

# BILL AND NATHAN RECORD LECTURE!!!!

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**UN-TIMED PART OF  
FINAL IS TUESDAY  
May 11 11:00A.  
NO DEAD CAT**

**FINAL IS THURSDAY**  
**May 13**  
**8:00PM-10:15PM**

**FILL OUT COURSE  
EVALS for ALL YOUR  
COURSES!!!**

# Review for Final

# Rules

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7. **Scope of the Exam**  
**Short Answer** HWs and lectures.  
**Long Answer** This Presentation.

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(Thats a lot!)

# Turing Machines

1. For this review we omit definitions and conventions.
2. There is a JAVA program for function  $f$  iff there is a TM that computes  $f$ .
3. Everything computable can be done by a TM.



# Decidable Sets

**Def** A set  $A$  is DECIDABLE if there is a Turing Machine  $M$  such that

$$x \in A \rightarrow M(x) = Y$$

$$x \notin A \rightarrow M(x) = N$$

# Terrible Def of DTIME

**Def** Let  $T(n)$  be a computable function (think increasing).  $A$  is in  $\text{DTIME}(T(n))$  if there is a TM  $M$  that decides  $A$  and also, for all  $x$ ,  $M(x)$  halts in time  $\leq O(T(|x|))$ .

Terrible Def since depends to much on machine model.

- ▶ Prove theorems about  $\text{DTIME}(T(n))$  where the model does not matter. (Time hierarchy theorem—We did not do this)).
- ▶ Define time classes that are model-independent (P, NP stuff)

# P, NP, Reductions

# P and EXP

## Def

1.  $P = \text{DTIME}(n^{O(1)})$ .
2.  $\text{EXP} = \text{DTIME}(2^{n^{O(1)}})$ .
3. PF is the set of **functions** that are computable in poly time.

# NP

**Def**  $A$  is in NP if there exists a set  $B \in \mathcal{P}$  and a polynomial  $p$  such that

$$A = \{x \mid (\exists y)[|y| = p(|x|) \wedge (x, y) \in B]\}.$$

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Intuition. Let  $A \in \text{NP}$ .

- ▶ If  $x \in A$  then there is a SHORT (poly in  $|x|$ ) proof of this fact, namely  $y$ , such that  $x$  can be VERIFIED in poly time. So if I wanted to convince you that  $x \in L$ , I could give you  $y$ . You can verify  $(x, y) \in B$  easily and be convinced.

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- ▶ If  $x \notin A$  then there is NO short verifiable proof that  $x \in A$ .



# Examples of Sets in NP

$$\text{SAT} = \{\phi : (\exists \vec{y})[\phi(\vec{y}) = T]\}$$

$$3\text{COL} = \{G : G \text{ is 3-colorable}\}$$

$$\text{CLIQ} = \{(G, k) : G \text{ has a clique of size } k\}$$

$$\text{HAM} = \{G : G \text{ has a Hamiltonian Cycle}\}$$

$$\text{EUL} = \{G : G \text{ has an Eulerian Cycle}\}$$

**Note** These all ask if something EXISTS. To FIND the (say) 3-coloring one can make queries to (say) 3COL.

**Note**  $\text{EUL} \in P$ . The rest are NPC hence likely NOT in P.

# Reductions

**Def** Let  $X, Y$  be languages. A **reduction** from  $X$  to  $Y$  is a polynomial-time computable function  $f$  such that

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**Easy Lemma** If  $X \leq Y$  and  $Y \in P$  then  $X \in P$ .

**Contrapositive** If  $X \leq Y$  and  $X \notin P$  then  $Y \notin P$ .

# Def of NP-Complete

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**Easy Lemma** If  $Y$  is NP-complete and  $Y \in \text{P}$  then  $\text{P} = \text{NP}$ .

# SAT is NP-Complete

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3. Thousands of problems are NP-complete. If any are in P then they are all in P.

