



Computational Complexity Theory

Lecture I: Course overview; Turing machines

Department of Computer Science,
Indian Institute of Science

Course overview

About the course

- Computational complexity attempts to classify computational **problems** based on the amount of **resources** required by **algorithms** to solve them.

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 - a. **Decision problem**

Example: i) Is vertex **t** reachable from vertex **s** in graph **G**?

ii) Is **n** a prime number?

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- Computational **problems** come in various flavors:
 - a. **Decision problem**
 - b. **Search problem**

Example: i) Find a satisfying assignment for a Boolean formula.
ii) Find a prime between n and $2n$.

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- Computational **problems** come in various flavors:
 - a. Decision problem
 - b. Search problem
 - c. Counting problem

Example: i) Count the number of cycles in a graph.

ii) Count the number of perfect matchings in a graph.

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 - a. Decision problem
 - b. Search problem
 - c. Counting problem
 - d. Optimization problem

Example: i) Find a minimum size vertex cover in a graph.
ii) Optimize a linear function subject to linear inequality constraints. (*linear programming*)

About the course

- Computational complexity attempts to classify computational **problems** based on the amount of **resources** required by **algorithms** to solve them.
- **Algorithms** are methods for solving problems; they are studied using formal models of computation, like **Turing machines**.



- a **memory** with head (like a RAM)
- a **finite control** (like a processor)

(...more later in this lecture)

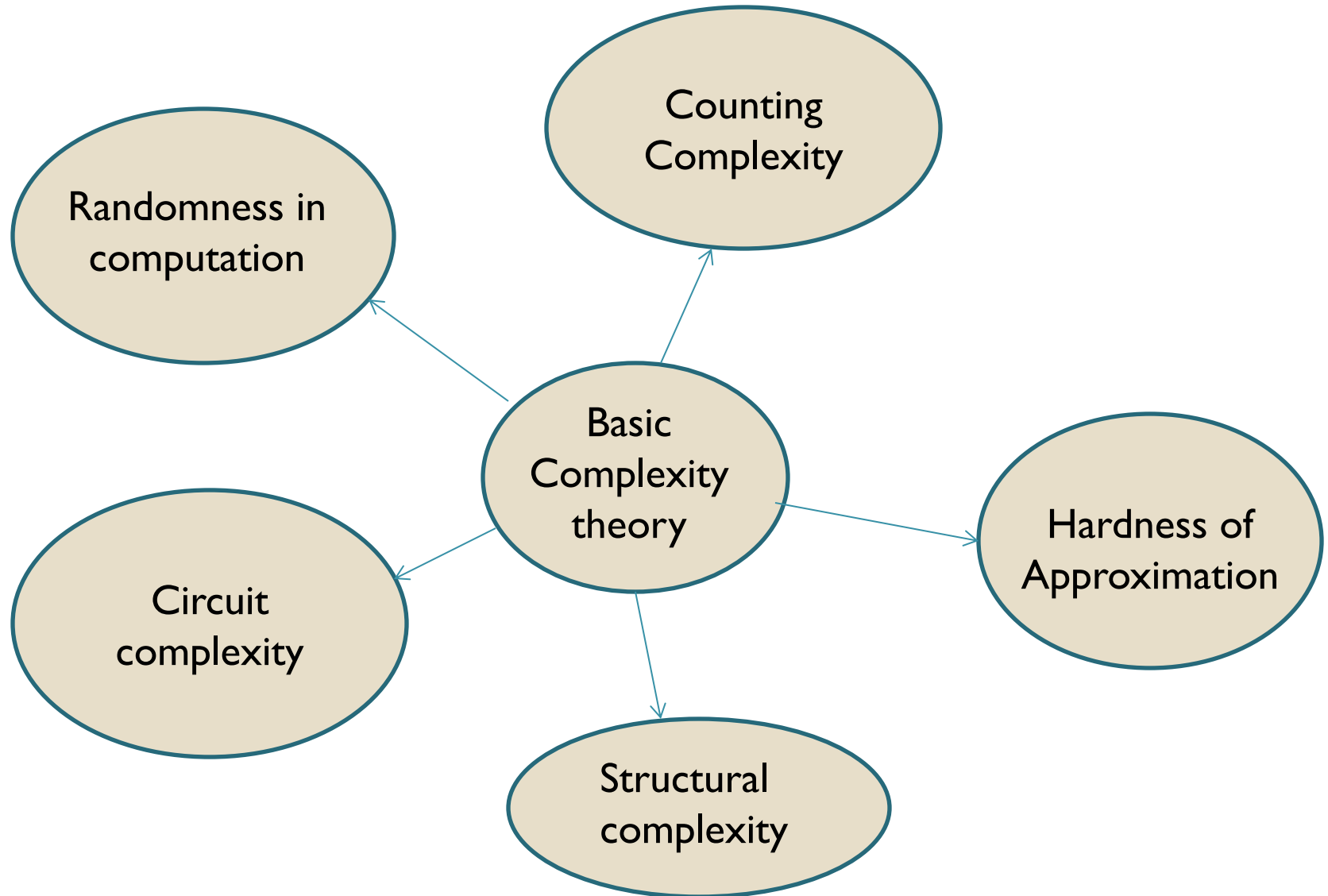
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- Computational complexity attempts to classify computational **problems** based on the amount of **resources** required by **algorithms** to solve them.
- Computational **resources** (required by models of computation) can be:
 - **Time** (bit operations)
 - **Space** (memory cells)

