### CS 154

Lecture 12:
Foundations of Math and
Kolmogorov Complexity

## Computability and the Foundations of Mathematics

#### The Foundations of Mathematics

A formal system describes a formal language for

- writing (finite) mathematical statements,
- has a definition of what statements are "true"
- has a definition of a proof of a statement

**Example:** Every TM M defines some formal system **F** 

- {Mathematical statements in  $\mathcal{F}$ } =  $\Sigma^*$ String w represents the statement "M accepts w"
- {True statements in F} = L(M)
- A proof that "M accepts w" can be defined to be an accepting computation history for M on w

#### **Consistency and Completeness**

A formal system **F** is **consistent** or **sound** if no false statement has a valid proof in **F** (Proof in **F** implies Truth in **F**)

A formal system F is complete if every true statement has a valid proof in F (Truth in F implies Proof in F)

#### **Interesting Formal Systems**

Define a formal system F to be interesting if:

- 1. Any mathematical statement about computation can be (computably) described as a statement of  $\mathcal{F}$ . Given (M, w), there is a (computable)  $S_{M,w}$  in  $\mathcal{F}$  such that  $S_{M,w}$  is true in  $\mathcal{F}$  if and only if M accepts w.
- 2. Proofs are "convincing" a TM can check that a proof of a theorem is correct

  This set is decidable: {(S, P) | P a proof of S in F}
- 3. If S is in  $\mathcal{F}$  and there is a proof of S describable as a computation, then there's a proof of S in  $\mathcal{F}$ .

  If M accepts w, then there is a proof P in  $\mathcal{F}$  of  $S_{M,w}$

#### **Limitations on Mathematics**

For every consistent and interesting F,

Theorem 1. (Gödel 1931) F is incomplete:

There are mathematical statements in **F** that are true but cannot be proved in **F**.

Theorem 2. (Gödel 1931) The consistency of  $\mathcal{F}$  cannot be proved in  $\mathcal{F}$ .

Theorem 3. (Church-Turing 1936) The problem of checking whether a given statement in F has a proof is undecidable.

#### **Unprovable Truths in Mathematics**

(Gödel) Every consistent interesting F is incomplete: there are true statements that cannot be proved.

Let S<sub>M, w</sub> in F be true if and only if M accepts w

**Proof:** Define Turing machine **G**(x):

- 1. Obtain own description **G** [Recursion Theorem]
- 2. Construct statement  $S' = \neg S_{G, \epsilon}$
- 3. Search for a proof of S' in F over all finite length strings. Accept if a proof is found.

Claim: S' is *true* in  $\mathcal{F}$ , but has no proof in  $\mathcal{F}$  S' basically says "There is no proof of S' in  $\mathcal{F}$ "

(Gödel 1931) The consistency of F cannot be proved within any interesting consistent F

Proof: Suppose we can prove " $\mathcal{F}$  is consistent" in  $\mathcal{F}$ We constructed  $\neg S_{G, \epsilon} =$  "G does not accept  $\epsilon$ "
which we showed is *true*, but *has no proof* in  $\mathcal{F}$ G does not accept  $\epsilon \Leftrightarrow \Gamma$  There is no proof of  $\neg S_{G, \epsilon}$  in  $\mathcal{F}$ 

But if there's a proof in  $\mathcal{F}$  of " $\mathcal{F}$  is consistent" then there is a proof in  $\mathcal{F}$  of  $\neg S_{G, \varepsilon}$  (here's the proof):

"If  $S_{G,\epsilon}$  is true, then there is a proof in  $\mathcal{F}$  of  $\neg S_{G,\epsilon}$ .  $\mathcal{F}$  is consistent, therefore  $\neg S_{G,\epsilon}$  is true. But  $S_{G,\epsilon}$  and  $\neg S_{G,\epsilon}$  cannot both be true. Therefore,  $\neg S_{G,\epsilon}$  is true"

This contradicts the previous theorem.