# The Cook-Levin Thm

# What does the Proof Involve

Proof involved coding a TM into a Boolean Formula which had parts:

1.  $z_{i,j,\sigma} = T$  iff the  $j$ th symbol in the  $i$ th configuration is  $\sigma$ .
2. First config: input  $x$ , start state, SOME  $y$  of the right length.
3. Last config: accepts
4.  $C_{i+1}$  follows from  $C_i$ .

# Closure of P

# Easy Closure Properties of P

Assume  $L_1, L_2 \in P$ .

1.  $L_1 \cup L_2 \in P$ . EASY. Uses polys closed under addition.
2.  $L_1 \cap L_2 \in P$ . EASY. Uses polys closed under addition.
3.  $\overline{L_1} \in P$ . EASY.
4.  $L_1 L_2 \in P$ . EASY. Uses  $p(n)$  poly then  $np(n)$  poly.

# Closure of P Under \*

**Thm** If  $L \in P$  then  $L^* \in P$ .

## **Proof**

First let's talk about what you **should not** do:

The technique of looking at **all** ways to break up  $x$  into pieces takes roughly  $2^n$  steps, so we need to do something clever.



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**New Problem** Given  $x = x_1 \cdots x_n$  want to know:

$$e \in L^*$$

$$x_1 \in L^*$$

$$x_1 x_2 \in L^*$$

$$\vdots$$

$$x_1 x_2 \cdots x_n \in L^*.$$

**Intuition**  $x_1 \cdots x_i \in L^*$  IFF it can be broken into TWO pieces, the first one in  $L^*$ , and the second in  $L$ .

# Final Algorithm

$A[i]$  stores if  $x_1 \cdots x_i$  is in  $L^*$ .  $M$  is poly-time Alg for  $L$ , poly  $p$ .

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$A[1] = A[2] = \dots = A[n] = \text{FALSE}$

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for  $i = 1$  to  $n$  do

    for  $j = 0$  to  $i - 1$  do

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**Note** Key is that the set of polynomials is closed under mult by  $n^2$ .

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The following defines  $L_1 \cup L_2$  in an NP-way.

$$L_1 \cup L_2 = \{x : (\exists y):$$

- ▶  $|y| = p_1(|x|) + p_2(|x|) + 1$ .  $y = y_1\$y_2$  where  $|y_1| = p_1(|x|)$  and  $|y_2| = p_2(|x|)$ .
- ▶  $(x, y_1) \in B_1 \vee (x, y_2) \in B_2$

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Witness:  $|y| = p_1(|x|) + p_2(|x|) + 1$  is short.

Verification:  $(x, y_1) \in B_1 \vee (x, y_2) \in B_2$ , is quick.

# Closure of NP under Intersection

**Thm** If  $L_1 \in \text{NP}$  and  $L_2 \in \text{NP}$  then  $L_1 \cap L_2 \in \text{NP}$ .

Proof is similar to closure under  $\cup$ .

# Closure of NP under Concatenation

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$$\{x : (\exists x_1, x_2, y_1, y_2)$$

- ▶  $x = x_1 x_2$
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# Closure of NP under \*

**Thm** If  $L \in \text{NP}$  then  $L^* \in \text{NP}$ .

Similar to the Closure of NP under CONCAT.

**Much** easier than the proof that P is closed under \*.

# Is NP closed under Complementation?

**Unknown to Science!**

But the common opinion is NO.

# Is NP closed under Complementation?

## Unknown to Science!

But the common opinion is NO.

Unlikely that there is a short poly-verifiable witness to  $G$  NOT being 3-colorable.

# IND SET is NP-Complete, 3COL is NP-Complete

GOTO the slides from Stanford

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SAT-solvers are so good that people now use reductions to map problem TO SAT.

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Let  $G = (V, E)$ .