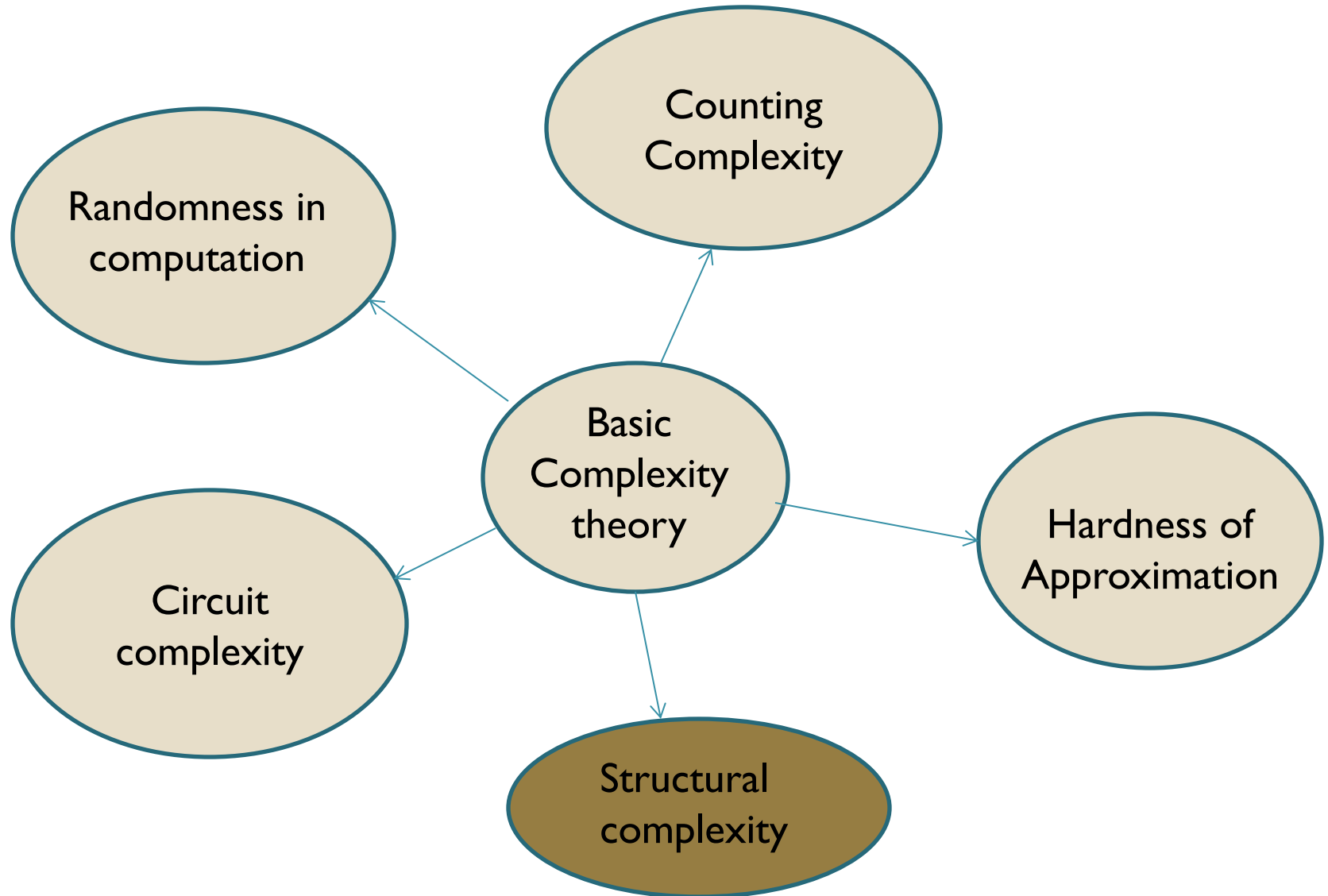
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- Computational **resources** (required by models of computation) can be:
 - **Time** (bit operations)
 - **Space** (memory cells)
 - **Random bits** (magic bits: **0** w.p $\frac{1}{2}$ and **1** w.p $\frac{1}{2}$)
 - **Communication** (bit exchanges)

