



Computational Complexity Theory

Lecture I: Intro; Turing machines

Department of Computer Science,
Indian Institute of Science

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 - a. **Decision problem**

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

Is **n** a prime number?

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 - b. **Search problem**

Example: Find a satisfying assignment for a Boolean formula.
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: Count the number of cycles in a graph.

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: Find a minimum size vertex cover in a graph

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)

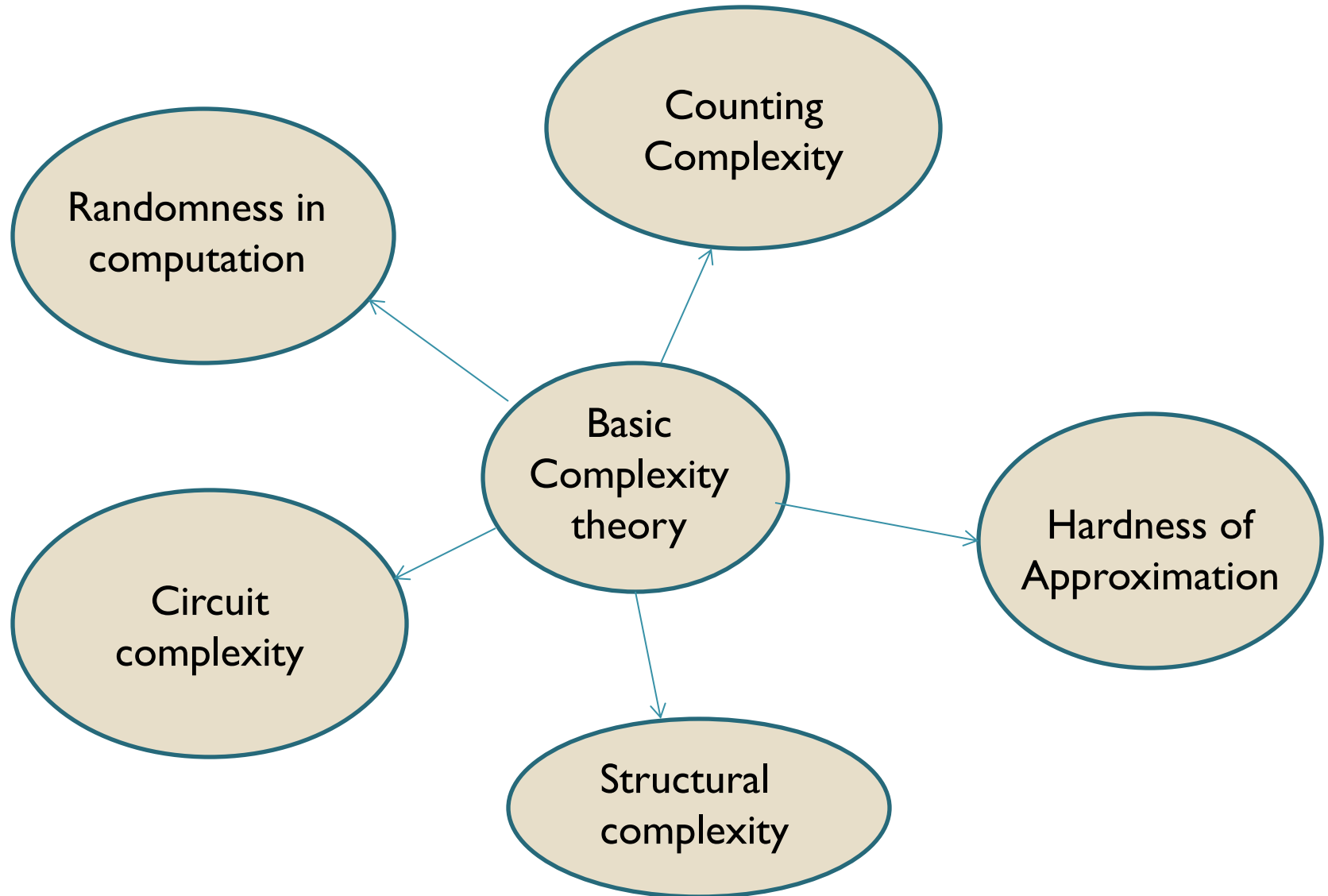
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- Computational complexity attempts to classify computational **problems** based on the amount of **resources** required by **algorithms** to solve them.
- Computational **resources** (required by models of computation) can be:
 - **Time** (bit operations)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

