

<div> <div><math>k</math>-tape Turing machine</div> <div> <div>turing-machine</div> <div>turing-machine-def</div> </div> </div>	<ul style="list-style-type: none"> <li>• A finite set <math>A</math>, called the <i>alphabet</i>.</li> <li>• A collection of <math>k</math> <i>tapes</i>, each an infinite sequence of <i>cells</i> each containing an element of <math>A</math>. One tape is declared to be the <i>input</i> and another one is declared to be the <i>output</i>.</li> <li>• A set <math>S</math> of <i>states</i>, including two special states: <math>S_{\text{init}}</math> and <math>S_{\text{halt}}</math>.</li> <li>• A <i>transition function</i> <math>\delta : A^k \times S \rightarrow A^k \times S \times \{L, N, R\}</math>.</li> <li>• A <i>head</i> which is in a state and a <i>position</i> on each tape.</li> </ul>
<div> <div>Complexity class <math>\mathbf{P}</math></div> <div> <div>complexity-class::p</div> <div>complexity-class-p-def</div> </div> </div>	<p><math>f \in \mathbf{P}</math> iff there exists a Turing machine <math>T</math> and a polynomial <math>p</math> such that, for all <math>x</math>, <math>T</math> computes <math>f(x)</math> in time at most <math>p(x)</math>.</p>
<div> <div>Non-deterministic Turing machine</div> <div> <div>turing-machine::nondet</div> <div>non-deterministic-turing-machine-def</div> </div> </div>	<p>A non-deterministic Turing machine <math>T</math> is like a Turing machine except that it has two transition functions, and we say that <math>T</math> computes <math>f</math> if, for all input <math>x</math>, <math>f(x) = 1</math> iff there's a sequence of choices between the two transition functions such that the output tape is 1 when <math>T</math> halts.</p>
<div> <div>Complexity class <math>\mathbf{NP}</math></div> <div> <div>complexity-class::np</div> <div>turing-machine::nondet</div> <div>complexity-class-np-def</div> </div> </div>	<p><math>f \in \mathbf{NP}</math> iff there exists a non-deterministic Turing machine <math>T</math> and a polynomial <math>p</math> such that, for all <math>x</math>, <math>T</math> computes <math>f(x)</math> in time at most <math>p(x)</math>.</p>

Alternative definition of **NP**

$f \in \mathbf{NP}$  iff there is a polynomial  $p$  and a function  $g \in P$  such that, for every  $x$ ,  $f(x) = 1$  iff  $\exists y \in \{0, 1\}^{p(|x|)}, g(x, y) = 1$ .

*Proof.*

$\implies$  Use  $y$  to encode the choices made by the non-deterministic TM. Then the TM for  $g$  is "Run the non-deterministic TM, reading  $y$  to know which transition function to follow."

$\impliedby$  Non-deterministically write down  $y$  and apply  $g$ .

□

complexity-class:np  
complexity-class-np-def-alt

Complexity class **co – NP**

$f \in \mathbf{co – NP}$  iff  $\neg f \in \mathbf{NP}$

Alternatively,  $f \in \mathbf{co – NP}$  iff there is a polynomial  $p$  and a function  $g \in P$  such that, for every  $x$ ,  $f(x) = 1$  iff  $\forall y \in \{0, 1\}^{p(|x|)}, g(x, y) = 1$ .

complexity-class:co-np  
complexity-class-co-np

Polynomial hierarchy

Define  $\Sigma_0^P$  and  $\Pi_0^P$  to be **P** and

$$\begin{aligned} f \in \Sigma_{k+1}^P &\iff \exists \text{ polynomial } p \text{ and } g \in \Pi_k^P, \\ &\qquad \forall x, f(x) = 1 \iff \exists y, g(x, y) = 1 \\ f \in \Pi_{k+1}^P &\iff \exists \text{ polynomial } p \text{ and } g \in \Sigma_k^P, \\ &\qquad \forall x, f(x) = 1 \iff \forall y, g(x, y) = 1 \end{aligned}$$

Define

$$\mathbf{PH} = \bigcup_k \Sigma_k^P \cup \Pi_k^P$$

complexity-class:ph  
complexity-class-ph

If **P** = **NP**, then **P** = **PH**

Induction on  $k$ :

- $\Sigma_1^P$  by assumption.  $\Pi_1^P = P$  since we can negate, calculate in polynomial time, negate again.
- If  $f \in \Sigma_{k+1}^P$ , then there exists  $g \in \Pi_k^P$  such that  $\forall x, f(x) = 1 \iff \exists y, g(x, y) = 1$ . By induction hypothesis,  $g \in \mathbf{P}$ . So  $f \in \mathbf{NP} = \mathbf{P}$ . Similarly,  $\Pi_{k+1}^P = \mathbf{P}$ .

