths students

# Clarifying two misconceptions

- Only prerequisite: Mathematical maturity
- ▶ I am well aware of possible overlaps with other modules and some prior background you may already have

## Lectures

Lectures (Week 1 to 10)

- ► Thursday 9:00 10:00 (PLT)
- Friday 15:00 16:00 (R0.21)
- Friday 16:00 17:00 (Woods-Scawen room)

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After each lecture, I will provide links to

- recording of the lecture
- ► relevant lecture notes/online book chapters in the module webpage

Be proactive yourself

#### **Seminars**

#### 9 seminars from Week 2

- ▶ Week 2 seminar questions will be posted in the module webpage before Monday morning of Week 2, and so on.
- ▶ Try to solve the questions yourself, before attending your seminar sessions.

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## Seminar Groups:

- ▶ Group 1: Thursdays 10 am 11 am in S0.17.
- ▶ Group 2: Thursdays 11 am 12 am at OC0.05.
- ► Group 3: Thursdays 12 pm 1 pm at H0.03.
- ► Group 4: Thursdays 1 pm 2 pm at S0.18.
- ► Group 5: Fridays 2 pm 3 pm at S0.10.

# Some important information (I)

## Grading

- ► Coursework 1: 10%
- Coursework 2: 10%
- ► In-person Examination: 80%

About coursework 1: will post on the module webpage.

- o Every relevant information will be posted in the module webpage.
- o Contact me via emails. I will respond within 2 days (if I'm not on travel).

# Some important information (II)

- ▶ you are having problems with tabula ⇒ email DCS.UG.Support@warwick.ac.uk
- Only exception: You wish to change your seminar group (for valid reason). Then write an email to DCS.UG.Support@warwick.ac.uk and cc me. In the email, explain your reason, and specify the groups you can join (see module webpage for group numbers).
- o valid reason: time conflicts with other lectures/seminars

## Recall: Induction and recursion

- ▶ induction: **proving** some universal statements from a smaller objects
- recursion: applied to **definition** in terms of smaller objects

# Example

Let  $n \in \mathbb{N}$ , if  $n \geq 4$ , we have  $2^n \geq n^2$ .

#### Recall: Induction and recursion

- ▶ induction: **proving** some universal statements from a smaller objects
- recursion: applied to definition in terms of smaller objects

# Example

Let  $n \in \mathbb{N}$ , if  $n \ge 4$ , we have  $2^n \ge n^2$ .

## Proof by induction.

Base case, let n=4, we have  $2^n=n^2=16$ .

For the induction step, assume  $2^n \ge n^2$ , we need to show  $2^{n+1} \ge (n+1)^2 = n^2 + 2n + 1$ .

$$2^{n+1} = 2 \times 2^n \ge 2n^2 \ge n^2 + 4n \ge n^2 + 2n + 1$$
.



#### Recursion

# Definition (Recursion)

A problem solving technique in which problems are solved by reducing them to **smaller problems** of the same form.

# Example (factorial)

$$1! = 1 \tag{1}$$

$$n! = n \cdot (n-1)! \tag{2}$$

$$6! = 6 \cdot 5! = 6 \cdot 5 \cdot 4! = 6 \cdot 5 \cdot 4 \cdot 3! = 6 \cdot 5 \cdot 4 \cdot 3 \cdot 2! = 6 \cdot 5 \cdot 4 \cdot 3 \cdot 2 \cdot 1! = 6 \cdot 5 \cdot 4 \cdot 3 \cdot 2 \cdot 1$$

# **Example: Frog jumping problem**

## Problem

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How many ways can a frog hop up a twelve-step staircase if the frog can hop either one or two steps on each hop?

## Solution

Denote f(n) as the number of ways that a frog hops up to the n-th stair. Clearly, f(1)=1 and f(2)=1+1=2.

for the n-th stair, there is only two ways to reach

- ▶ hop from the (n-1)-th stair
- ▶ hop from the (n-2)-th stair

$$\Rightarrow f(n)=f(n-1)+f(n-2)\,.\quad \textit{with } f(1)=1, f(2)=2.$$

This is a Fibonacci sequence.

#### References |

- [0] Sitan Chen, Adam R Klivans, and Raghu Meka, Learning deep relu networks is fixed-parameter tractable, 2021 IEEE 62nd Annual Symposium on Foundations of Computer Science (FOCS), IEEE, 2022, pp. 696–707. (Cited on pages 10 and 11.)
- [0] Ian J Goodfellow, Jonathon Shlens, and Christian Szegedy, *Explaining and harnessing adversarial examples*, International Conference on Learning Representations, 2015. (Cited on page 7.)