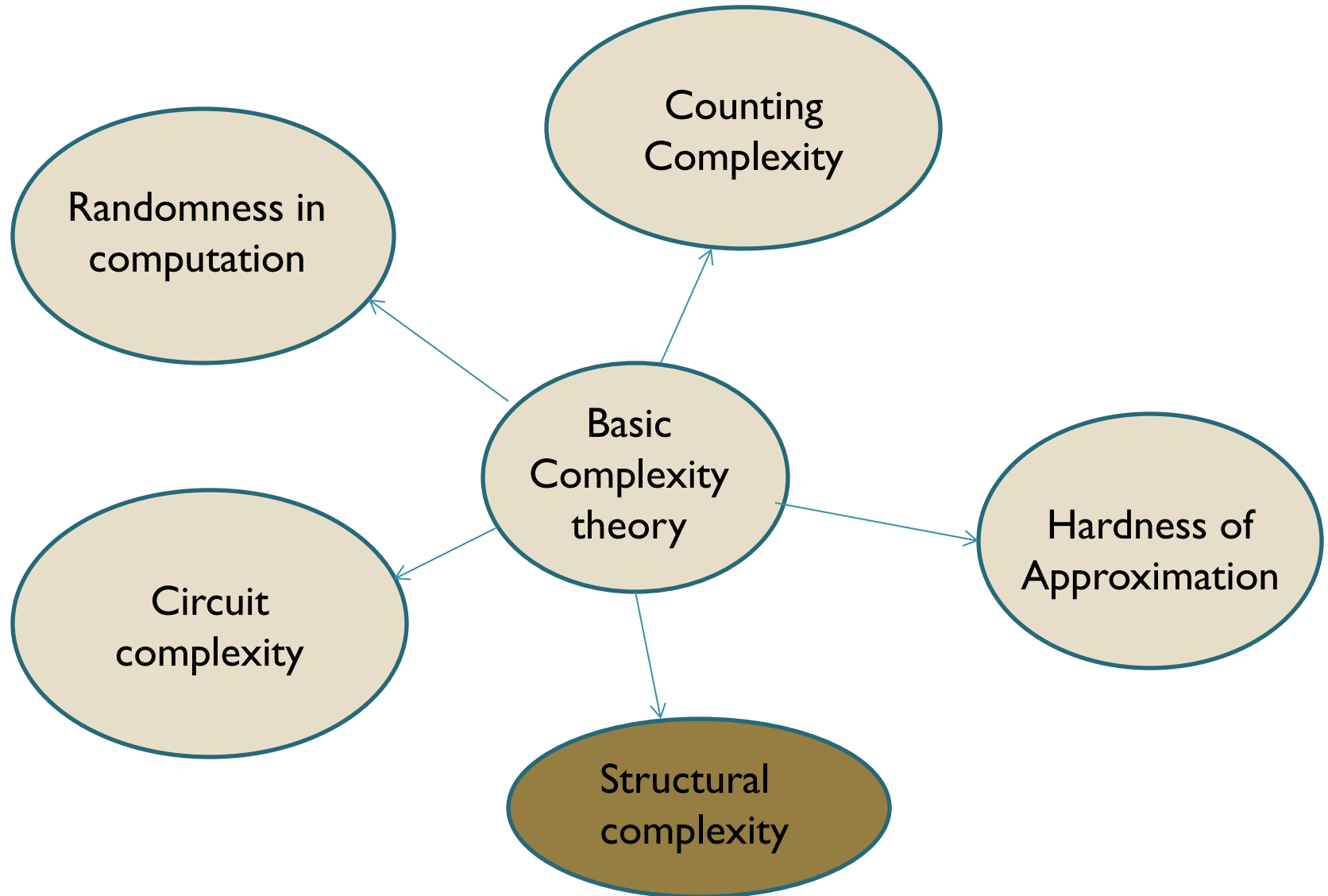
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- Computational **resources** (required by models of computation) can be:
 - **Time** (bit operations)
 - **Space** (memory cells)
 - **Random bits** (magic bits: **0** w.p $\frac{1}{2}$ and **1** w.p $\frac{1}{2}$)
 - **Communication** (bit exchanges)

Topics to be covered in this course



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Structural Complexity

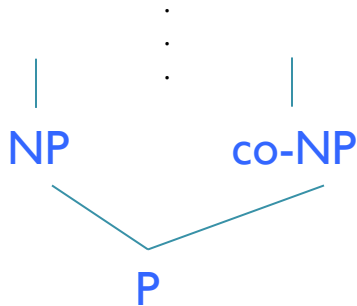
- Classes P, NP, co-NP... NP-completeness.
 - How hard is it to check **satisfiability** of a Boolean formula?
 - What if the formula has **exactly one or no** satisfying assignment?

