

# Arithmetic and Incompleteness

Finalizing the computability half of the course

# Theory of Arithmetic

- The theory  $\mathsf{Th}(\mathbb{N})$  of all the facts about the structure of natural numbers is LIFE
- Naturally there is a desire to capture it through a manageable set of axioms
- By manageable I mean finite, or just computable
- By capture I mean axiomatize
- Sadly, this isn't possible (Gödel's Incompleteness Theorem)

#### Peano Axioms

A suggested axiomatization for Th(N)

 From those axioms one can deduce (using a formal proof) many facts about the natural numbers

• ... but not every fact

# Gödel's First Incompleteness

- Within the language of arithmetic, Gödel used his numbering tricks to make sentences speak about themselves (self reference)
- The idea is to create a formula P(x, y) using  $0, +, \times, (, ), s, \rightarrow, \neg, ...$  such that y is the Gödel number of a proof in PA of the sentence whose Gödel number is x
- Look now at this sentence:  $\neg \exists y P(e, y)$  where  $e = gn(\neg \exists y P(e, y))$
- It says e (myself), not provable
- We see (as outsiders to PA) that it is true, but PA does not

#### Gödel's Second Incompleteness

- Gödel decided to play more with his numbering trick and created a sentence that speaks about PA (about the system from within the system)
- The sentence said: PA is consistent
- Consis(PA):  $\neg \exists y P(gn(\neg(0=0)), y)$  (there is no proof of  $0 \neq 0$ )
- In other words, PA cannot prove its own consistency

#### Generalizability of the Incompleteness Theorems

 All those proofs of Gödel just required that the system is powerful enough to express arithmetic

• So, he was able to prove similar facts about, e.g., set theory

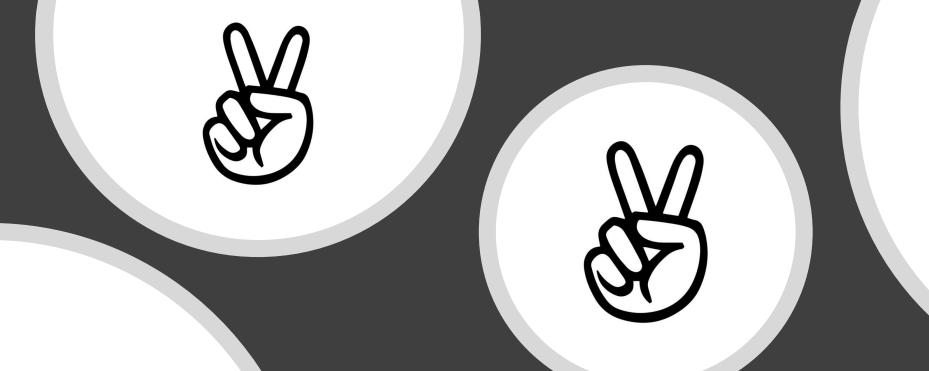
• 
$$0 = \emptyset$$
,  $1 = \{\emptyset\}$ ,  $2 = \{\emptyset, \{\emptyset\}\}\}$ , ...,  $n = \{0, 1, ..., n - 1\}$ 

#### In philosophical terms

 A system which is powerful (enough to describe arithmetic) does not have a decidable list of axioms from which every fact would follow

 Imagine yourself creating a manageable (finite or computable) list of rules (laws) from which everything in your system of interest should follow.

Unless the system is very weak, we can't







Peace out Computability

# Theory of Complexity

Inside what computers can do

#### Computation

- Formality produced models of computation: Turing Machines, Recursive Functions
- Other weaker models with restricted memory: Finite Automata, Pushdown Automata
- Turing Machines are a much more accurate model of a general purpose computer
- Church-Turing thesis connects real-world with theory
- Formality enable us to tell what computers can't do
- Formality made concepts like randomness tangible
- I would like you to take a look at Sipser's book

#### Complexity Analysis

- Formality does not only help us tell what computers can't do, it also allows a general rigorous way to discuss computation resources (time and space)
- What is efficient computation? Turing Machines formalize efficiency and enable measuring it in a standard way
- Time as the number of steps (or transitions). Space as the number of tape cells used.