Topics to be covered in this course



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Structural Complexity: Overview

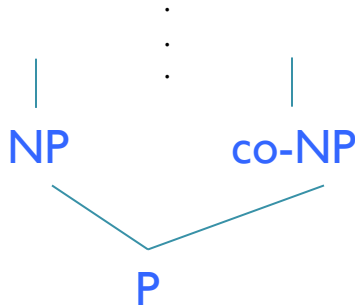
- Classes P , NP , $co-NP$... NP -completeness.
 - How hard is it to check **satisfiability** of a Boolean formula?
 - What if the formula has **exactly one or no** satisfying assignment?

Structural Complexity: Overview

- Classes **P**, **NP**, **co-NP**... **NP-completeness**.
- Space bounded computation.
 - How much **space** is required to check **s-t connectivity**?

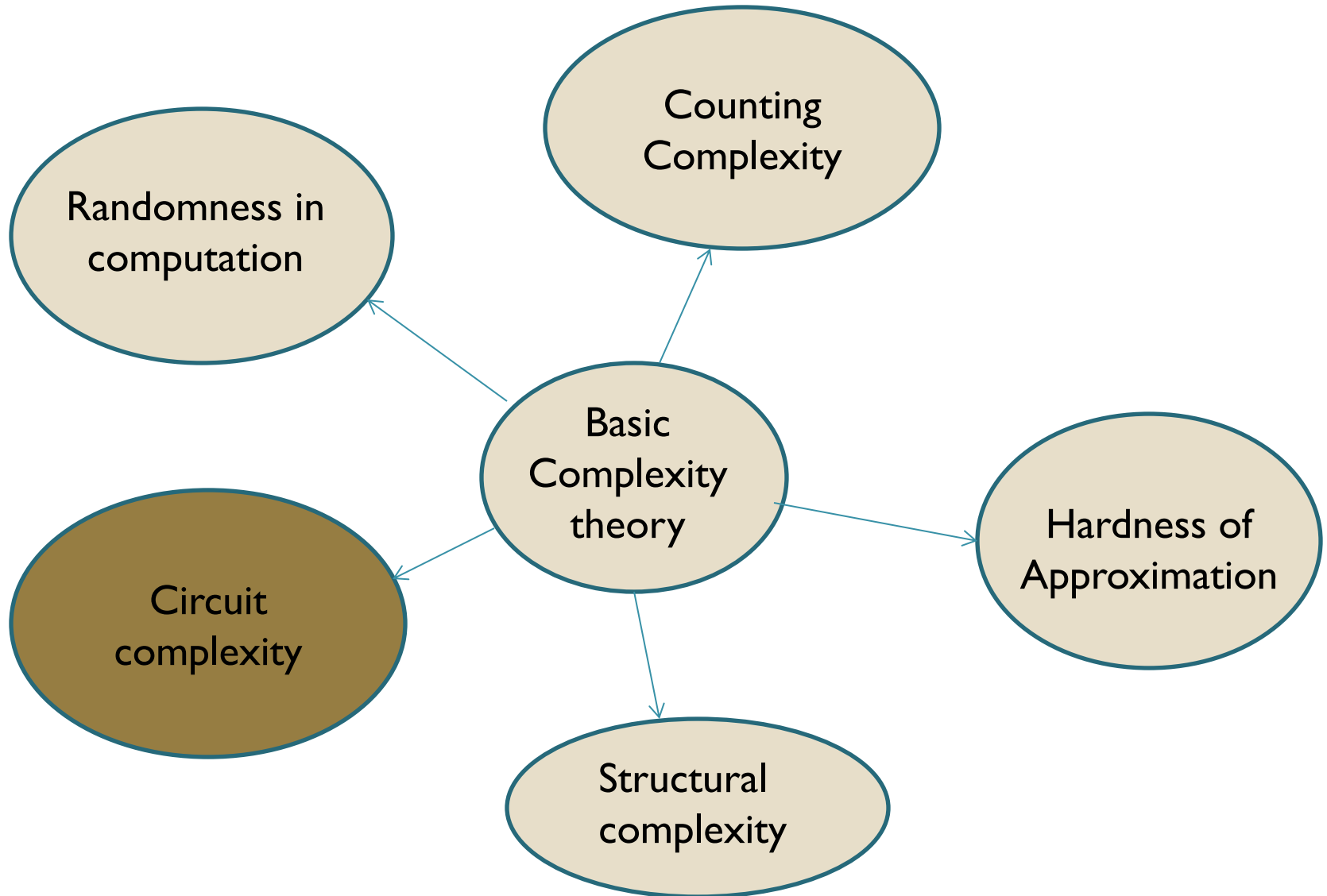
Structural Complexity: Overview

- Classes P , NP , $co-NP$... NP -completeness.
- Space bounded computation.
- Polynomial Hierarchy (PH).