Structural Complexity

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

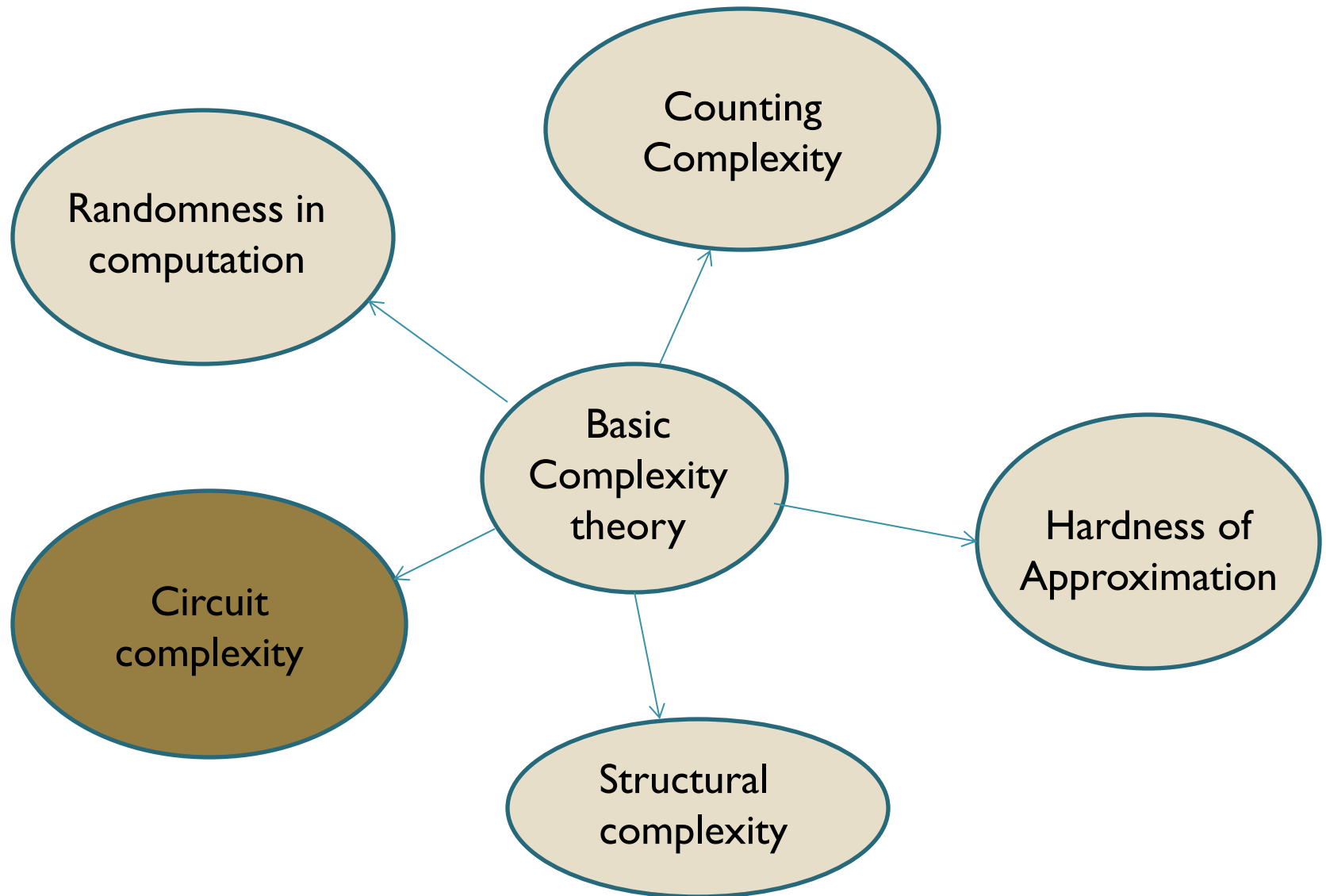
Structural Complexity