#### **Undecidability in Mathematics**

PROVABLE<sub>F</sub> = {S | there's a proof in  $\mathcal{F}$  of S, or there's a proof in  $\mathcal{F}$  of  $\neg$ S}

(Church-Turing 1936) For every interesting consistent  $\mathcal{F}$ , PROVABLE, is undecidable

**Proof:** Suppose PROVABLE  $_{\sigma}$  is decidable with TM P.

Then we can decide A<sub>TM</sub> using the following procedure:

On input (M, w), run the TM P on input S<sub>M,w</sub>

If P accepts, examine all possible proofs in F

If a proof of  $S_{M,w}$  is found then accept If a proof of  $-S_{M,w}$  is found then reject

If P rejects, then reject.

Why does this work?

# Kolmogorov Complexity: A Universal Theory of Data Compression

#### The Church-Turing Thesis:

Everyone's
Intuitive Notion = Turing Machines
of Algorithms

This is not a theorem — it is a falsifiable scientific hypothesis.

**A Universal Theory of Computation** 

## Is there a Universal Theory of *Information*?

Can we quantify how much *information* is contained in a string?

A = 010101010101010101010101010101

B = 110010011101110101101001011001011

Idea: The more we can "compress" a string, the less "information" it contains....

#### Information as Description

**Thesis:** The amount of information in a string

= Shortest way of describing that string

How should we "describe" strings?

**Use Turing machines with inputs!** 

Let  $x \in \{0,1\}^*$ 

Definition: The shortest description of x, denoted as d(x), is the lexicographically shortest string <M,w> such that M(w) halts with only x on its tape.

#### **A Specific Pairing Function**

Theorem. There is a 1-1 computable function  $<,>: \Sigma^* \times \Sigma^* \to \Sigma^*$  and computable functions  $\pi_1$  and  $\pi_2: \Sigma^* \to \Sigma^*$  such that:

$$z = \langle M, w \rangle$$
 iff  $\pi_1(z) = M$  and  $\pi_2(z) = w$ 

For  $x_i \in \Sigma$ , let  $Z(x_1 x_2 ... x_k) = 0 x_1 0 x_2 ... 0 x_k 1$ Then we can define:

$$:=Z(M)w$$

(Example: <10110,101> = 01000101001101)

Note that 
$$|| = 2|M| + |w| + 1$$

#### **Kolmogorov Complexity (1960's)**

Definition: The shortest description of x, denoted as d(x), is the lexicographically shortest string <M,w> such that M(w) halts with only x on its tape.

Definition: The Kolmogorov complexity of x, denoted as K(x), is |d(x)|.

#### **EXAMPLES??**

Let's first determine some properties of K. Examples will fall out of this.

#### **Kolmogorov Complexity**

Theorem: There is a fixed c so that for all x in  $\{0,1\}^*$  $K(x) \le |x| + c$ 

"The amount of information in x isn't much more than |x|"

Proof: Define a TM M = "On input w, halt."On any string x, M(x) halts with x on its tape. Let c = 2|M|+1Then  $K(x) \le |\langle M, x \rangle| \le 2|M| + |x| + 1 \le |x| + c$ 

#### Repetitive Strings have Low Information

Theorem: There is a fixed c so that for all  $x \in \{0,1\}^*$  $K(xx) \le K(x) + c$ 

"The information in xx isn't much more than that in x"

Proof: Let N = "On < M, w>, let s = M(w). Print ss."

Suppose <M,w> is the shortest description of x.

Then <N,<M,w>> is a description of xx

**Therefore** 

$$K(xx) \le |\langle N, \langle M, w \rangle \rangle| \le 2|N| + |\langle M, w \rangle| + 1$$
  
  $\le 2|N| + K(x) + 1 \le c + K(x)$ 

#### Repetitive Strings have Low Information

```
Corollary: There is a fixed c so that for all n \ge 2, and all x \in \{0,1\}^*, K(x^n) \le K(x) + c \log n
```

"The information in x" isn't much more than that in x"

```
Proof: Define the TM

N = "On input <n,M,w>,

Let x = M(w). Print x for n times."
```

Let <M,w> be the shortest description of x. Then  $K(x^n) \le K(<N,<n,M,w>>) \le 2|N| + d log n + K(x)$   $\le c log n + K(x)$ for some constant c and d

#### Repetitive Strings have Low Information

Corollary: There is a fixed c so that for all  $n \ge 2$ , and all  $x \in \{0,1\}^*$ ,  $K(x^n) \le K(x) + c \log n$ 

"The information in x" isn't much more than that in x"

#### **Recall:**

A = 010101010101010101010101010101

For  $w = (01)^n$ ,  $K(w) \le K(01) + c \log_2 n$ 

So for all n,  $K((01)^n) \le d + c \log_2 n$  for a fixed c, d

#### **Does The Computational Model Matter?**

Turing machines are one "programming language." If we use other programming languages, could we get significantly shorter descriptions?

