# Computational Complexity Theory

Lecture I: Course overview;

Turing machines

Department of Computer Science, Indian Institute of Science

## Course overview

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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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- Computational problems come in various flavors:
  - 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 <u>linear</u> inequality constraints. (linear programming)

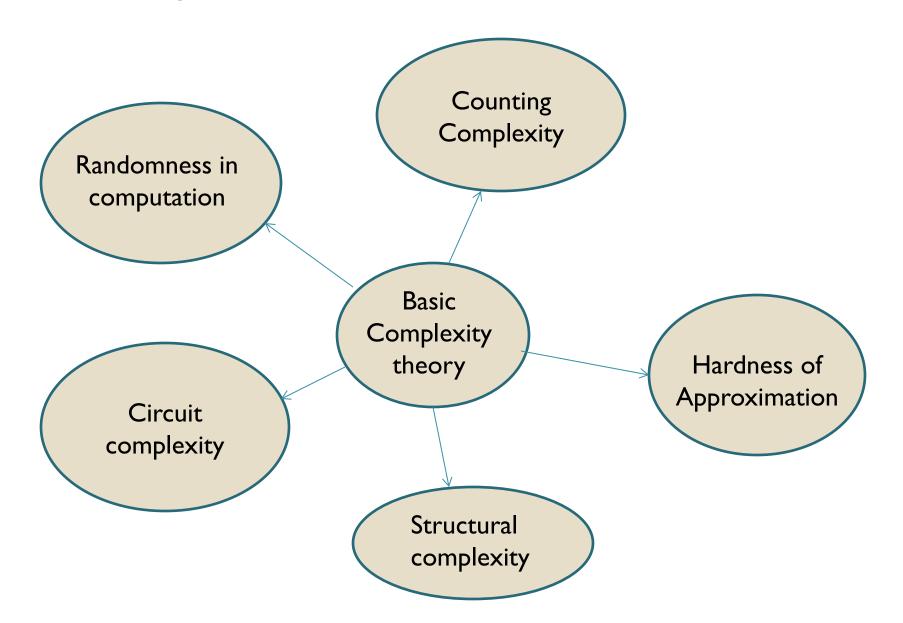
- Computational complexity attempts to classify computational problems based on the amount of resources required by algorithms to solve them.
- Algorithms are <u>methods</u> for solving problems; they are studied using formal <u>models of computation</u>, like <u>Turing machines</u>.
  - **↓**
  - a memory with head (like a RAM)
  - a finite control (like a processor)

(...more later in this lecture)

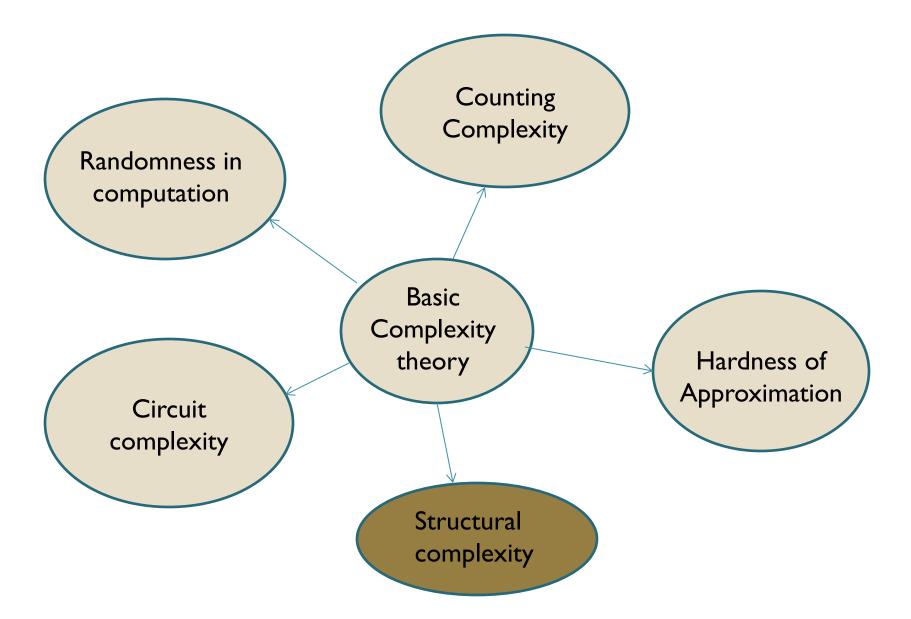
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- Computational resources (required by models of computation) can be:
  - Time (bit operations)
  - Space (memory cells)
  - Random bits (magic bits: 0 w.p ½ and 1 w.p ½)
  - Communication (bit exchanges)

