

# Midterm Announcements

- Topics on midterm:
  - Beginning of course through Monday 2/19 lecture
  - ⇒ Includes Turing reductions, but not the type where you need to construct another machine
- Midterms from previous terms have been released. Format will be similar, but see HW and discussion worksheets to review all topics
- You may bring one double-sided 8.5 x 11 study sheet, that you prepare
- Review of past exams today 6-8pm BBB 1670 with Eric K.
- Wednesday 3/6:
  - No lecture
  - Office hours end at 5pm
  - Midterm 7-9pm

# Techniques/concepts

## Algorithmic techniques

- Potential method
- Divide-and-Conquer + Master Theorem
- Dynamic Programming
- Greedy + Induction/Exchange

## Models of Computation:

- DFAs
- Turing machines + Church-Turing thesis
- Terminology: countable vs uncountable, language, (un)decidable

## Techniques for proving undecidability

- Diagonalization/paradox
- Reduction

# Reminder of problems + algorithms from class

- **Potential method:** GCD (Euclid)
- **Divide-and-conquer:** sorting (mergesort), closest pair, integer multiplication (Karatsuba)
- **Dynamic programming:** weighted task selection, LIS, LCS, knapsack, SSSP (Bellman-Ford), APSP (Floyd-Warshall)
- **Greedy:** unweighted task selection, MST (Kruskal)
- **Countable vs uncountable sets:** integers, rationals, reals, TMs, TM inputs, languages
- **Undecidable languages:**  $L_{\text{BARBER}}$ ,  $L_{\text{ACC}}$ ,  $L_{\text{HALT}}$

Some reference slides copied from  
past lectures

# Potential Method

Intuitively, a **potential function argument** says:  
If I start with a finite amount of water in a leaky bucket, then eventually water must stop leaking out.



## Ingredients of the argument:

1. Define the “unit of time” e.g. one iteration of an algorithm
2. Define how we measure the amount of water in the bucket. This is the **potential function  $S_i$**  ← amount of water in bucket at timestep  $i$
3. Prove that the  **$S_0$**  is finite and  **$S_i$**  can never be negative
4. Prove that the bucket “leaks quickly”. I.e. that  **$S_i$**  decreases by at least some fixed amount per unit time.
5. Use this to upper bound the total number of units of time.

# Overview: Divide-and-Conquer Algorithms

## Main Idea:

1. **Divide** the input into smaller sub-problems
2. **Conquer:** solve each sub-problem recursively and combine their solutions

## Designing the Algorithm + Proving Correctness: an “art”

- Depends on problem structure, ad-hoc, creative

## Running time Analysis: “mechanical”

- Express runtime using a recurrence
- Can often solve using the “Master Theorem”

# Solving Recurrences

## The Master Theorem

Consider the following recurrence relation where  $k \geq 1$ ,  $b > 1$ ,  $d \geq 0$ ,  $w \geq 0$  are constants (they do not depend on  $n$ ):

$$T(n) = kT(n/b) + O(n^d \log^w n) \\ T(1) = O(1)$$

Then:

$$T(n) = \begin{cases} O(n^d \log^w n) & \text{if } k < b^d \\ O(n^d \log^{w+1} n) & \text{if } k = b^d \\ O(n^{\log_b k}) & \text{if } k > b^d \end{cases}$$

# The DP Recipe

1. Write recurrence ← usually the trickiest part
2. Size of table: How many dimensions? Range of each dimension?
3. What are the base cases?
4. To fill in a cell, which other cells do I look at? In which order do I fill the table?
5. Which cell(s) contain the final answer?
6. Running time = (size of table) • (time to fill each entry)
7. To reconstruct the solution (instead of just its size) follow arrows from final answer to base case



# General strategy commonly used for analyzing greedy algorithms:

Proof by induction using an “**exchange**” argument

**The idea:** Show that we can transform any **optimal solution** into the **solution given by our algorithm** by **exchanging** each piece of it out one-by-one without increasing the cost.

**Key part of proof:** **Exchange** shows that my greedy choice is **safe** i.e. it is in some optimal solution.

Induction formalizes the idea that *each successive choice* is **safe**.

# String notation

**Alphabet:** A nonempty finite set  $\Sigma$  of symbols.

$\Sigma = \{0,1\}$  is a popular choice.

**String:** A finite sequence of 0 or more symbols.

(or “word”)

The empty string is denoted  $\varepsilon$ .

For any  $a \in \Sigma$ :

$a^k$  means  $k$   $a$ 's

$a^*$  means  $\geq 0$   $a$ 's

$a^+$  means  $\geq 1$   $a$ 's

$\Sigma^k$  means all strings over  $\Sigma$  of length  $k$ .

$\Sigma^*$  means **all** (finite) strings over  $\Sigma$ .

$\Sigma^+$  means all nonempty (finite) strings over  $\Sigma$

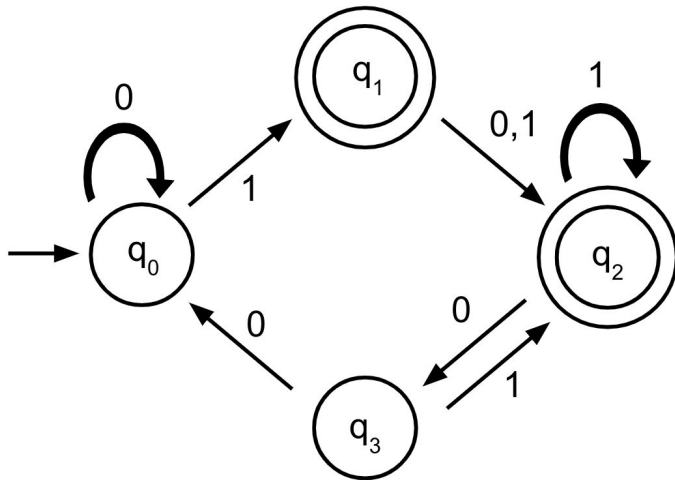
For any  $a, b \in \Sigma$ :  $a|b$  means  $a$  OR  $b$

