

Computable Analysis toolkit

All the definitions that follow are directly taken from the book ?. We specify on which page the definition is given with brackets, for example we write {12} for page 12.

1 Questions

2 Basics in Discrete Complexity Theory

2.1 Deterministic computation

{13} In this book, Turing machines are described to have multiple states and tapes, with two distinguished states (input and halting state), and two distinguished tapes: a read-only input tape, and a write-only output tape. They also can only move right or left (they cannot stay on the same cell).

{14} A recognizer is a Turing machine that has two different halting states: an accepting state, and a rejecting state. For such a machine, we denote by $L(M)$ the set of strings for which the machine halts in the accepting state.

{14} $\text{time}_M(s)$ is the number of steps that the machine M makes before halting on input $s \in \Gamma^*$. $\text{space}_M(s)$ is the number of cells *that are not on the input and output tapes* that the machine M visits before halting on input $s \in \Gamma^*$. We define, for a machine M its time and space complexity functions as

$$t_M(n) := \max\{\text{time}_M(s) : s \in \Gamma^n\}, \quad (1)$$

$$s_M(n) := \max\{\text{space}_M(s) : s \in \Gamma^n\}. \quad (2)$$

For a computable set A , it is said to have time (space) complexity $\psi : \mathcal{N} \rightarrow \mathcal{N}$ if there exists a Turing machine M computing A such that $t_M(n) \leq \psi(n)$ ($s_M(n) \leq \psi(n)$). The complexity of a computable function ϕ is defined analogously. {15} We say that a computable set (function) A (ψ) is computable in polynomial time if ψ is a polynomial. \mathbf{P} is the class of polynomial-time computable sets, and \mathbf{FP} is the class of polynomial-time computable functions.

{15} **Conjecture: extended Church-Turing thesis.** For every two reasonable computation model, the time and space complexity of a given problem differ at most of a polynomial factor.

2.2 Nondeterministic computation

{15} In this book, nondeterministic Turing machines are described as Turing machines which transition function maps to a subset of all possible transitions.

{16} $A \subset \Gamma^*$ is recognized by a ND Turing machine M if there exists at least one computation starting at s path that halts in accepting state iff $s \in A$.

{16} $\text{time}_M(s)$ is the number of steps that the machine M makes before at least one accepting computation path halts on input $s \in \Gamma^*$. $\text{space}_M(s)$ is the number of cells that the machine M visits before at least one accepting

computation path halts on input $s \in \Gamma^*$. We define, for a machine M its time and space complexity functions as

$$t_M(n) := \max\{\text{time}_M(s) : M \text{ accepts } s, s \in \Gamma^n\}, \quad (3)$$

$$s_M(n) := \max\{\text{space}_M(s) : M \text{ accepts } s, s \in \Gamma^n\}. \quad (4)$$

NP is the class of sets accepted by polynomial time nondeterministic Turing machines.

{17} More sets are introduced, with the obvious meaning: **LOGSPACE**, **NLOGSPACE**, **EXP**, **PEXP**, **PSPACE**, **NPSPACE**, **FLOGSPACE**.

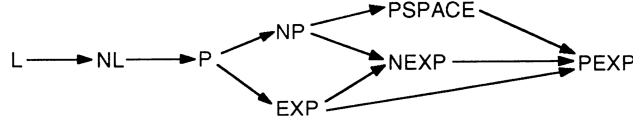


Figure 1.1. Inclusion relations on some complexity classes. In the above we use L for **LOGSPACE** and NL for **NLOGSPACE**, and we write $A \rightarrow B$ for $A \subseteq B$.

2.3 Reducibilities

Definition D1.1 - {18} (*Many-to-one reducibility*). A set A is polynomial-time many-to-one reducible to a set B , written $A \leq_m^P B$, if there exists $\phi \in \mathbf{FP}$ such that for all $s \in \Gamma^*$, $s \in A \iff \phi(s) \in B$.