- 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



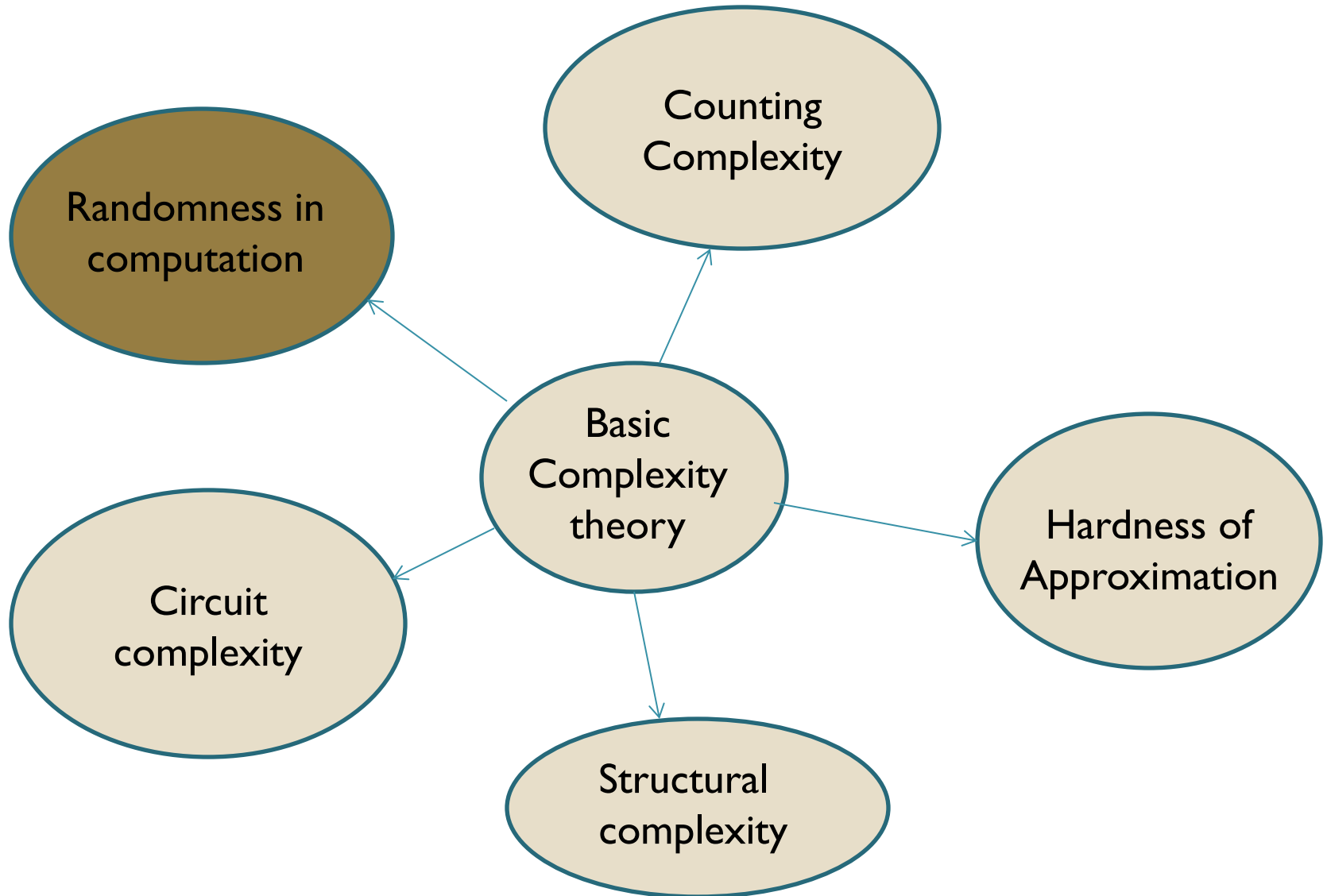
Circuit Complexity

- 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.