- How hard is it to check if the largest independent set in G has size k ?
- How hard is it to check if there is a circuit of size k that computes the same Boolean function as a given Boolean circuit C ?

Topics to be covered in this course



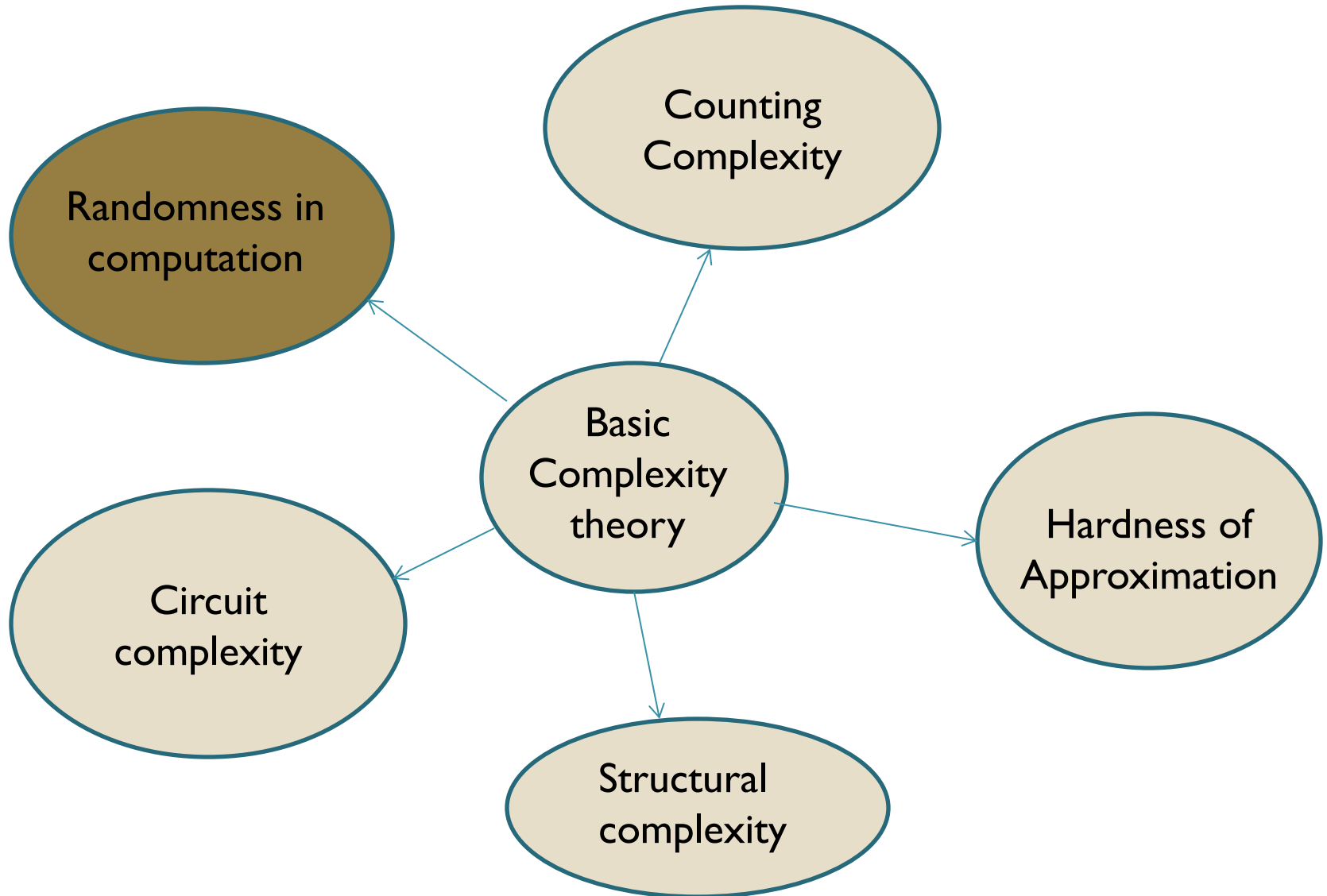
Circuit Complexity: Overview

- The internal workings of an algorithm can be viewed as a **Boolean circuit** -- a nice combinatorial model of computation that is closely related to Turing Machines.
- The size, depth & width of a circuit correspond to the sequential, parallel & space complexity, respectively, of the algorithm that it represents.