Definition D1.2 - {18} (*Hardness and completeness*). A set A is \leq_m^P -hard for the class \mathcal{C} if for every $B \in \mathcal{C}$, $B \leq_m^P A$. A set A is \leq_m^P -complete for the class \mathcal{C} , or alternatively \mathcal{C} -complete if $A \in \mathcal{C}$ and A is \leq_m^P -hard for the class \mathcal{C} .

Definition {19} (*Log-space many-to-one reducible*). We say that a set A is log-space many-to-one reducible to a set B , written $A \leq_m^{LS} B$, if there exists $\phi \in \mathbf{FLOGSPACE}$ such that for all $s \in \Gamma^*$, $s \in A \iff \phi(s) \in B$. We define \leq_m^{LS} -hardness and \leq_m^{LS} -completeness analogously.

Definition {20} (*Oracle Turing machine*). An oracle Turing machine is an ordinary Turing machine, with an extra tape, called the query tape, and two extra states, called the query state and answer state, and associated with a polynomially length-bounded function $\phi : \{0, 1\}^* \rightarrow \{0, 1\}^*$, i.e. $\ell(\phi(s)) \leq p(\ell(s))$ {22}. In query state, it reads the string $t \in \{0, 1\}^*$ on the query tape, computes $\phi(t)$, and replaces t by $\phi(t)$ on query tape, and places the head of the Turing machine on the leftmost nonempty cell of the query tape.

{20} If M is a recognizer, we write $L(M, \phi)$ the set of string for which M^ϕ halts in accepting state.

{21} $\text{time}_{M^\phi}(s)$ counts all the regular steps of the Turing machine, but all the steps in the query state count only as one step. $\text{space}_{M^\phi}(s)$ counts the cell

that the Turing machine visits, except the ones on the input, output and query tape. We define

$$t_M(\phi, n) := \max\{\text{time}_{M^\phi}(s) : s \in \Gamma^n\}, \quad (5)$$

$$s_M(\phi, n) := \max\{\text{space}_{M^\phi}(s) : s \in \Gamma^n\}. \quad (6)$$

An oracle TM is said to run in time (space) $\psi : \mathcal{N} \rightarrow \mathcal{N}$ if $t_M(\phi, \cdot) \leq \psi(\cdot)$ for all ϕ .

Definition D1.3 - {22} (*Turing reducibility*). A set A is polynomial-time Turing reducible to a set B , written $A \leq_T^P B$, if there is a polynomial time oracle TM M which computes A when B is used as oracle.

Definition D1.4 - {22} (*Hardness and completeness*). A set A is \leq_T^P -hard for the class \mathcal{C} if for every $B \in \mathcal{C}$, $B \leq_T^P A$. A set A is \leq_T^P -complete for the class \mathcal{C} , or alternatively \mathcal{C} -complete if $A \in \mathcal{C}$ and A is \leq_T^P -hard for the class \mathcal{C} .

Lemma P1.5 - {22} (*Obvious statements*). (a) $A \leq_m^P B \Rightarrow A \leq_T^P B$.

(b) \leq_m^P -hardness implies \leq_T^P -hardness.

(c) If $A \leq_T^P B$ and $B \in P$ then $A \in P$.

(d) If $A \leq_m^P B$ and $B \in NP$ then $A \in NP$.

2.4 Polynomial hierarchy

{23} Definition of nondeterministic oracle Turing machines. We let $\mathbf{P}(\mathcal{C})$ to be the class of set that are accepted by polynomial-time deterministic oracle Turing machines using some $A \in \mathcal{C}$ as oracle. We define accordingly $\mathbf{NP}(\mathcal{C})$, $\mathbf{PSPACE}(\mathcal{C})$, etc. with the obvious meaning.

{23} Let \mathcal{C} be a class of sets, we let $\text{co-}\mathcal{C}$ be the class of sets such that $\bar{A} \in \mathcal{C}$.