Polynomial hierarchy collapse

If  $\Sigma_k^P = \Sigma_{k+1}^P$  or  $\Sigma_k^P = \Pi_k^P$ , then  $\mathbf{PH} = \Sigma_k^P$ .

*Proof.* Prove  $\Sigma_\ell^P = \Pi_\ell^P = \Sigma_k^P$  by induction on  $\ell$ . The trick is to use the induction hypothesis to turn  $\exists\forall$  or  $\forall\exists$  into  $\forall\forall$  or  $\exists\exists$  and then collapse the quantifiers.  $\square$

complexity-class:ph  
polynomial-hierarchy-collapse

Complexity class **PSPACE**

$f \in \mathbf{PSPACE}$  iff  $f$  can be computed by a Turing machine using a polynomial amount of tape.

complexity-class:pspace  
complexity-class:pspace

$\mathbf{NP} \subseteq \mathbf{PSPACE}$

Assume  $f(x) = 1 \iff \exists y, g(x, y) = 1$  where  $g \in \mathbf{P}$ . It takes a polynomial amount of space to iterate through all possible  $y$  and a polynomial amount of space to compute  $g(x, y)$  for each  $y$ . Hence it takes a polynomial amount of space to compute  $f$ .

complexity-class:np complexity-class:pspace  
np-subset-pspace

$\mathbf{PH} \subseteq \mathbf{PSPACE}$

Prove that  $\Sigma_k^P, \Pi_k^P \subseteq \mathbf{PSPACE}$  by induction on  $k$ , using the fact that it's possible to brute-force search  $y$  of polynomial size such that  $g(x, y) = 0$  or  $g(x, y) = 1$  in polynomial time if  $g \in \mathbf{P}$ .

complexity-class:ph complexity-class:pspace  
ph-subset-pspace

Complexity class **EXPTIME**

$f \in \mathbf{EXPTIME}$  iff  $f$  can be computed by a Turing machine in time  $\exp(O(n^k))$  for some  $k$

complexity-class:exptime  
complexity-class:exptime

**PSPACE**  $\subseteq$  **EXPTIME**

If a Turing machine takes polynomial space to compute inputs of size  $n$ , say  $p(n)$ , then its **configuration** (combination of the state, the position on the tapes, the values of each cell on the tapes) goes through at most  $p(n)^k |S| |A|^{kp(n)} = \exp(O(p(n)))$  possibilities. Further, if it went through one possibility twice, it would loop. Hence the computation takes exponential time.

complexity-class:pspace complexity-class:exptime  
pspace-subset-exptime

Complexity class **NEXPTIME**

$f \in \mathbf{NEXPTIME}$  iff there is a polynomial  $p$  and a function  $g \in \mathbf{EXPTIME}$  such that, for every  $x$ ,  $f(x) = 1$  iff  $\exists y \in \{0, 1\}^{p(|x|)}, g(x, y) = 1$ .

complexity-class:nexptime  
complexity-class:nexptime

Complexity class **EXPSPACE**

$f \in \mathbf{EXPSPACE}$  iff  $f$  can be computed using tapes of length  $\exp(O(n^k))$  for some  $k$

complexity-class:expspace  
complexity-class:expspace

$$\begin{array}{c} \mathbf{P} \subseteq \mathbf{NP} \subseteq \mathbf{PSPACE} \\ \subseteq \\ \mathbf{EXPTIME} \subseteq \mathbf{NEXPTIME} \subseteq \mathbf{EXPSpace} \end{array}$$

complexity-class:pspace complexity-class:pspace complexity-class:exptime  
basic-complexity-classes-hierarchy

Circuit

A *circuit* is a directed acyclic graph (DAG) such that each vertex is labelled as either an *input*, an **AND** gate, an **OR** gate or a **NOT** gate.