Circuit Complexity: Overview

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- The size, depth & width of a circuit correspond to the sequential, parallel & space complexity, respectively, of the algorithm that it represents.
- Proving **P** $\neq$ **NP** reduces to showing circuit lower bounds.
 - We will see lower bounds for restricted classes of circuits.

Topics to be covered in this course



Randomness in Computation: Overview

- Probabilistic complexity classes (BPP, RP, co-RP).
 - Does randomization help in improving efficiency?
 - Quicksort has $O(n \log n)$ expected time but $O(n^2)$ worst case time.
 - Can SAT be solved in polynomial time using randomness?

Theorem (Schoening, 1999): 3SAT can be solved in randomized $O((4/3)^n)$ time.

(brute force takes 2^n time)

(best randomized algorithm for 3SAT: $\sim O(1.307^n)$ time.)

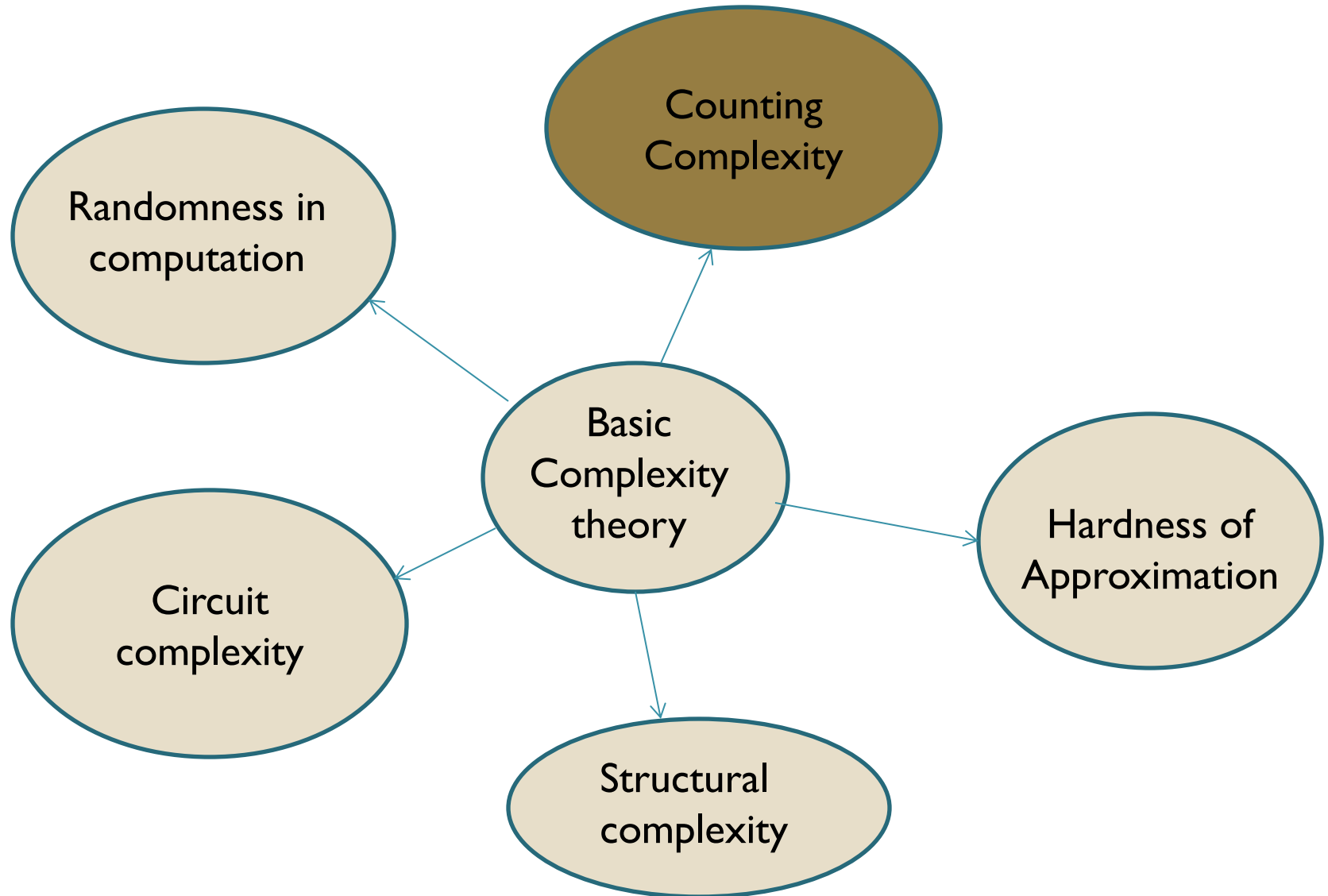
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- Access to random bits can help improve computational efficiency... but, to what extent?