$G$  has a clique of size  $k$  is EQUIVALENT TO:

*There is a 1-1 function  $\{1, \dots, k\} \rightarrow V$  such that for all  $1 \leq a, b \leq k$ ,  $(f(a), f(b)) \in E$ .*

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Given  $G$  and  $k$  We want to know:

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We formulate this as a Boolean Formula:

1. For  $1 \leq i \leq k$ ,  $1 \leq j \leq n$ , have Boolean Vars  $x_{ij}$ . Intent:

$$x_{ij} = \begin{cases} T & \text{if vertex } i \text{ maps to vertex } j \\ F & \text{if vertex } i \text{ does not maps to vertex } j \end{cases} \quad (1)$$

2. Part of formula says  $x_{ij}$  is a bijection.
3. Part of formula says that the  $k$  points map to a clique.

# Decidability and Undecidability

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4. Everything computable is computable by some TM.
5. A TM that halts on all inputs is called **total** .

# Computable Sets

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Computable sets are also called decidable or solvable. A machine such as  $M$  above is said to **decide**  $A$ .

# Computable Sets

**Def** A set  $A$  is *computable* if there exists a Turing Machine  $M$  that behaves as follows:

$$M(x) = \begin{cases} Y & \text{if } x \in A \\ N & \text{if } x \notin A \end{cases} \quad (2)$$

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**Notation** DEC is the set of Decidable Sets.

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# Noncomputable Sets

Are there any noncomputable sets?

1. Yes—ALL SETS: uncountable. DEC Sets: countable, hence there exists an uncountable number of noncomputable sets.
2. YES—HALT is undecidable, and once you have that you have many other sets undec.
3. YES—the problem of telling if a  $p \in \mathbb{Z}[x_1, \dots, x_n]$  has an int solution is undecidable.

# The HALTING Problem

**Def** The HALTING set is the set

$$HALT = \{(e, d) \mid M_e(d) \text{ halts} \}.$$

# HALT is Undecidable

**Thm** HALT is not computable.

**Proof** Assume HALT computable via TM  $M$ .

$$M(e, d) = \begin{cases} Y & \text{if } M_e(d) \downarrow \\ N & \text{if } M_e(d) \uparrow \end{cases} \quad (3)$$

We use  $M$  to create the following machine which is  $M_e$ .

1. Input  $d$
2. Run  $M(d, d)$
3. If  $M(d, d) = Y$  then RUN FOREVER.
4. If  $M(d, d) = N$  then HALT.

$$M_e(e) \downarrow \implies M(e, e) = Y \implies M_e(e) \uparrow$$

$$M_e(e) \uparrow \implies M(e, e) = N \implies M_e(e) \downarrow$$

We now have that  $M_e(e)$  cannot  $\downarrow$  and cannot  $\uparrow$ . **Contradiction.**



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Proofs by reductions. Similar to NPC. We will not do that.

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# Compare NP to $\Sigma_1$

$A \in \text{NP}$  if there exists  $B \in \text{P}$  and poly  $p$  such that

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4. Are ideas from Computability theory useful in complexity theory? Yes, to a limited extent. My thesis was on showing some of those limits.

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TOT is **harder** than HALT.

# WS1S Formulas and Sentences

1. Variables  $x, y, z$  range over  $\mathbb{N}$ ,  $X, Y, Z$  range over finite subsets of  $\mathbb{N}$ .
2. Symbols:  $<, \in$  (usual meaning),  $S$  (meaning  $S(x) = x + 1$ ).
3. A *Formula* allows variables to not be quantified over. A Formula is neither true or false. Example:  $(\exists x)[x + y = 7]$ .
4. A *Sentence* has all variables quantified over. Example:  $(\forall y)(\exists x)[x + y = 7]$ . So a Sentence is either true or false IF domain is

WS1S: Weak Second order Theory of One Successor. Weak Second order means quantify over finite sets.