- An input is a vertex of in-degree 0.
- A **NOT** gate has in-degree 1.
- All vertices of in-degree  $> 1$  are **AND** or **OR** gates.
- Vertices of out-degree 0 are outputs.

The value at an **AND/OR** is the min / max of its predecessors. The value at a **NOT** is  $1 - x$  where  $x$  is the value at its predecessor.

circuit  
circuit-def

Fan-in of a circuit

The fan-in of a circuit is the maximum in-degree of any **AND** or **OR** gate.

circuit  
fan-in-def

Straight-line computations

A straight-line computation of  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  of length  $m$  is a sequence of functions  $f_1, \dots, f_m$  starting with  $f_i(x) = x_i$  for  $i = 1, \dots, n$ , ending with  $f_m = f$ , and for each  $i > n$  there are some  $j_1, \dots, j_k < i$  such that either

$$\begin{array}{l} f_i = f_{j_1} \wedge \dots \wedge f_{j_k} \\ f_i = f_{j_1} \vee \dots \vee f_{j_k} \end{array}$$

This is the same as taking intersections and unions of half-spaces in an hypercube in order to get to some set.

straight-line-computation  
straight-line-computation-def

The smallest size of a circuit computing  $f$  is the shortest length of a straight line computing  $f$

circuit straight-line-computation  
straight-line-computation-circuit

Every function  $f : \{0,1\}^n \rightarrow \{0,1\}$  can be computed by a circuit of size exponential in  $n$ .

circuit  
circuit-exponential

Family of circuits that each computes the output of a Turing machine on inputs of a given size

circuit turing-machine  
turing-machine-to-circuits

Complexity class **P/poly**

complexity-class:p-poly  
complexity-class-p-poly

A straight-line computation is the same as a circuit whose vertices have been totally ordered in a way that respects its edges.

For every possible input  $x$ , build a circuit that recognises  $x$  (using at most  $n$  **AND** gates and  $n$  **NOT** gates) and outputs  $f(x)$  if it's recognised. Then take a giant **OR** gate of all of those. This circuit has size at most  $2^{n+1}$  and computes  $f$  since

$$f(x_1) \wedge (x_1 = x_i) \vee \dots f(x_{2^n}) \wedge (x_{2^n} = x_i) = f(x_i)$$

Let  $f$  be a function computed by a  $k$ -tapes Turing machine  $T$  in time  $p(n)$  for inputs of size  $n$ . Then there is a family  $C_n$  of circuits such that  $C_n$  computes  $f$  for inputs of size  $n$  and

$$|C_n| = O(p(n)^{k+2})$$

*Proof.* WLOG assume the alphabet is  $\{0,1\}$ . Encode the configuration of the machine in  $|S|$  variables for the state,  $2kp(n)$  variables for the position of the heads,  $2kp(n)$  variables for the values of the reachable cells. The transition function has  $k + |S|$  inputs, hence can be computed in a circuit of size  $O(1)$ . Updating the variables takes a circuit of size  $O(p(n))$ . Hence the whole circuit has size  $O(p(n)^2)$ .  $\square$

$f \in \mathbf{P/poly}$  if one (hence all) of the following holds

- There is a family  $C_n$  of polynomial-size circuits such that  $C_{|x|}$  computes  $f(x)$ .
- There is a polynomial  $p$  and a sequence  $y_n$  with  $|y_n| = p(n)$  and a function  $g \in \mathbf{P}$  such that  $f(x) = 1 \iff g(x, y_{|x|}) = 1$ .  $y_n$  should be thought of as an "advice string" to help compute  $f$ .
- There is a sequence of Turing machines  $T_n$  and a polynomial  $p$  such that  $T_n$  has  $\leq p(n)$  states and computes  $f(x)$  when  $|x| = n$ .

