

Chapter 4: Uncomputable functions and the halting problem

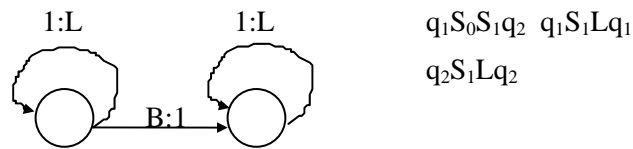
There are some functions that cannot be computed using a Turing machine. The chapter provides a **general argument** that uncomputable functions exist, and **three examples** of uncomputable functions.

General argument. The set of Turing machines is enumerable. So the set of Turing-computable functions $f:P \rightarrow P$ is enumerable. But (chapter 2) the set of all functions $f:P \rightarrow P$ is not enumerable.

I. Example 1: diagonal function**1. Effective list (or coding) of all Turing-computable 1-place functions.**

Step 1: A canonical representation for Turing machines.

a) Start with the quadruple representation.



b) Add a halting state with no instructions, if needed, as the highest-numbered state.

c) Ensure that a quadruple is present for each state (except the halted state) and each possible scanned symbol (both B and 1). (Add quadruples that do nothing, if needed.)

$q_1S_0S_1q_2$ $q_1S_1Lq_1$
 $q_2S_0S_0q_3$ $q_2S_1Lq_2$

d) Abbreviate by dropping the first two symbols in the quadruple, which are obvious from position, and write as a single list:

$S_1q_2, Lq_1, S_0q_3, Lq_2$

Step 2: Code each Turing machine with a positive integer.

a) Convert the canonical representation to a finite sequence of numbers using 1-4 for the overt action [$S_0 = 1, S_1 = 2, L = 3, R = 4$] and i for state i .

2,2,3,1,1,3,3,2

Important: no two machines yield the same list.

b) Convert the finite sequence to a single number, using powers of the prime numbers taken in increasing order:

$$2^2 \cdot 3^2 \cdot 5^3 \cdot 7 \cdot 11 \cdot 13^3 \cdot 17^3 \cdot 19^2$$

c) Place all Turing Machines in order by their code number. This gives an effective listing:

M: M_1, M_2, \dots

and also an effective list of all 1-place Turing-computable functions

L: f_1, f_2, \dots

2. Diagonalization.

Given the above list L, define the *diagonal function* d as follows:

$$d(n) = \begin{cases} 2 & \text{if } f_n(n) \text{ is defined and } = 1 \\ 1 & \text{otherwise} \end{cases}$$

Then d is a total function from $P \rightarrow P$, but d is distinct from each entry on the list and hence is not Turing-computable.

1) If Turing's thesis is correct, d is not effectively computable.

2) Puzzle: why isn't d computable? (Halting problem)

4) Second puzzle: take the list of functions f_1, f_2, f_3, \dots that you get from the above enumeration of Turing machines, but omitting all the partial functions. Define d as before, but using this new list. Then d is total but distinct from each f_n and hence not Turing-computable. Why not?

II. Example 2: Halting function

With reference to the above list M of Turing machines, define the *halting function* $h(m, n)$ by:

$$h(m, n) = \begin{cases} 1 & \text{if machine } M_m \text{ halts when started with input } n \\ 2 & \text{if machine } M_m \text{ does not halt when started with input } n. \end{cases}$$

- *Halting problem*: find an effective procedure that, given any Turing machine M and any input n , will enable us to determine whether M will halt given input n . [This amounts to showing that h is an effectively computable function.]
- If h were effectively computable, then d would be effectively computable. So if we use Turing's thesis, we know that h is not Turing-computable.

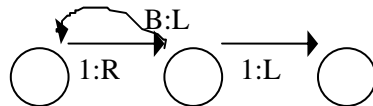
Here, we give a different direct argument that h is not Turing-computable.

Theorem 4.2: The halting function h is not Turing computable.

Proof:

Step 1: The *copying machine*, C . Given input n [n 1's], outputs two blocks of n 1's separated by a blank, halting at leftmost 1. *Exercise:* C can be implemented as a Turing machine.

Step 2: The *dithering machine*, D . Given input n , this machine halts (at leftmost 1) if $n > 1$, but otherwise it never halts.



Step 3: Assume, for a contradiction, that there is a machine H that computes h . "Glue together" the machines for H and C . The combined machine (call it G) computes the *self-halting function*

$$g(n) = h(n, n) = \begin{cases} 1 & \text{if } M_n \text{ halts on input } n \\ 2 & \text{if } M_n \text{ does not halt on input } n. \end{cases}$$

Now glue on D to produce the machine M . If M is started on input n :

$$\begin{aligned} M \text{ halts} & \leftrightarrow g(n) = 2 & \leftrightarrow M_n \text{ does not halt on input } n \\ M \text{ does not halt} & \leftrightarrow g(n) = 1 & \leftrightarrow M_n \text{ halts on input } n \end{aligned}$$

We have a contradiction. M is M_m for some m . But if M_m halts on input m , then $M (= M_m)$ does not halt on m (a contradiction); and if M_m does not halt on m , $M (= M_m)$ does halt on m (also a contradiction).

III. Example 3: The Scoring and Productivity Functions

a) **Definition of the scoring function s .**

- A k -state Turing machine is a Turing machine with k -states (not counting the halted state).