Definition D1.6 - {23} (*Polynomial-time hierarchy*).

$$\Sigma_0^P = \Pi_0^P = \Delta_0^P = P \quad (7)$$

$$\Sigma_{k+1}^P = \mathbf{NP}(\Sigma_k^P) \quad (8)$$

$$\Pi_k^P = \text{co-}\Sigma_k^P \quad (9)$$

$$\Delta_{k+1}^P = \mathbf{P}(\Sigma_k^P) \quad (10)$$

$$\mathbf{PH} = \bigcup_{k=0}^{\infty} \Sigma_k^P. \quad (11)$$

Lemma P1.7 - {23} (*First inclusions*).

$$\Sigma_k^P \cup \Pi_k^P \subseteq \Delta_{k+1}^P \subseteq \Sigma_{k+1}^P \cap \Pi_{k+1}^P \subseteq \mathbf{PSPACE}. \quad (12)$$

Theorem T1.10 - {24} (*Formula for elements in Σ_k^P*). $A \in \Sigma_k^P \forall s \in A$ iff there exists a polynomial-time computable predicate R , and p a polynomial such that

$$s \in A \iff \exists t_1 \forall t_2 \exists t_3 \dots \forall t_{k-1} \exists t_k R(s, t_1, t_2, \dots, t_k), \quad (13)$$

with $\ell(t_1), \ell(t_2), \dots, \ell(t_k) \leq p(\ell(s))$.

2.5 Relativization

Definition {26} (*Relativized polynomial-time hierarchy*). Let A be a set. We define

$$\Sigma_0^P(A) = \Pi_0^P(A) = \Delta_0^P(A) = P(A) \quad (14)$$

$$\Sigma_{k+1}^P(A) = \mathbf{NP}(\Sigma_k^P(A)) \quad (15)$$

$$\Pi_k^P(A) = \text{co-}\Sigma_k^P(A) \quad (16)$$

$$\Delta_{k+1}^P(A) = \mathbf{P}(\Sigma_k^P(A)) \quad (17)$$

$$\mathbf{PH}(A) = \bigcup_{k=0}^{\infty} \Sigma_k^P(A). \quad (18)$$

See Theorem 1.12 for many solutions to $P=NP$ and other similar problems, in relativized version.

{26} Let \mathcal{S} be the set of all subsets of $\{0,1\}^*$, and $\Phi : \mathcal{S} \rightarrow \mathcal{S}$. We say that Φ is polynomial-time computable if there exists a polynomial-time oracle Turing machine M such that for each $A \in \mathcal{S}$, M^A accepts $\Phi(A)$. {27} We define analogously \mathbf{NP} -computable, \mathbf{PSPACE} -computable and Σ_k^P -computable operators. We denote each of these classes \mathbf{P}_{op} , \mathbf{NP}_{op} , \mathbf{PSPACE}_{op} and $\Sigma_{k,op}^P$.

2.6 Probabilistic Complexity Classes

{28} A probabilistic TM is a TM that receives an input $s \in \{0,1\}^*$ and a sequence of coin flips $\alpha \in \{0,1\}^\infty$. The transfer function of the machine can sometimes have up to two choices available. In the case that the machine has a choice, it is made by reading the first bit of α . Then, this first bit is popped.

{28} We suppose that $\alpha \in \{0,1\}^\infty$ is generated according to the uniform distribution on $\{0,1\}^\infty$. We say that M accepts s if the probability of accepting s is greater than $1/2$. \mathbf{PP} is the set of sets accepted by polynomial-time probabilistic Turing machines. \mathbf{BPP} , \mathbf{RP} and \mathbf{ZPP} are on page {29}.

Many more results are on page {30}.

2.7 Complexity of counting

Definition D1.18 - {31} (*Number of paths*). A function $\phi : \Gamma^* \rightarrow \mathcal{N}$ is in the class $\#P$ if there is a polynomial-time nondeterministic TM M such that for each $s \in \Gamma^*$, $\phi(s)$ is the number of accepting computations of $M(s)$.

Many results on that {31 – 32}.

2.8 One-way functions

Definition {32} (*One way function*). ϕ is polynomially honest if $\ell(s) \leq p(\ell(\phi(s)))$. A one way function is a polynomially honest function that is polynomial time computable, one-to-one, and whose inverse is not polynomial-time computable.