The three definitions of **P/poly** are equivalent

complexity-class:p-poly  
complexity-class:p-poly-alt

**P**-uniformity

p-uniformity  
p-uniformity-def

If **P** = **NP**, then search problems are equivalent to decision problems

complexity-class:p complexity-class:np  
p-np-search-decision-problem

If **NP**  $\subseteq$  **P/Poly**, and  $f \in \mathbf{NP}$ , then we can compute certificates for  $f$  using polynomial-size circuits

complexity-class:np complexity-class:p-poly  
np-subset-p-poly-certificate

- Family of circuits  $\implies$  advice string. Let  $y_n$  be an encoding of  $C_n$  and let  $g(x, y) = 1$  if the circuit encoded by  $y$  outputs 1 with input  $x$ .
- Advice string  $\implies$  family of circuits. Let  $C'_n$  compute  $g$  and make  $C_n$  to be  $C'_n$  with the last  $p(n)$  inputs set to  $y_n$ .
- Advice string  $\implies$  family of TMs. Let  $T$  compute  $g$  and let  $T_n$  be a Turing machine that prints out  $y_n$  and then uses  $T$  to compute  $g(x, y_n)$ .
- Family of TMs  $\implies$  advice string. Let  $y_n$  be an encoding of  $T_n$  and let  $g(x, y) = 1$  if the Turing machine encoded by  $y$  outputs 1 with input  $x$  ( $g$  is encoded by a universal TM).

A family of circuits  $C_n$  is **P**-uniform if there is an algorithm that generates it in polynomial-time (in  $n$ ).

If **P** = **NP** and  $f \in \mathbf{NP}$ , say  $f(x) = 1 \iff \exists y, |y| = p(|x|)$  and  $g(x, y) = 1$  where  $g \in \mathbf{P}$ , then there is some polynomial-time algorithm  $h$  such that if  $f(x) = 1$  then  $g(x, h(x)) = 1$ .

*Proof.* For each  $i$ , let  $g_i$  be the function with input  $x$  and  $u_i$  where  $|u_i| = i$  and output whether it can be extended by a 1. Now calculate  $u_1 = g_0(x)$ ,  $u_2 = g_1(x, u_1)$ ,  $u_3 = g_1(x, u_1, u_2)$ , etc. At the end we obtain  $h(x) = (u_1, \dots, u_{p(|x|)})$  such that  $g(x, u) = 1$ . Each  $g_i$  is obviously in **NP** = **P**, so  $h$  can be computed in polynomial time (assuming the  $g_i$  are uniformly in polynomial time).  $\square$

Say  $f(x) = 1 \iff \exists y, g(x, y) = 1$  where  $g \in \mathbf{P}$ . Define  $g_i(x, u_1, \dots, u_i) = 1 \iff \exists y, g(x, u_1, \dots, u_i, 1, y)$ .  $g_i \in \mathbf{NP} \subseteq \mathbf{P/Poly}$ , hence find a polynomial-size circuit family  $C_{i,n}$  that computes  $g_i$ . Now put circuits  $C_{0,n}, \dots, C_{p(n),n}$  together as follows:

- $C_{0,n}$  takes inputs  $x_1, \dots, x_n$  and outputs  $u_1$
- $C_{1,n}$  takes inputs  $x_1, \dots, x_n, u_1$  and outputs  $u_2$
- ...

The resulting circuit  $C_n$  is such that if  $\exists y, g(x, y) = 1$  then  $g(x, C_n(x)) = 1$ .

Karp-Lipton theorem

complexity-class:np complexity-class:p-poly complexity-class:ph  
karp-lipton

For every  $k$ , there's a boolean function  $f$  that can be computed by a circuit family of size  $n^{k+1}$  but not by a circuit family of size  $n^k$ .

circuit  
boolean-function-precise-polynomial

For every  $k$ , there is a boolean function  $f \in \Sigma_4^P$  that cannot be computed by a family of circuits of size  $n^k$

circuit complexity-class:ph  
boolean-function-sigma-four-not-polynomial

Kannan's theorem

complexity-class:np complexity-class:ph circuit  
kannan

If  $\mathbf{NP} \subseteq \mathbf{P/poly}$ , then  $\Sigma_2^P = \Pi_2^P$ .