#### Topics to be covered in this course



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## Structural Complexity: Overview

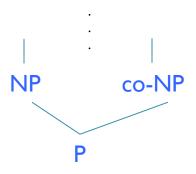
- Classes P, NP, co-NP... NP-completeness.
  - How hard is it to check satisfiability of a Boolean formula?
  - What if the formula has exactly one or no satisfying assignment?

## Structural Complexity: Overview

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

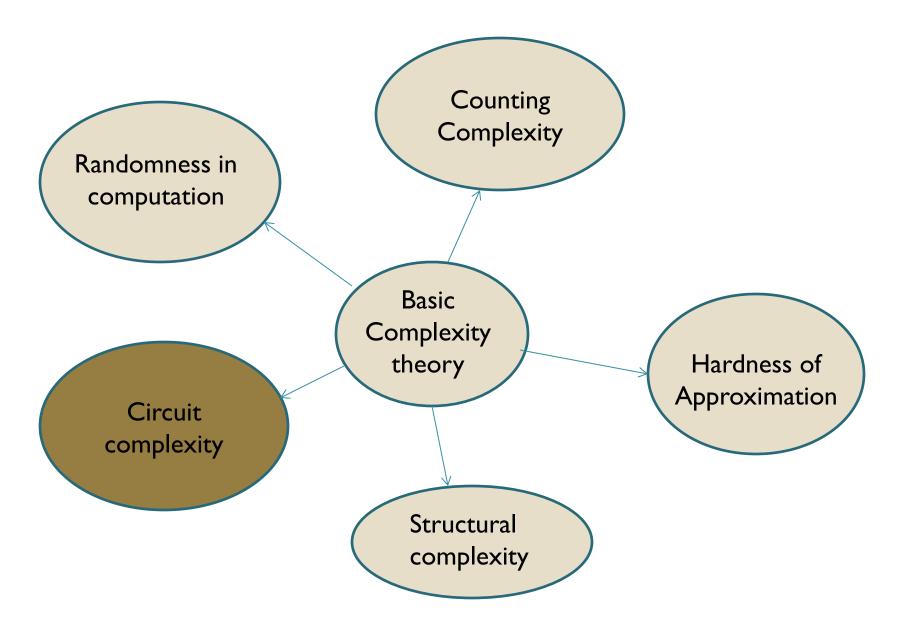
## Structural Complexity: Overview

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



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

#### Topics to be covered in this course



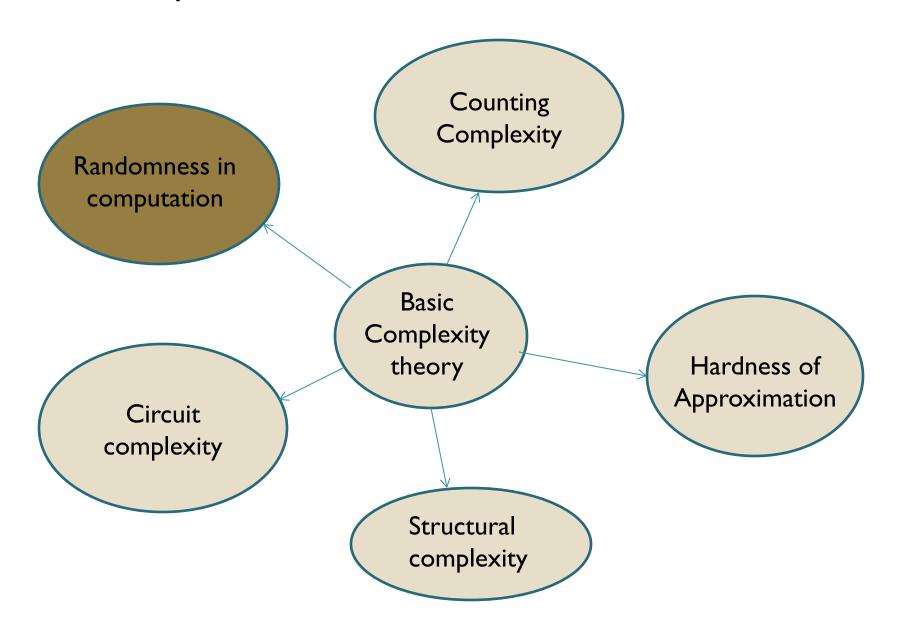