# Atomic Formulas

An *Atomic Formula* is:

1. For any  $c \in \mathbb{N}$ ,  $x = y + c$  is an Atomic Formula.
2. For any  $c \in \mathbb{N}$ ,  $x < y + c$  is an Atomic Formula.
3. For any  $c, d \in \mathbb{N}$ ,  $x \equiv y + c \pmod{d}$  is an Atomic Formula.
4. For any  $c \in \mathbb{N}$ ,  $x + c \in X$  is an Atomic Formula.
5. For any  $c \in \mathbb{N}$ ,  $X = Y + c$  is an Atomic Formula.

# WS1S Formulas

Build up formulas from atomic formulas using  $\wedge$ ,  $\vee$ ,  $\neg$ ,  $\exists$ ,  $\forall$ .  
Hence can define the set of formulas.

Can put formulas into **Prenex Normal Form** :

$$(Q_1 v_1)(Q_2 v_2) \cdots (Q_n v_n)[\phi(v_1, \dots, v_n)]$$

**Def** If  $\phi(x_1, \dots, x_n, X_1, \dots, X_m)$  is a WS1S Formula then  
 $TRUE(\phi)$  is the set

$$\{(a_1, \dots, a_n, A_1, \dots, A_m) \mid \phi(a_1, \dots, a_n, A_1, \dots, A_m) = T\}$$

This is the set of  $(x_1, \dots, x_n, X_1, \dots, X_m)$  that make  $\phi$  TRUE.

# KEY THEOREM

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We prove this by induction on the formation of a formula. If you prefer- induction on the LENGTH of a formula.

# DECIDABILITY OF WS1S

**Thm:** WS1S is Decidable.

**Proof:**

1. Given a SENTENCE in WS1S put it into the form

$$(Q_1 X_1) \cdots (Q_n X_n) (Q_{n+1} x_1) \cdots (Q_{n+m} x_m) [\phi(x_1, \dots, x_m, X_1, \dots, X_n)]$$

2. Assume  $Q_1 = \exists$ . (If not then negate and negate answer.)
3. View as  $(\exists X)[\phi(X)]$ , a FORMULA with ONE free var.
4. Construct DFA  $M$  for  $\{X \mid \phi(X) \text{ is true}\}$ .
5. Test if  $L(M) = \emptyset$ .
6. If  $L(M) \neq \emptyset$  then  $(\exists X)[\phi(X)]$  is TRUE.  
If  $L(M) = \emptyset$  then  $(\exists X)[\phi(X)]$  is FALSE.

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4. For WS1S, WS2S, S1S, can state things about low-level verification of code (see MONA group). For S2S can state actual Mathematical Theorems of interest. However, the program would take too long to use this, and it would not offer mathematical insights anyway.

# Def of Randomness

Taking a cue from the above two examples, we will define the **Randomness of a string  $x$**  to be the size of the shortest Turing Machine (TM) that prints  $x$ .

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3. A string is **Kolmogorov random** if  $C(x) \geq n$ . A string is **Kolmogorov random relative to  $y$**  if  $C(x|y) \geq n$ .

**Note** Java-Random, Python-Random, 1-tape-TM-Random will all give different values. But all within  $O(1)$ .

# Do Random Strings Exist?

Is there a string of length  $n$  that has  $C(x) \geq n$ ?

YES- there are more Strings of length  $n$  than TMs of length  $\leq n - 1$ .

# Application of Kolmogorov Complexity to Proving Languages Not Regular

# $L_1 = \{a^n b^n : n \in \mathbb{N}\}$ is Not Regular

Assume  $L_1$  is regular via  $M = (Q, \{a, b\}, \delta, s, F)$ .

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Pick  $n$  such that  $C(n) > A$ . Then you have a program of size  $A < C(n)$  printing out  $n$ , which is a contradiction.

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# BILL AND NATHAN RECORD LECTURE!!!!

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**UN-TIMED PART OF  
FINAL IS TUESDAY  
May 11 11:00A.  
NO DEAD CAT**

**FINAL IS THURSDAY**  
**May 13**  
**8:00PM-10:15PM**

**FILL OUT COURSE  
EVALS for ALL YOUR  
COURSES!!!**