An interpreter is a "semi-computable" function

$$p: \Sigma^* \to \Sigma^*$$

Takes programs as input, and (may) print their outputs

Definition: Let  $x \in \{0,1\}^*$ . The shortest description of x under p, called  $d_p(x)$ , is the lexicographically shortest string w for which p(w) = x.

**Definition:** The  $K_p$  complexity of x is  $K_p(x) := |d_p(x)|$ .

#### **Does The Computational Model Matter?**

Theorem: For every interpreter p, there is a integer c so that for all  $x \in \{0,1\}^*$ ,  $K(x) \le K_p(x) + c$ 

Moral: Using another programming language would only change K(x) by some additive constant

Proof: Define M = "On w, simulate p(w) and write its output to tape"

Then  $< M, d_p(x) > is$  a description of x, and

$$K(x) \le |\langle M, d_p(x) \rangle|$$
  
  $\le 2|M| + K_p(x) + 1 \le c + K_p(x)$ 

#### There Exist Incompressible Strings

Theorem: For all n, there is an  $x \in \{0,1\}^n$  such that  $K(x) \ge n$ 

"There are incompressible strings of every length"

Proof: (Number of binary strings of length n) = 2<sup>n</sup> but (Number of descriptions of length < n) ≤ (Number of binary strings of length < n) = 1 + 2 + 4 + ··· + 2<sup>n-1</sup> = 2<sup>n</sup> - 1

Therefore there is at least one n-bit string x that does not have a description of length < n

#### Random Strings Are Incompressible!

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Theorem: For all n and c \ge 1,
\Pr_{x \in \{0,1\}^n}[K(x) \ge n-c] \ge 1 - 1/2^c
```

"Most strings are highly incompressible"

Proof: (Number of binary strings of length n) = 2<sup>n</sup> but (Number of descriptions of length < n-c) ≤ (Number of binary strings of length < n-c) = 2<sup>n-c</sup> - 1

Hence the probability that a  $random \times satisfies$ K(x) < n-c

is at most  $(2^{n-c}-1)/2^n < 1/2^c$ .

Give short algorithms for generating the strings:

- 1. 01000110110000010100111001011101110000
- 2. 123581321345589144233377610987
- 3. 126241207205040403203628803628800

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This seems hard to determine in general. Why? We'll give a formal answer in just one moment...

#### **KOLMOGOROV DIRECTIONS**



WHEN PEOPLE ASK FOR STEP-BY-STEP DIRECTIONS, I WORRY THAT THERE WILL BE TOO MANY STEPS TO REMEMBER, SO I TRY TO PUT THEM IN MINIMAL FORM.

#### **Determining Compressibility**

Can an algorithm perform optimal compression? Can algorithms tell us if a given string is compressible?

COMPRESS =  $\{(x,c) \mid K(x) \le c\}$ 

**Theorem:** COMPRESS is undecidable!

**Intuition:** If decidable, we could design an algorithm that prints the **shortest incompressible string of length n** 

But such a string could then be succinctly described, by providing the algorithm code and n in binary!

Berry Paradox: "The smallest integer that cannot be defined in less than thirteen words."

## Determining Compressibility $COMPRESS = \{(x,c) \mid K(x) \le c\}$

**Theorem:** COMPRESS is undecidable!

**Proof:** Suppose it's decidable. Consider the TM:

M = "On input  $x \in \{0,1\}^*$ , interpret x as a number N. For all  $y \in \{0,1\}^*$  in lexicographical order, If  $(y,N) \notin COMPRESS$  then print y and halt."

M(x) prints the shortest string y' with K(y') > N.

But  $< M, x > describes y', and <math>|< M, x > | \le d + log N$ 

So N <  $K(y') \le d + 2 \log N$ . CONTRADICTION!