- If M is a k -state Turing machine, and we run M with input k (i.e., k 1's), define the *score* of M as follows:

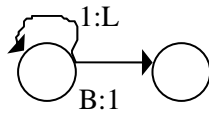
$$\text{Score of } M = \begin{cases} 0 & \text{if } M \text{ does not halt or halts out of standard final position} \\ n & \text{if } M \text{ halts in standard position with output } n \end{cases}$$

- Define $s(k)$ = highest score achieved by any k -state Turing machine.

b) The scoring function s is not Turing computable.

Proof:

i) If s is Turing-computable, then so is t given by $t(k) = s(k) + 1$. Just glue the following on to the flow diagram for s (letting the leftmost node be the former halting state for s):



ii) t is not Turing-computable. For suppose the k -state machine M computes t . Then if we run M with input k , we halt with output $t(k)$. So the score of M is $t(k)$, which is impossible because $t(k) = s(k) + 1$, but $s(k)$ is the highest score achievable by a k -state Turing machine.

Analysis: Why isn't s effectively computable by surveying the finitely many k -state machines? At any given time, we may not know whether or not all the k -state machines that are going to halt have halted. Again, we come back to the halting problem.

c) The productivity (busy-beaver) function p is not Turing-computable.

- Start a Turing machine M on a blank tape. Define the productivity of M as follows:

$$\text{productivity of } M = \begin{cases} 0, & \text{if } M \text{ does not halt or halts out of s.f.p.} \\ n, & \text{if } M \text{ halts in s.f.p. with output } n \end{cases}$$

- Define the *productivity function* p as follows:

$$p(n) = \text{productivity of the most productive } n\text{-state machine (not counting halted state)}$$

Claim: p is not Turing-computable.

Stage 1: Facts about productivity and p

Fact 1: $p(1) = 1$.

Proof: There are finitely many 1-state machines that use just B and 1. We take a survey (using the flow-graphs) and show that the maximum productivity is 1. There are always two arrows coming out of state 1, but the machines differ in whether 0, 1 or 2 arrows come back to state 1, and in what is printed on the arrows.

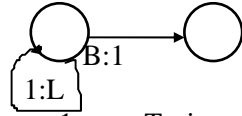
- 0 arrows: productivity is 0 (most of the time) or 1 (if it writes a single 1).
- 2 arrows back to q_1 . These machines never halt (because there is always a next instruction to follow), so all have productivity 0.
- 1 arrow back to q_1 .

Could be labeled "1:...", where "..." is R or L or B or 1. Never executed, since the arrow pointing to the halted state q_2 is labeled "B: ...". The machine has productivity 0 or 1.

The arrow could be labeled "B:...", where "..." is R or L or B or 1. Either the machine never halts and productivity is 0, or (if B:1) it prints a 1 and then follows the other arrow to the halted state; if the instruction is 1:1, the productivity is 1 and otherwise (1:B or 1:L or 1:R) the productivity is 0.

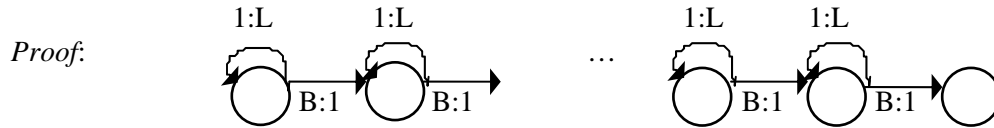
Fact 2: $p(n+1) > p(n)$

Proof: Take an n -state machine of maximum productivity. Modify its last (halted) node:



This is an $n+1$ -state Turing machine with productivity $p(n) + 1$. So $p(n+1) > p(n)$.

Fact 3: There is an n -state machine (not counting halted state) that begins on a blank tape, writes n 1's, and halts at the leftmost 1.



Fact 4: $p(n+11) \geq 2n$ for all n .

Proof: Take an n -state machine M_n that writes n 1's (by Fact 3). On the right, glue the 11-state doubler of chapter 3, so that the halted state of M_n is modified to look like state 1 of the doubler. The combined machine takes a blank input, produces a block of n 1's which it feeds to the doubler routine, and finally halts with a block of $2n$ 1's. So its productivity is $2n$. Since this machine has $n+11$ states, this shows that $p(n+11) \geq 2n$.

Step 2: Proof that the busy beaver problem is unsolvable

Suppose there is a machine BB that has k states (not counting the halted state) and computes $p(n)$: starting on leftmost on n 1's, it halts in the halted state scanning leftmost of $p(n)$ 1's.

First: $p(n+2k) \geq p(p(n))$ for any n . (1)

As in Fig. 4-3: string "Write n 1's", "BB" and "BB" together (plus a final halted state added at the end). This is an $n+2k$ -state machine that writes $p(p(n))$ 1's. This machine has $n+2k$ states. So we've shown that $p(n+2k) \geq p(p(n))$.

Second: since p is strictly increasing (from Fact 2), it must be that $n+2k \geq p(n)$, for all n ; (2)

otherwise, the inequality just proven would be wrong.

Third: to see that (2) is wrong, substitute " $n+11$ " for n in (2) to get:

$$n+11+2k \geq p(n+11)$$

But $p(n+11) \geq 2n$, so for all n we have $n+11+2k \geq 2n$,

which implies that for all n ,

$$11+2k \geq n. \quad (3)$$

But (3) is absurd! If $n = 12 + 2k$, for example, we get $11 \geq 12$, which is plainly false. ////

So p is not computable by a Turing machine in standard form. (Again, the halting problem is to blame.)