Definition D1.22- {33} (*Unambiguous polynomial-time*). $A \in \mathbf{UP}$ if there is a polynomial-time nondeterministic Turing machine M computing A with exactly one accepting computation path. $\mathbf{P} \neq \mathbf{UP}$ is equivalent to the existence of one-way functions.

Relativized results on \mathbf{UP} are discussed in page {34}.

2.9 Polynomial-size circuits and sparse sets

Definition D1.28 - {36} (*Advice functions*). Let \mathcal{C} be a class of sets and Φ be a class of functions $\Phi : \mathcal{N} \rightarrow \Gamma^*$. The class \mathcal{C}/Φ is the class of sets A for which there exist a set $B \in \mathcal{C}$ and a function $\phi \in \Phi$ such that for all $s \in \Gamma^*$, $s \in A \iff \langle s, \phi(\ell(s)) \rangle \in B$.

Definition 1. {36} [*Sparse set, tally set*] A set $S \subset \Gamma^*$ is sparse if there is a polynomial p such that $\#\{s \in S : \ell(s) = n\} \leq p(n)$. A tally set is a set $T \subset \{0\}^*$.

Definition P1.29 - {36} (*Characterisation of polynomial circuit sets*).

$$\mathbf{P}(\mathbf{SPARSE}) = \mathbf{P}(\mathbf{TALLY}) = \mathbf{P}/\text{poly} \quad (19)$$

Many final discussions on relativized $\mathbf{P} = \mathbf{NP}$, and sparse sets used as $\mathbf{P} = \mathbf{NP}$ bridges are on pages {37 – 39}.

3 Computational Complexity of Real Functions

3.1 Computational complexity of real numbers

{41} We denote by \mathcal{D} the dyadic numbers. $\ell(s)$ is the total length of a representation s of a dyadic number, and $\text{prec}(s)$ is the length after the dot.

Definition D2.1 - {42} (*Representations of real numbers*). (a) A function $\phi : \mathcal{N} \rightarrow \mathcal{D}$ is said to binary converge towards $x \in \mathbb{R}$ if $\prec(\phi(n)) = n$ and $|\phi(n) - x| \leq 2^{-n}$. CF_x is the set of all functions that binary converge towards x .

(b) $LC_x = \{d \in \mathcal{D} \mid d < x\}$ is called the Dedekind left-cut of x .

(c) $BE_x : \mathcal{N} \rightarrow \{0, 1\}$ is such that

$$x = \sum_{i=1}^{\infty} BE_x(i) 2^{-i}. \quad (20)$$

{43} A real number x is computable \iff there is a computable function in $CF_x \iff LC_x$ is computable $\iff BE_x$ is computable.

{44} Computable real numbers form a closed field. Equality testing is incomputable.

{47} We defined \mathbf{P}_{CF} , \mathbf{P}_{BE} , \mathbf{P}_{LC} to be the polynomial-time complexity classes associated with the CF, BE and LC representations. It is shown that $P_{LC} = P_{BE} \subsetneq P_{CF}$. {49} Moreover, P_{CF} is a closed field, while $P_{LC} = P_{BE}$ is not even closed under addition. So CF is the best representation to study real numbers complexity.

3.2 Computable real functions

Definition D2.11 - {51} (*Computable real function*). Let $f : \mathbb{R} \rightarrow \mathbb{R}$. f is computable if there exists an oracle Turing machine M such that for all $x \in \mathbb{R}$ and all $\phi \in CF_x$, $n \mapsto M^\phi(0^n) \in CF_{f(x)}$. We can restrict the notion of computability on interval.

Definition D2.12 - {52} (*Modulus functions*). Let $f : [a, b] \rightarrow \mathbb{R}$ be a continuous function. A function $m : \mathcal{N} \rightarrow \mathcal{N}$ is said to be a modulus function of f if for all $n \in \mathcal{N}$, and all $x, y \in [a, b]$, we have

$$|x - y| \leq 2^{-m(n)} \Rightarrow |f(x) - f(y)| \leq 2^{-n}. \quad (21)$$

Theorem T2.13 - {52} (*Uniform modulus theorem*). If $f : [a, b] \rightarrow \mathbb{R}$ is computable, then f is continuous and has a recursive modulus functions.