*Proof.* By symmetry, it's enough to prove  $\Pi_2^P \subseteq \Sigma_2^P$ . Let  $f \in \Pi_2^P$  and let  $g \in \mathbf{NP}, h \in \mathbf{P}$  be such that

$$\begin{aligned} f(x) = 1 &\iff \forall y, g(x, y) = 1 \\ g(x, y) = 1 &\iff \exists z, h(x, y, z) = 1 \end{aligned}$$

By the existence of polynomial certificates when  $\mathbf{NP} \subseteq \mathbf{P/poly}$ , find a polynomial-size circuit family  $C_n$  such that  $g(x, y) = 1 \implies h(x, y, C_n(x, y))$ . Then

$$f(x) = 1 \implies \exists C, \forall y, h(x, y, C_n(x, y)) = 1$$

The converse is true by assumption. Hence  $f \in \Sigma_2^P$ . □

TODO

For sufficiently large  $n$ , the lemma gives us  $f'_n : \{0, 1\}^n \rightarrow \{0, 1\}$  that can be computed by a circuit of size  $n^{k+1}$  but not by a circuit of size  $n^k$ . Choose an ordering of the circuits of size  $\leq n^{k+1}$  that's computable in polynomial time. Let  $C_n$  be the first circuit in this ordering such that no circuit of size  $\leq n^k$  computes the same function as  $C_n$ . Let  $f(x) = C_{|x|}(x)$ . Then

$$\begin{aligned} f(x) = 1 &\iff \exists C_n, |C_n| \leq n^{k+1} \text{ and } C_n(x) = 1 \\ &\text{and } \forall D, |D| \leq n^k, \exists y, C_n(y) \neq D(y) \\ &\text{and } \forall E < C_n, \exists F, |F| \leq n^k, \forall z, E(z) = f(z) \end{aligned}$$

The  $\exists \forall \exists \forall$  shows that  $f \in \Sigma_4^P$ .

For every  $k$ , there is a function  $f \in \Sigma_2^P \cap \Pi_2^P$  that cannot be computed by a circuit family of size  $n^k$ .

*Proof.* If  $\mathbf{NP} \subseteq \mathbf{P/poly}$ , then  $\mathbf{PH} \subseteq \Sigma_2^P \cap \Pi_2^P$  by Karp-Lipton. So the function in  $\Sigma_4^P$  that cannot be computed by a circuit of size  $\leq n^{k+1}$  does the job.

If  $\mathbf{NP} \not\subseteq \mathbf{P/poly}$ , then there is some  $f \in \mathbf{NP} \subseteq \Sigma_2^P \cap \Pi_2^P$  that cannot be computed by *any* polynomial-size circuit family. □



Complexity class **L**

$f \in \mathbf{L}$  iff  $f$  can be computed with a logarithmic amount of memory. Formally,  $f \in \mathbf{L}$  iff it can be computed by some Turing machine with a read-only input tape, a write-only output tape and worktapes of size  $O(\log n)$  for inputs of size  $n$ .

complexity-class:l  
complexity-class-l

Complexity class **NL**

$f \in \mathbf{NL}$  iff  $f$  can be non-deterministically computed with a logarithmic amount of memory. Formally,  $f \in \mathbf{NL}$  iff it can be computed by some non-deterministic Turing machine with a read-only input tape, a write-only output tape and worktapes of size  $O(\log n)$  for inputs of size  $n$ .

complexity-class:nl  
complexity-class-nl

**NL**  $\subseteq$  **P**

Let  $f \in \mathbf{NL}$ ,  $T$  be a non-deterministic Turing machine that computes  $f$  in log-space and  $G$  be the *configuration graph* of  $T$ . Then  $f(x) = 1$  iff there is a directed path in  $G$  from the initial configuration to one such that  $T$  has halted with output 1. Since  $T$  runs on  $O(\log n)$  space, the number of vertices of  $G$  is polynomial in  $n$  ( $|V(G)| \leq 2^{O(\log n)}|S| = n^{O(1)}$ ). But note that **REACHABILITY**, the problem of determining whether there is a directed path from a vertex  $x$  to a subset  $S$  of vertices on a directed graph is easily seen to be in **P**: just brute force search to find the neighbours.