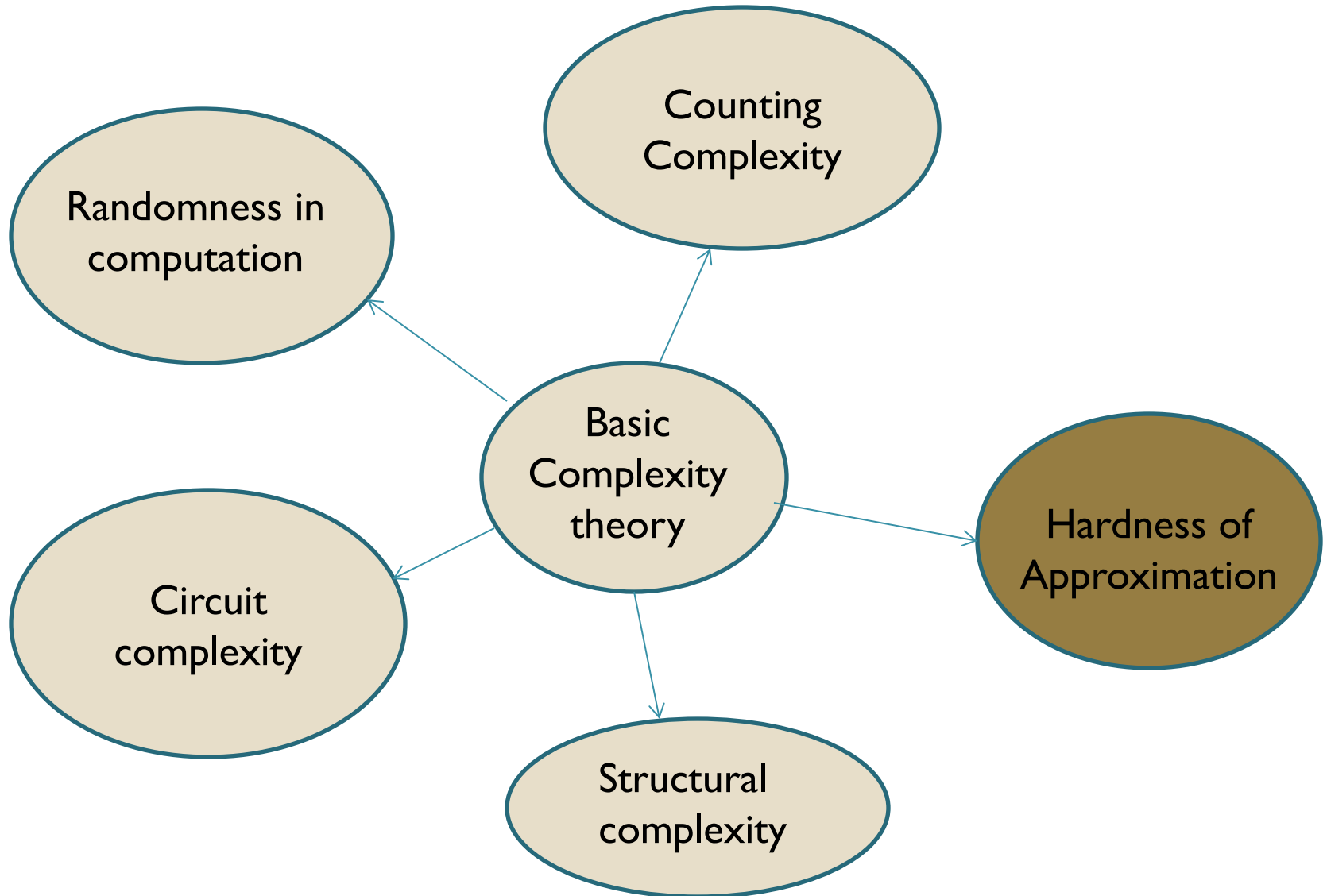
Topics to be covered in this course



Counting Complexity: Overview

- Counting complexity classes (class $\#P$).
 - How hard is it to count the number of perfect matchings in a graph?
 - How hard is it to count the number of cycles in a graph?
 - Can we compute the number of simple paths between s and t in G efficiently?
 - Is counting much harder than the corresponding decision problem?

