Olet's try another reduction SAT SAT SAT - Very common type of reduction from a general to specific version of a problem. - Recall, the difference between SAT and 3-SAT is that in 3-SAT each clause has 3 literals. Reduction Given an instance I of SAT, take every clause with ≤ 3 literals as is (note, you can always "pad" clauses with ≤ 3 literals to have 3). Let $C = (A, VA_2V...VA_k)$ be a clause with k > 3 literals. Create a new set of clauses in the 3-SAT instance! $(a_1 v a_2 v y_1) \wedge (\overline{y}_1 v a_3 v y_2) \wedge (\overline{y}_2 v a_4 v y_3) \dots (\overline{y}_{k-3} v a_{k-1} v a_k)$ Where each of the y: are new variables. Call the resulting 3-SAT instance I. The conversion of I to I' is clearly polynomial-time.

<u>Claim</u>: There is a satisfying assignment for C= (a, vazv... vak) iff is a satisfying assignment for C'= (a, vazvy) v (y, vazvyz)... Proof: C'has a satisfying assignment -> C has a satisfying assignment => It c' is satisfied, then at least 1 of a, , ... , a, is true; otherwise, y would have to be true, which would force yz to be true ... so that eventually y would be fulse, along with the whole clause. Thus, if c' is satisfied, some q' is true and so C is satistied. <= If c has a satisfying assignment -> c' has a satisfying assignment Assume w.l.o.g that ai is the first true literal in c. Then, set g1)...) yi-2 to true and the rest to Salse. This satisfies all of C'. Thus, any instance of SAT can be (poly-time) transformed into an instance of 3-SAT such that the SAT instance has a satisfying assignment iff the 3-SAT instance does. So SAT = 3-SAT

Beyond P, NP and NP-complete.

Note that we have talked about NP-complete problems as decision problems that are at least as hard as any problem in NP.

Do not confuse <u>NP-complete</u> with <u>NP-hard</u>.

NP-complete problems are <u>required</u> to be in NP (to have efficient certifiers) NP-hard problems are not, and some NP-hard problems are not decidable in any Sinite time.

think of the following sketch:

NP-hard WP-completer NPPP

NP-hard P=NP=NP-complete IF P=NP

If P+NP

Some NP-hard problems have decision versions that are NP-complete.

For example, the Traveling Selesmon Problem asks to find the shortest Hamiltonian cycle (a cycle that visits each vertex exactly once, except the stort vertex to which it returns) in a graph.

The decision problem asks if a graph \mathcal{C} has a Hamiltonian cycle of length $\leq K$, For which the cycle itself is an efficient certificate.

Clearly the decision version of TSP is in NP. However, some decision problems are NP-hard but not NR-complete (because they are not in NP). One tamous example is the halting problem.

Informally the halting problem asks us to deterime, for pairs (i, x) of programs and input, will program is halt (eventually terminate) on input X.

We wish to know if there exists an algorithm (Turing Machine) that can solve the halting problem in <u>Finite</u> time.

Unfortunately, there is not. The proof itself is a bit involved, but here is the basic concept.

Assume that we had some algorithm, halts (i, x) TRUE if program i halts and FALSE if it does not.

Now, recognize that a program (i) is just data, so we could write the following function

det SelfHalt(p) E return halts(p,p) 3 That is, does program p halt when run on itself?

Finally, consider Feeding the following program p to Self Halt () def paradox (prog) { if Self Halt (prog) & while (True) & return False; 3 else & 3 return True Now, imagine calling Self Halt (paradox). This will run the paradox () function on itself. If paradox (paradox) halts, then Self Halt (paradox) should be True, but in that case we take the if branch, and the call goes into an infinite loop. IS puradox (paradox) does not halt, then we go into the else branch and return True ... which is wrong because Self Halt (paradox) should be False. This means that such a general algo as halts () can not exist!

So, the halting problem is undecidable. There is no way to solve it on all instances in finite (let alone polynomial) time.

- There are many interesting ideas related to this: e.g. Gödel's incompleteness theorem Implications for static program analysis Determination of what is even completable (in Finite time).

We do not have the time to go deeply into these topics, but I highly recommend you explore them a bit.

The strange case of co-NP

Our definition of NP is fundamentally asymmetric.

An input string is a "yes" instance iff $\exists t$ with $\lvert t \rvert = p(\lvert s \rvert)$ so that $B(s,t) \lor = "yes"$. Negating this statement, a string s is a "no" instance iff for all t', B(s,t') = no. That is:

- It is easy to verify we have a solution - It is hard to vesify that no solution exists

NP is concerned with the efficient verification of yes instances.

For every problem X, the complementary problem X is defined so that for all inputs S:

SEX iff S&X Note: if XEP then XEP, but this is not necessarily the case for NP.

Det: co-NP - A problem belongs to co-NP if its complementary problem belongs to NP.

Example: Subset Sum E NP. It asks, given a collection of numbers, is there a non-empty subset whose sum is O.

Consider the complement : Given a collection C of nombers, return True iff no non-empty subset of C sums to O. How does one efficiently verify that no such subset exists? I can efficiently verify "no" instances of this problem using the same certificate as Sor subset sum, but how do I verify "yes" instances?

Open Question:

NP = (0-NP? This is widely believed not to be the case. In fact

If NP = co-NP then P = NP

If NP = co-NP then P = NP Proof (contrapositive): P=NP => NP= co-NP P is closed under complementation, so if P=NP, then NP is closed under complementation as well. Specifically, assume P=NP: $x \in NP \Rightarrow x \in P \Rightarrow \overline{x} \in P \Rightarrow \overline{x} \in NP \Rightarrow x \in c_0 - NP$ and $x \in co - NP \implies \overline{x} \in NP \implies \overline{x} \in P \implies x \in P \implies x \in NP$ Hence NP = co - NP and co - NP = NP so NP = co - NP The set of decidable problems is called K.

Ket's end on a somewhat disturbing note: Almost all decision problems & R Why (proof concept from E. Demaine) Consider the following sets - Decision problems = Function from input > 20,13 input ~ binary string ~ natural number ~ M - Turing Machines Binary Strings 22 so it's a sunction from N->20,13 Natural Numbers ~~ 0 1 1 0 1 0 0 tupto ~~ } (countably infinite) IN $/ \rightarrow input 0 1 2 3 4 5 6 \dots$ Dinfinite set of bits. # of functions $N \rightarrow \{0,1\}$ has a randinality much greater than $N \cdot It$ is unrountably infihite. Almost all problems have no algorithm that solve them.