complexity-class:nl complexity-class:p  
nl-subset-p

Low depth-computation classes

The class  $\mathbf{NC}^i$  where  $i \in \mathbb{N}$  consists of all functions that can be computed by a family of circuits of polynomial size, fan-in 2 and depth  $O(\log^i n)$ , where *depth* of a circuit is the length of the longest directed path in the associated DAG.  $\mathbf{AC}^i$  is like  $\mathbf{NC}^i$  except that we allow unbounded fan-in. We then define

$$\mathbf{NC} = \bigcup_i \mathbf{NC}^i, \mathbf{AC} = \bigcup_i \mathbf{AC}^i$$

complexity-class:nc complexity-class:ac  
complexity-class-nc-ac

log-space uniformity

$f \in \mathbf{u-NC}^i$  iff  $f$  can be computed by a family of circuits of polynomial size, fan-in 2 and depth  $O(\log^i n)$  that can be generated in log-space.

complexity-class:nc  
log-space-uniformity-def

$\mathbf{AC} = \mathbf{NC}$

Obviously,  $\mathbf{NC}^i \subseteq \mathbf{AC}^i$ . But we also have  $\mathbf{AC}^i \subseteq \mathbf{NC}^{i+1}$  since a circuit of fan-in  $k$  can be replace by a circuit of fan-in 2 which is at most  $\log k$  bigger by replacing each gate of in-degree  $d$  by  $\log d$  gates of in-degree 2.

complexity-class:nc complexity-class:ac  
ac-eq-nc

Complexity classes  $\mathbf{RP}$ ,  $\mathbf{co-RP}$  and  $\mathbf{ZPP}$

$f \in \mathbf{RP}$  (randomised polynomial time) iff there is a polynomial  $p$  and a function  $g \in \mathbf{P}$  such that if  $|x| = n$  and  $m = p(n)$  then

$$\mathbb{P}_{y \in \{0,1\}^m}(g(x,y) = 1) \begin{cases} = 0 & \text{if } f(x) = 0 \\ \geq \frac{1}{2} & \text{if } f(x) = 1 \end{cases}$$

$f \in \mathbf{co-RP}$  iff  $\neg f \in \mathbf{RP}$ .

$\mathbf{ZPP}$  (zero-error probabilistic polynomial time) is  $\mathbf{RP} \cap \mathbf{co-RP}$

complexity-class:rp complexity-class:co-rp complexity-class:zpp  
complexity-class-rp-co-rp-zpp

How to improve the accuracy of an algorithm in  $\mathbf{RP}$

Run the algorithm many times. Say we're computing  $f$  and  $g \in \mathbf{P}$  is such that

$$\mathbb{P}(g(x,y) = 1) \begin{cases} = 0 & \text{if } f(x) = 0 \\ \geq \frac{1}{2} & \text{if } f(x) = 1 \end{cases}$$

Then if  $y_1, \dots y_k$  are independent samples, we get

$$\mathbb{P}(\exists i, g(x,y_i) = 1) \begin{cases} = 0 & \text{if } f(x) = 0 \\ \geq 1 - 2^{-k} & \text{if } f(x) = 1 \end{cases}$$

complexity-class:rp  
rp-improved-accuracy

Complexity class **BPP**

$f \in \mathbf{BPP}$  (bounded-error probabilistic polynomial time) iff there is a polynomial  $p$  and a function  $g \in \mathbf{P}$  such that if  $|x| = n$  and  $m = p(n)$  then

$$\mathbb{P}_{y \in \{0,1\}^m}(g(x,y) = f(x)) \geq \frac{2}{3}$$

complexity-class:bpp  
complexity-class:bpp

How to improve the accuracy of an algorithm in **BPP**

Run the algorithm many times. Say we’re computing  $f$  and  $g \in \mathbf{P}$  is such that

$$\mathbb{P}(g(x,y) = f(x)) \geq \frac{2}{3}$$

Take  $y_1, \dots, y_k$  independent samples. Compute  $g(x, y_1), \dots, g(x, y_k)$ . Output the majority. The probability of getting the wrong answer is at most  $\exp(-\frac{k}{48})$  by Chernoff.

complexity-class:bpp  
bpp-improved-accuracy

**BPP**  $\subseteq$  **P**/poly

If  $k \geq 48n$ , then the probability that the majority is wrong is  $< 2^{-n}$ . Therefore there exist  $y_1, \dots, y_k$  such that for **every**  $x$  the majority vote is correct.  $y_1 \dots y_k$  serves as an advice string, together with the function that computes the majority vote.

complexity-class:bpp complexity-class:p-poly  
bpp-subset-p-poly