**Language:** A collection of strings.

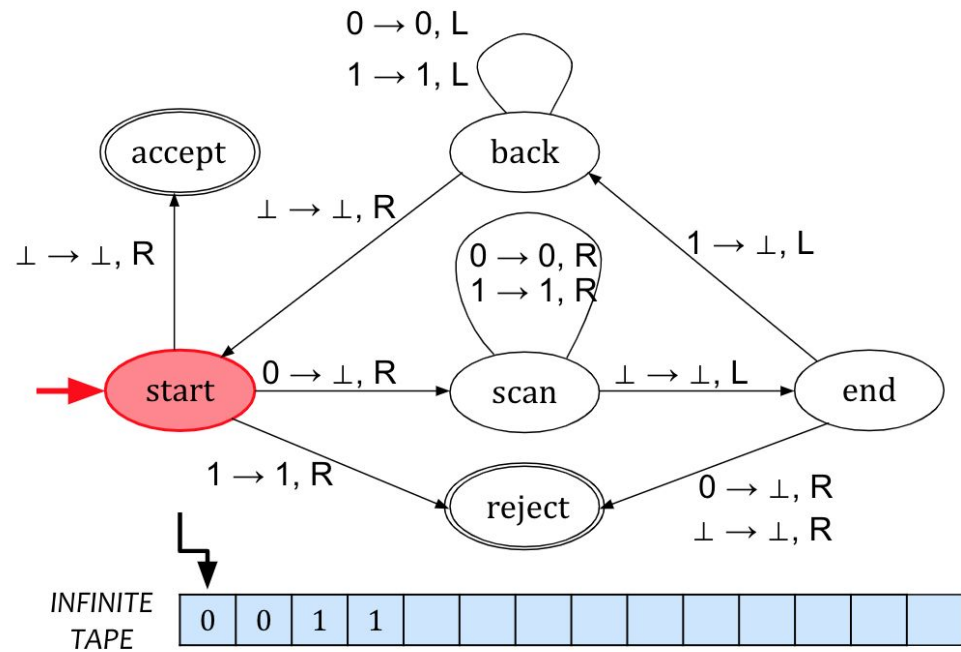
I.e. any subset  $L \subseteq \Sigma^*$ .

The empty language is denoted  $\emptyset$ .

# DFA



# Turing Machine



# Undecidability

**Question:** What are the possible outcomes of a TM **M**?

**Answer:** **M** either (i) accepts, (ii) rejects, or (iii) it “**loops**” (**forever**)

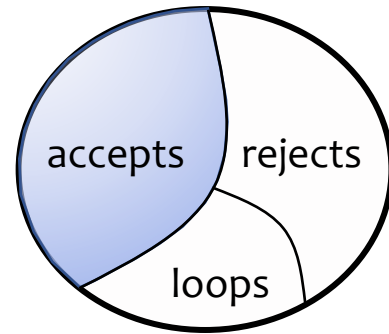
The language of a TM is the set of strings it accepts:

$$L(M) = \{x : M \text{ accepts } x\}$$

**Definition:** A Turing Machine **M** **decides** a language **L** if it:

1. accepts every string in **L**, and
2. rejects every string not in **L**  
(and never loops forever)

A language **L** is **decidable** if there is a TM that decides **L**.  
Otherwise **L** is **undecidable**.



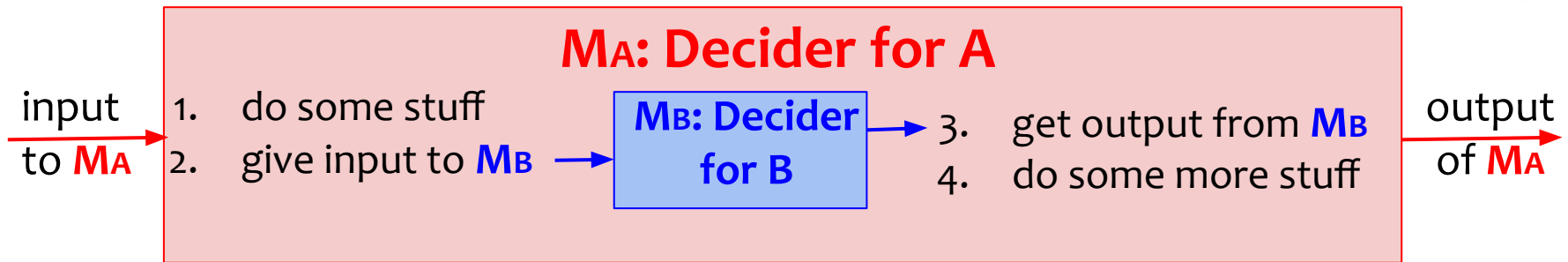
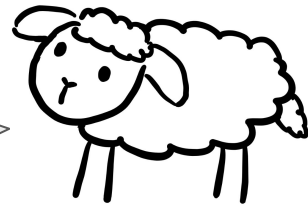
# Turing Reduction from **A** to **B** (denoted $A \leq_T B$ ):

“We can use a black-box decider for **B**  
as a subroutine to decide **A**.”

What it implies:

1. If **B** is decidable then **A** is decidable.
2. Contrapositive: If **A** is undecidable then **B** is undecidable.

You can even invoke  $M_B$   
multiple times inside of  
 $M_A$  if you want



“Problem **B** is at least as hard as Problem **A**”