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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



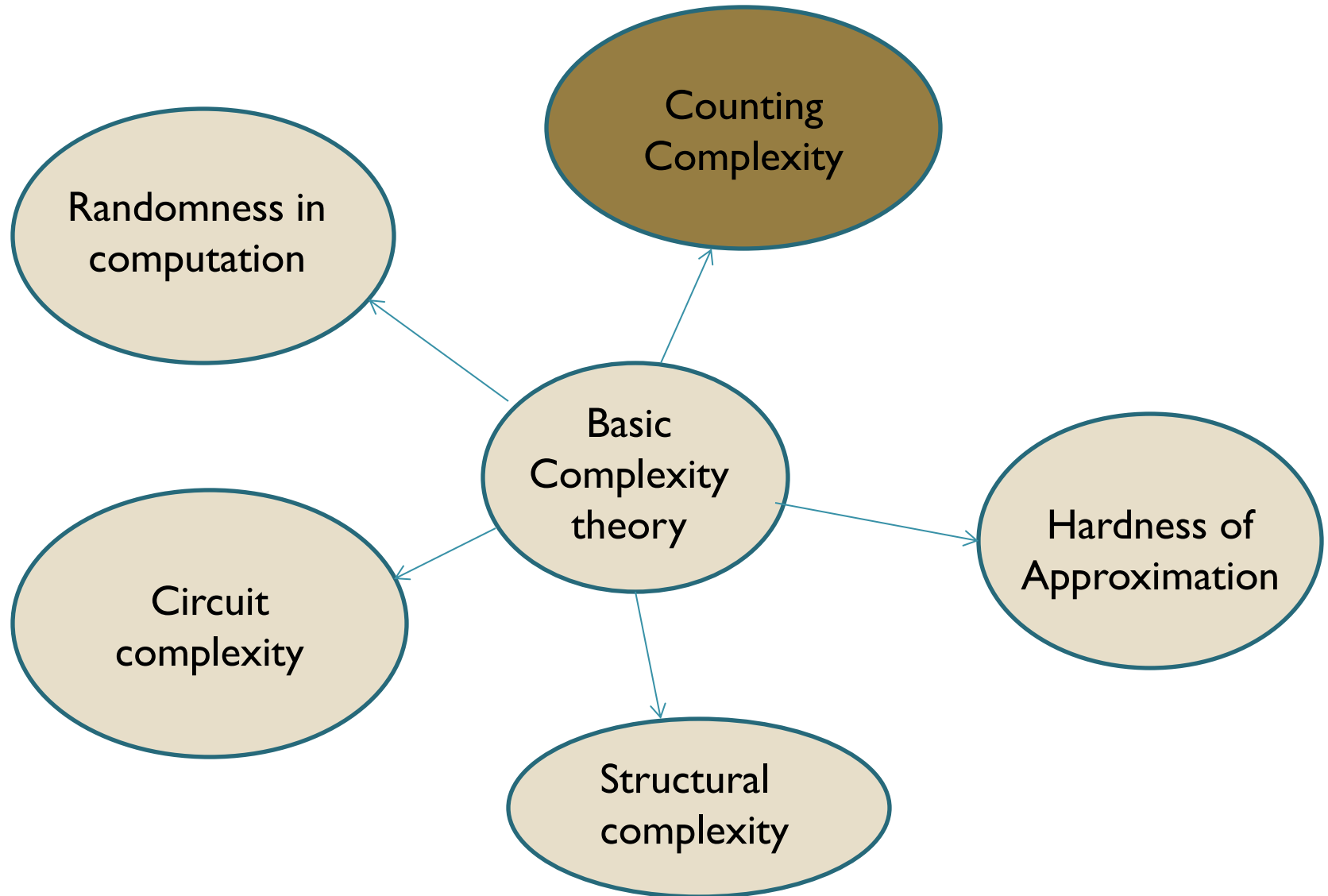
Randomness in Computation

- 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.