Topics to be covered in this course



Hardness of Approximation: Overview

- Probabilistically Checkable Proofs (PCPs).

Hardness of approximation results.

Theorem (Hastad, 1997): If there's a poly-time algorithm to compute an assignment that satisfies at least $7/8 + \epsilon$ fraction of the clauses of an input 3SAT, for any constant $\epsilon > 0$, then $P = NP$.

- In contrast, there is a poly-time algorithm to compute an assignment that satisfies at least $7/8$ fraction of the clauses.

Course Info

- **Course no.:** E0 224 **Credits:** 3:1
- **Instructor:** Chandan Saha
- **Lecture time:** M,W 11:30-1pm. **Venue:** CSA 252

- **Course homepage:**

<https://www.csa.iisc.ac.in/~chandan/courses/complexity25/home.html>

Course Info

- **Prerequisites:** Basic familiarity with algorithms; Mathematical maturity.
- **Primary reference:** [Computational Complexity – A Modern Approach](#) by Sanjeev Arora and Boaz Barak.
- **Lectures:** Slides will be posted on the course homepage.
- **Number of lectures:** ~30.

Course Info

- **Grading policy:** Three assignments - 45%
Midterm exam - 25%
Final exam - 30%

Assignments

- **First assignment:** Will posted on Aug 31; due date will be Sep 14.
- **Second assignment:** Will posted on Sep 30; due date will be Oct 14.
- **Third assignment:** Will posted on Oct 31; due date will be Nov 14. (Last day of class is also Nov 14.)
- **Mode:** Assignments will be posted on the course homepage. You need to e-mail me your assignment as a pdf file (use Latex).

Midterm exam

- Would be a 3 hr long written test.
- **When?** Likely in the last week of Sep.

Final exam

- Would be a 3 hr long written test.
- **When?** Likely in the last week of Nov.

Turing Machines

Turing Machines

- An algorithm is a set of instructions or rules.
- To understand the performance of an algorithm we need a model of computation. Turing machine is one such *natural* model (introduced by Alan Turing in 1936).
- Turing called it an “*a-machine*” (automatic machine). His doctoral advisor Alonzo Church coined the term “Turing machine”.

Turing Machines

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- A TM consists of:
 - Memory tape(s)
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- Turing machines  A mathematical way to describe algorithms.

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(e.g. of a physical realization of a TM is a simple adder)

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 has a **blank** symbol

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- δ is a function from $Q \times \Gamma^k$ to $Q \times \Gamma^k \times \{L, S, R\}^k$

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known as **transition function**; it captures the dynamics of M

Turing Machines: Computation

- Start configuration.
 - All tapes other than the input tape contain blank symbols.
 - The input tape contains the input string.
 - All the head positions are at the start of the tapes.
 - The machine is in the start state q_{start} .

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- Computation.

- A **step of computation** is performed by applying δ .

- Halting.

- Once the machine enters q_{halt} it stops computation.

Turing Machines: Running time

- Let $f: \{0,1\}^* \rightarrow \{0,1\}^*$ and $T: \mathbb{N} \rightarrow \mathbb{N}$ and M be a Turing machine.
- **Definition.** We say M **computes** f if on every x in $\{0,1\}^*$, M halts with $f(x)$ on its output tape beginning from the start configuration with x on its input tape.

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- **Definition.** M computes f in $T(|x|)$ **time**, if for every x in $\{0,1\}^*$, M halts within $T(|x|)$ steps of computation and outputs $f(x)$.

Turing Machines

- In this course, we would be dealing with
 - Turing machines that halt on every input.
 - Computational problems that can be solved by Turing machines.

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 - Turing machines that halt on every input.
 - Computational problems that can be solved by Turing machines.
- Can every computational problem be solved using Turing machines?

Turing Machines: Uncomputability

- There are problems for which there exists **no** TM that halts on every input instances of the problem and outputs the correct answer.
 - **Input:** A system of polynomial equations in many variables with integer coefficients.
 - **Output:** Check if the system has **integer solutions** .
 - **Question:** Is there an algorithm to solve this problem?

Turing Machines: Uncomputability

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➤ A typical input instance:

$$x^2y + 5y^3 = 3$$

$$x^2 + z^5 - 3y^2 = 0$$

$$y^2 - 4z^6 = 0$$



Integer solutions for x, y, z ?