# Circuit Complexity: Overview

- 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 <u>size</u>, <u>depth</u> & <u>width</u> of a circuit correspond to the <u>sequential</u>, <u>parallel</u> & <u>space</u> complexity, respectively, of the algorithm that it represents.

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- Proving P ≠ 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(n²) worst case time.
  - Can SAT be solved in polynomial time using randomness?

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

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(brute force takes 2^n time) (best randomized algorithm for 3SAT: \sim O(1.307^n) time.)
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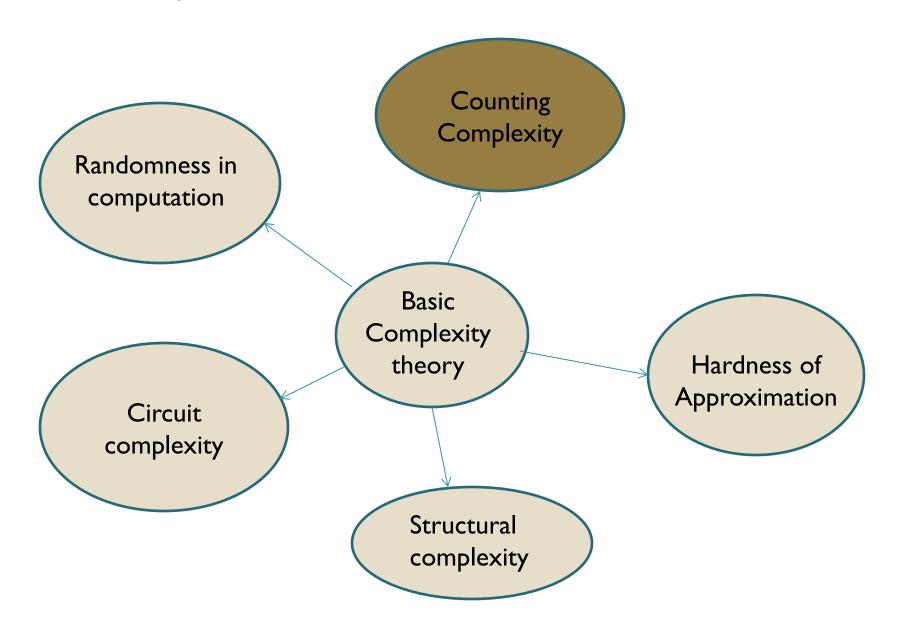
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 Access to random bits can help improve computational efficiency... but, to what extent?