Theorem C2.14 - {53} (*Characterization of Computable functions*). A real function $f : [a, b] \rightarrow \mathbb{R}$ is computable iff there exists two recursive functions $m : \mathcal{N} \rightarrow \mathcal{N}$ and $\psi : (\mathcal{D} \cap [a, b]) \times \mathcal{N} \rightarrow \mathcal{D}$ such that

- (i) m is a modulus for f ,
- (ii) for all $d \in \mathcal{D} \cap [a, b]$ and all $n \in \mathcal{N}$, $|\psi(d, n) - f(d)| \leq 2^{-n}$.

3.3 Complexity of real functions

Definition D2.18 - {57} (*Time Complexity of Real functions*). Let G be a closed bounded interval or \mathbb{R} . Let M be an oracle Turing machine, and let

$$\text{time}'_M(x, n) := \max\{\text{time}_M(\phi, n) : \phi \in CF_x\}, \quad \forall x \in G, \forall n \in \mathcal{N}. \quad (22)$$

Let $f : G \rightarrow \mathbb{R}$ be a computable function. We say that the time complexity of f on G is bounded by a function $t : G \times \mathcal{N} \rightarrow \mathcal{N}$ if there exists an oracle TM M which computes f , such that for all $x \in G$ and all $n > 0$,

$$\text{time}'_M(x, n) \leq t(x, n). \quad (23)$$

We say that f has uniform time complexity $t'(n)$ if the time complexity $t(x, n)$ of f satisfies $t(x, n) \leq t'(n)$, for all $x \in G$. We say that f has nonuniform time complexity $t'(n)$ if the time complexity $t(x, n)$ of f satisfies $t(x, n) \leq t'(n)$, for all $x \in [-2^n, 2^n] \cap G$.

Definition D2.18 - {58} (*Space Complexity of Real functions*). Let $f : G \rightarrow \mathbb{R}$ be a computable function. We say that the space complexity of f on G is bounded by a function $s : G \times \mathcal{N} \rightarrow \mathcal{N}$ if there exists an oracle TM M which computes f , such that for all $x \in G$ and all $n > 0$,

$$\text{space}_M(\phi, n) \leq s(x, n), \quad \forall \phi \in CF_x. \quad (24)$$

We say that f has uniform space complexity $s'(n)$ if the time complexity $s(x, n)$ of f satisfies $s(x, n) \leq s'(n)$, for all $x \in G$.

We denote by $P_{C[a,b]}$ the set of polynomial-time computable real functions.

Theorem T2.19 - {58} (*Time complexity and modulus*). If $f : [a, b] \rightarrow \mathbb{R}$ has uniform time complexity $t(n)$, then $t(n+2)$ is a modulus function for f .

Generalizations to multiple variables appear on pages {58–62}. Especially, the generalization implies that there is one oracle for each input variable. However, the characterization is NOT in terms of linear functions: it is a polynomial of degree 1.

3.4 Recursively Open sets

Definition D2.29 - {63} (*Partial computable real functions*). Let $S \subseteq \mathbb{R}$, and $f : S \rightarrow \mathbb{R}$. f is said to be partial computable if there exist an oracle TM M such that

- (i) for all $x \in S$, and all $\phi \in CF_x$, $M^\phi(n)$ halts and $|M^\phi(n) - f(x)| \leq 2^{-n}$, and
- (ii) for all $x \notin S$ and all $\phi \in CF_x$, $M^\phi(n)$ does not halt for all n .

Definition D2.30 - {63} (*Recursively open set*). A set $S \subseteq \mathbb{R}$ is recursively open if $S = \emptyset$ or if there exists a recursive function $\phi : \mathcal{N} \rightarrow D$ such that

- (i) for each $n \in \mathcal{N}$, $\phi(2n) < \phi(2n+1)$, and
- (ii) $S = \bigcup_{n=0}^{\infty} (\phi(2n), \phi(2n+1))$,

A set S is recursively closed if $\mathbb{R} - S$ is open.