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

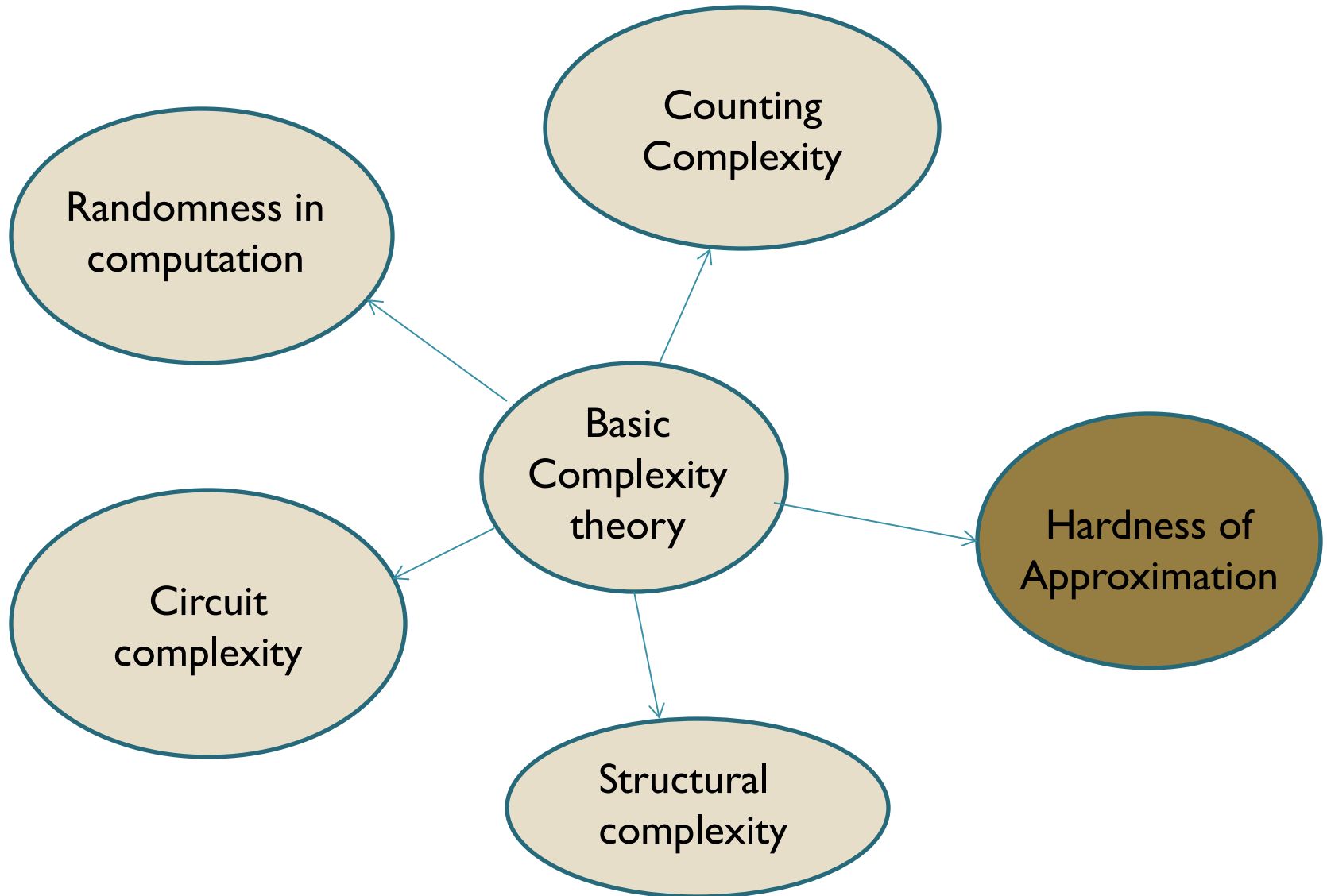
Topics to be covered in this course



Counting Complexity

- 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

- 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 $P = NP$.

Course Info

- **Course no.:** E0 224 **Credits:** 3:1
- **Instructor:** Chandan Saha
- **Lecture time:** M,W 11:30-1pm. **Venue:** CSA 112
- **Course homepage:**

<https://www.csa.iisc.ac.in/~chandan/courses/complexity23/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:** ~27.

Course Info

- **Grading policy:** Three assignments - 45%
Presentation or midsem 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.
- **Mode:** Assignments will be posted on the course homepage. You need to e-mail me your assignment as a pdf file (use Latex).

Presentations

- A group of 2 students would present a paper/result.
 - **Duration of a presentation:** 1-1.5 hr.
 - **Mode:** In class, use slides.
-
- I will start giving topics to present from mid-Sep. All topics will be handed out by mid-Oct.
 - You will get ~4 weeks to prepare a presentation.
 - We will finish all the presentations by Nov 23 (Wed).

Final exam

- Would be a 3 hr long written test.
- **When?** First week of Dec.

First few lectures

- Lecture 1: Today
- Lecture 2: **Saturday (Aug 5)**, 11:30-1, Room 112
- Lecture 3: Monday (Aug 7), 11:30-1, Room 112
- Lecture 4: Wednesday (Aug 9), 11:30-1, Room 112
- **No lectures on Aug 14 (Mon) and Aug 16 (Wed).**
- Lecture 5: **Saturday (Aug 19)**, 11:30-1, Room 112.

Let's begin...


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).

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 - 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 $f(x)$.