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(Hilbert's tenth problem, 1900)

Turing Machines: Uncomputability

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 - **Input:** A system of polynomial equations in many variables with integer coefficients.
 - **Output:** Check if the system has **integer solutions** .
 - **Question:** Is there an algorithm to solve this problem?
- **Theorem.** There doesn't exist any algorithm (realizable by a TM) to solve this problem. (Davis, Putnam, Robinson, Matiyasevich 1970)

Why Turing Machines?

- TMs are natural and intuitive.
- **Church-Turing thesis:** *“Every physically realizable computation device – whether it’s based on silicon, DNA, neurons or some other alien technology – can be simulated by a Turing machine”.*
— [quoted from Arora-Barak’s book]

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 - [quoted from Arora-Barak’s book]
- Several other computational models can be simulated by TMs.

Why Turing Machines?

- TMs are natural and intuitive.
- **Strong Church-Turing thesis:** “Every *physically realizable computation device* – whether it’s based on silicon, DNA, neurons or some other alien technology – can be simulated *efficiently* by a Turing machine”.

Possible exception: Quantum machines!

Basic facts about TMs

Turing Machines

- **Time constructible functions.** A function $T: \mathbb{N} \rightarrow \mathbb{N}$ is time constructible if $T(n) \geq n$ and there's a TM that computes the function that maps x to $T(\underbrace{|x|}_{\text{in binary}})$ in $O(T(|x|))$ time.
- Examples: $T(n) = n^2$, or 2^n , or $n \log n$

Turing Machines: Robustness

- Let $f: \{0,1\}^* \rightarrow \{0,1\}^*$ and $T: \mathbb{N} \rightarrow \mathbb{N}$ be a time constructible function.
- Binary alphabets suffice.
 - If a TM M computes f in $T(n)$ time using Γ as the alphabet set, then there's another TM M' that computes f in time $4 \cdot \log |\Gamma| \cdot T(n)$ using $\{0, 1, \text{blank}\}$ as the alphabet set.

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- A single tape suffices.
 - If a TM M computes f in $T(n)$ time using k tapes then there's another TM M' that computes f in time $5k \cdot T(n)^2$ using a single tape that is used for input, work and output.

Turing Machines: As strings

- Every TM can be represented by a finite string over $\{0,1\}$.

...simply encode the description of the TM.

Turing Machines: As strings

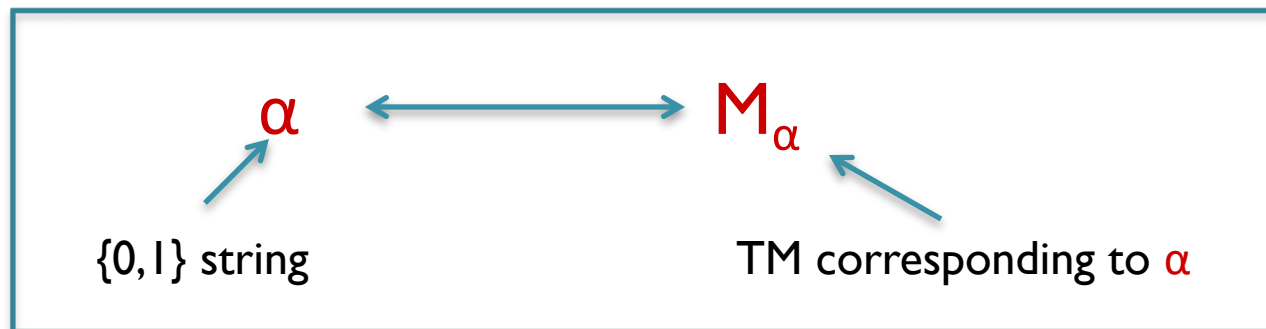
- Every TM can be represented by a finite string over $\{0,1\}$.
- Every string over $\{0,1\}$ represents some TM.
...invalid strings map to a fixed, trivial TM.

Turing Machines: As strings

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- Every string over $\{0,1\}$ represents some TM.
- Every TM has infinitely many string representations.
 - ... allow padding with arbitrary number of 0's

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Turing Machines: As strings

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- Every string over $\{0,1\}$ represents some TM.
- Every TM has infinitely many string representations.
- A TM (i.e., its string representation) can be given as an input to another TM !!

Universal Turing Machines

- **Theorem.** There exists a TM U that on every input x , α in $\{0,1\}^*$ outputs $M_\alpha(x)$.
- Further, if M_α halts within T steps then U halts within $C \cdot T \cdot \log T$ steps, where C is a constant that depends only on M_α 's alphabet size, number of states and number of tapes.

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- Physical realization of UTMs are modern day electronic computers.