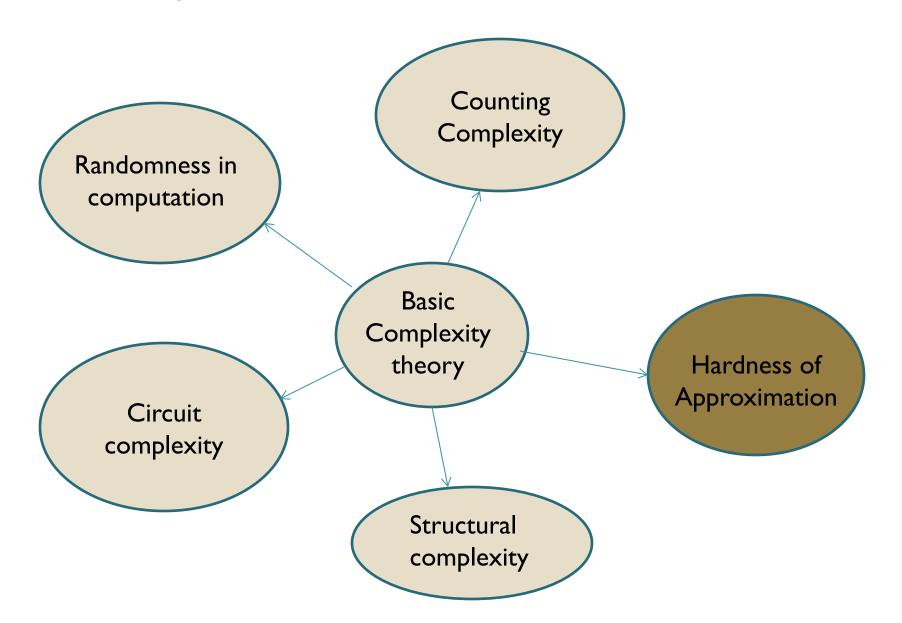
#### Topics to be covered in this course



# 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  $7/8 + \varepsilon$  fraction of the clauses of an input 3SAT, for any constant  $\varepsilon > 0$ , 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 112
- TA: Agrim Dewan (agrimdewan@iisc.ac.in)

Course homepage:

https://www.csa.iisc.ac.in/~chandan/courses/complexity24/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%

Presentation - 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).

### Presentations

- A group of 2 students would present a paper/result.
- Duration of a presentation: ~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 16 (Sat).

### Final exam

Would be a 3 hr long written test.

• When? Likely in the last week of Nov.

- An algorithm is a set of instructions or rules.
- To understand the performance of an algorithm we need a <u>model of computation</u>. 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".

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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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- Q is a finite set of states. (special states: q<sub>start</sub>, q<sub>halt</sub>)
   δ is a function from Q x Γ<sup>k</sup> to Q x Γ<sup>k</sup> {L,S,R}<sup>k</sup>

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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.
  - $\triangleright$  The machine is in the start state  $q_{start}$ .

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- Computation.
  - $\triangleright$  A **step of computation** is performed by applying  $\delta$ .
- Halting.
  - $\triangleright$  Once the machine enters  $q_{halt}$  it stops computation.

#### Turing Machines: Running time

- Let f:  $\{0,1\}^* \rightarrow \{0,1\}^*$  and T: N  $\rightarrow$  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).

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  - Turing machines that halt on every input.
  - Computational problems that can be solved by Turing machines.

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 Can every computational problem be solved using Turing machines?

- 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?

- There are problems for which there exists no TM that halts on every input instances of the problem and outputs the correct answer.
  - > A typical input instance:

$$x^{2}y + 5y^{3} = 3$$
  
 $x^{2} + z^{5} - 3y^{2} = 0$   
 $y^{2} - 4z^{6} = 0$ 

Integer solutions for x, y, z?

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(Hilbert's tenth problem, 1900)

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  - Input: A system of polynomial equations in many variables with integer coefficients.
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  - 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

• Time constructible functions. A function T:  $N \rightarrow N$  is <u>time constructible</u> if  $T(n) \ge n$  and there's a TM that computes the function that maps x to T(|x|) 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: N  $\rightarrow$  N be a time constructible function.
- Binary alphabets suffice.
  - If a TM M computes f in T(n) time using  $\Gamma$  as the alphabet set, then there's another TM M' that computes f in time  $4.\log |\Gamma| . T(n)$  using  $\{0, 1, blank\}$  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.

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

...simply encode the description of the TM.

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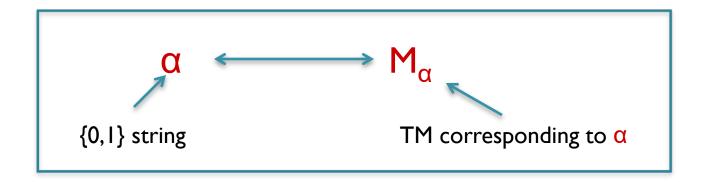
Every string over {0, I} represents some TM.
 ...invalid strings map to a fixed, trivial TM.

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

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

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- Physical realization of UTMs are modern day electronic computers.