Theorem T2.31 - {63} (*Characterization of recursively open sets*). A set $S \subseteq \mathbb{R}$ is recursively open iff there is a partial computable function f whose domain is S

Lemma C2.32 - {64} (*Continuity of partial computable functions*). Partial computable functions are continuous on their domain.

Definition D2.33 - {64} (*r.e. numbers*). A real number x is left computable enumerable (left r.e.) if $(-\infty, x) \cap D$ is r.e., and a real number is right recursively enumerable (right r.e.) if $(x, \infty) \cap D$ is r.e.

Theorem T2.34 - {65} (*Characterization of r.e.o. sets*). (a) If $x < y$, x is right r.e. and y is left r.e., then (x, y) is r.e.o.

(b) If $S \subset \mathbb{R}$ is a recursively open set, then for each component (x, y) of S (i.e., $(x, y) \subseteq S$ but $x \notin S$ and $y \notin S$), x is right r.e. and y is left r.e.

Definition D2.35 - {67} (*Computable functional*). A numerical function F on $D \subset C[0, 1]$ is computable if there exists a TM M such that for all $f \in D$, given two oracle m and ϕ representing f ,

$$|M^{m, \phi}(n) - F(f)| \leq 2^{-n}. \quad (25)$$

Definition D2.35 - {67} (*Computable operator*). A numerical operator F on $D \subset C[0, 1]$ is computable if there exists a TM M such that for all f , and all $x \in [0, 1]$, given two oracle m and ϕ representing f , and ψ representing x , then

$$|M^{m, \phi, \psi}(n) - F(f)(x)| \leq 2^{-n}. \quad (26)$$

Definition D2.35 - {67} (*Polynomial-time computable functional*). A computable function F is polynomial time if there exists p, q polynomials such that for all $f \in D$ represented by m, ϕ , M_F computes in time $p(m(0, 1, q(n)))$. We define similarly the complexity of operators.

4 Maximization

Theorem T3.1 - {72} (*Maximal points of computable functions*). Let S be a nonempty subset of $[0, 1]$. Then, the following are equivalent

- (a) S is recursively closed,
- (b) there exists a computable function $f : [0, 1] \rightarrow \mathbb{R}$ whose maximum points are exactly S ,
- (c) there exists a polynomial-time computable function $f : [0, 1] \rightarrow \mathbb{R}$ whose maximum points are exactly S .

Lemma C3.2/3 - {75} (*Recursiveness of maximum points*). Let f be a polynomial-time computable function on $[0, 1]$.

- (a) f has at least one left r.e. and one right r.e. maximum point,
- (b) If x is an isolated maximum point of f , then x is recursive,
- (c) If f has finitely many maximum points, they all are recursive,
- (d) If f has countably infinitely many maximum points, then infinitely many of them are recursive.

However, there exists a polynomial-time computable function f with uncountably many maximum points, all being nonrecursive.

Definition {76} (*Max function*). For $\phi : \{0, 1\}^* \rightarrow \{0, 1\}^*$, we define

$$\max_{\phi}(s) := \max\{\phi(\langle t, s \rangle) : \ell(t) = \ell(s)\}, \quad (27)$$

$$\max'_{\phi}(n) := \max\{\phi(\langle t, s \rangle) : \ell(t) = n\}. \quad (28)$$

Theorem P3.4/5 - {76} (*NP-hardness of maximization*). The problem of computing \max_{ϕ} is **NP**-complete, and \max'_{ϕ} is **NP**₁-hard.

In pages {81 – 99}, the authors establish the notion of NP real numbers, and make links with maximizing polynomial-time computable functions. In pages {100 – 105}, they establish a hierarchy of function classes by passing to the maximum and minimum, which establishes a relation with the usual polynomial hierarchy.

References

Ker-I Ko. *Complexity Theory of Real Functions*. Birkhäuser, 1991.