#### Complexity Measures

- Time Measure:  $t(i, x) = \min\{s: \varphi_{i,s}(x) \downarrow\}$
- Space Measure:

$$M(i,x) = \begin{cases} The \ number \ of \ cells \ visited \ by \ the \ reading \\ head \ while \ computing \ \varphi_{i,s}(x) & \text{if} \ \varphi_{i,s}(x) \downarrow \\ \uparrow & \text{otherwise} \end{cases}$$

- Those are examples of *complexity measures*
- A complexity measure is a more general concept (check Blum Axioms)

# Polynomial Time Computability

- A function *f* is polynomial time computable if:
- 1. There is e such that  $f = \varphi_e$
- 2. There is a polynomial p(n) such that  $t(e, x) \le p(|x|)$  for every binary string input x

• Such a function is called tractable, or efficiently computable

#### We should think in Turing Machine terms

 Since the concepts we are discussing now are mechanical, we switch our terminology from p.c. functions to TMs

 We will work with TMs that halt on all inputs (total). In other words, all our TMs will be deciders

• time(M, x) = The number of steps M takes to accept/reject input x

#### Determinism vs Nondeterminism

- When a TM is in a given state and reads the next input symbol, we know what the next state will be (determined)
- In a nondeterministic machine, several choices may exist for the next state at any point
- Transition function (Deterministic)

$$\delta: Q \times \Gamma \to Q \times \Gamma \times \{L, R\}$$

Transition function (Nondeterministic)

$$\delta \subseteq (Q \times \Gamma) \times (Q \times \Gamma \times \{L, R\})$$

In some references:  $\delta: Q \times \Gamma \to P(Q \times \Gamma \times \{L, R\})$ 

#### Deterministic vs Nondeterministic

- Deterministic is a special case of Nondeterministic
- However, every Nondeterministic TM can be simulated by a Deterministic one (why? Hint: breadth-first search)

#### Time Complexity

• Let M be a deterministic TM. The *running time* or *time complexity* of M is the **function**  $f: \mathbb{N} \to \mathbb{N}$  where f(n) is the maximum number of steps that M uses on any input of size n

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f(n) = \max\{s: M(x) \text{ halts in exactly } s \text{ steps, } |x| = n\}
= \max\{time(M, x): x \in \Sigma^n\}
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- So, for all input strings x of length n, M(x) halts within f(n) steps
- We say M runs in time f(n), or that M is an f(n) time TM

#### Asymptotic Analysis (O notation)

- Running time is often a complex expression
- We usually are only interested in estimating it
- Example: if the running time is  $f(n) = 6 n^3 + 2 n^2 20 n + 45$ , then we describe the running time as  $O(n^3)$
- Generally, we write f(n) = O(g(n)) if  $\exists c \ \exists n_0 \ \forall n \geq n_0, f(n) \leq c \ g(n)$
- g(n) is said to be an asymptotic upper bound

#### Example: The sorting problem

- Input: a sequence of n numbers  $a_1, a_2, ..., a_n$
- Output: a reordering  $a'_1, a'_2, \dots, a'_n$  of  $a_1, a_2, \dots, a_n$  such that  $a'_1 \le a'_2 \le \dots \le a'_n$

#### Idea:

Look at  $a_2$ . If  $\leq a_1$ , move it before  $a_1$ . So we obtain  $a_2, a_1, \ldots, a_n$ . Else, leave the ordering as it is, look at  $a_3$ , and compare it with  $a_2$ 

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This is known as insertion sorting

#### Clarification with numbers:

Input: 5,2,4,6,1,3

- (a) 5,2,4,6,1,3 (At most 1 step)
- (b) 2,5,4,6,1,3 (At most 2 steps)
- (c) 2,4,5,6,1,3 (At most 3 steps)
- (d) 2,4,5,6,1,3
- (e) 1,2,4,5,6,3
- (f) 1,2,3,4,5,6
- Total number of steps in worst-case scenario = 1+2+...+6
- In general, with input of size n, it will be  $\frac{n(n+1)}{2} = O(n^2)$

# Complexity Classes

• For any function  $f: \mathbb{N} \to \mathbb{R}^+$ , and  $n \in \mathbb{N}$ :

$$TIME(f(n)) =$$

 $\{L: L \text{ is a language decidable by some TM that runs in worst case time } O(f(n))\}$ 

$$SPACE(f(n)) =$$

 $\{L: L \text{ is a language decidable by some TM that runs in worst case space } O(f(n))\}$ 

#### The Class P

•  $P = \{L: L \text{ is a language decidable by some polytime TM}\}$ 

• Note that  $P = \bigcup_k TIME(n^k)$ 

# Polytime Reducibility

•  $A \leq_p B$  if  $A \leq_m B$  via an m-reduction f which is polytime

• Fact: If B is decidable in polytime, and  $A \leq_p B$ , then A is also decidable in polytime