

Automata and Formal Languages

October 16, 2019

1 Register Machines and Computability

Books: PTJ (Chapter 4)

NOTE: HERE $\mathbb{N} = \{0, 1, 2, \dots\}$

A **register machine (RM)** consists of:

1. A sequence of **registers** R_1, R_2, R_3, \dots where at discrete time steps $t = 0, 1, 2, \dots$ have $R_i(t) \in \mathbb{N}$. In fact, we only have finitely many registers, and regard $R_i \equiv 0$ for all $i \geq I$.
2. A finite **program** consisting of a fixed number of **states** S_0 (HALT), S_1 (START), S_2, \dots, S_n . Each state comes with a fixed instruction performed when in state S_i . When the computer reaches HALT, we get the output from R_1 . Otherwise, for $1 \leq i \leq n$ we have 2 types of **commands**:
 - (a) Increment R_j , then move to state S_k . We write this $S_i : (j, +, k)$.
 - (b) If $R_j \neq 0$ then decrement R_j , then move to state S_k . Otherwise move to state S_l . We write this $S_i : (j, -, k, l)$.

A **sequence of instructions** for a RM is the ordered list of the instructions for the program. An **input** for a RM is, for some $k \geq 1$, a finite k -tuple $(n_1, \dots, n_k) \in \mathbb{N}^k$ which are the initial values of R_1, \dots, R_k . The other registers are set to 0.

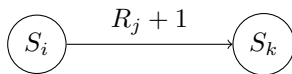
A **program diagram** for a RM is a directed graph with vertices being the states of the machine and the labelled arrows denote the instructions: $S_i : (j, +, k)$

We can then use these to describe programs:

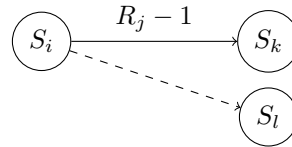
For any $k > 0$ a program P **halts** on input $(m_1, m_2, \dots, m_k) \in \mathbb{N}^k$ if it ever reaches state S_0 , written $P(m_1, \dots, m_k) \downarrow$

The **halting set** $\Omega(P)$ is a set of inputs on which P halts.

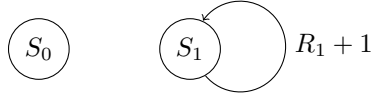
$$\Omega(P) = \cup_{k>0} \{(m_1, \dots, m_k) : P(m_1, \dots, m_k) \downarrow\}$$



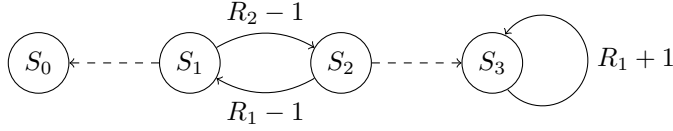
(a) $S_i : (j, +, k)$



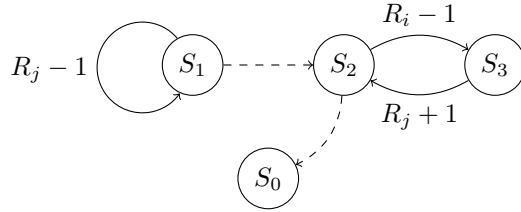
(b) $S_i : (j, -, k, l)$



(a) Repeatedly increment R_1 , never halting



(b) For input (n_1, n_2) returns $n_1 - n_2$ if $n_1 \geq n_2$, else never halt



(c) Transfer R_i to R_j , emptying R_i

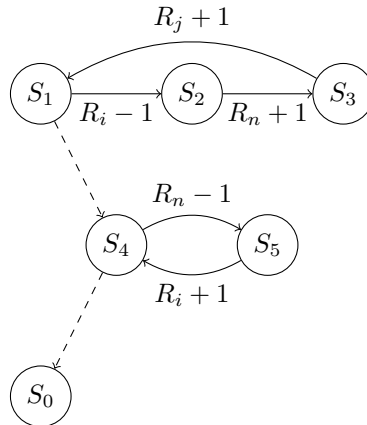
If P does not halt, we write $P(m_1, \dots, m_k) \uparrow$.

For each program P , the **upper register index** $\text{upper}(P)$ is the largest index of a register appearing in the instructions for P . So if $i > \text{upper}(P)$ then R_i never changes.

A **partial function** $f : \mathbb{N}^k \rightarrow \mathbb{N}$ is one where the domain of f is a subset of \mathbb{N}^k , and undefined otherwise. If f is defined everywhere then we call it a **total function**. This lets us define these programs as functions - we say f is **partial computable** by a program P such that $\forall (m_1, \dots, m_k) \in \text{dom}(f)$ have $P(m_1, \dots, m_k) \downarrow$ with $f(m_1, \dots, m_k) = R_1$ on halting, and $\forall (m_1, \dots, m_k) \notin \text{dom}(f)$ we have $P(m_1, \dots, m_k) \uparrow$. Hence any program P and $k > 0$ gives a partial function $f : \mathbb{N}^k \rightarrow \mathbb{N}$.

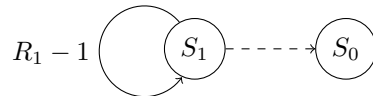
Lemma 1.1. *We can add R_i to R_j leaving R_i unchanged.*

Proof.



Thus by setting $(i = 2, j = 1)$ we see that $(n_1, n_2) \mapsto n_1 + n_2$ is total computable. \square

We have already seen that the function $n \mapsto 0$ is also computable. This can be done with the machine:



Corollary 1.2. *There exists a routine which can copy R_i to R_j leaving R_i unchanged.*

Proof. First empty R_j , then use **1.1** to add R_i to R_j . □

We can use these as subroutines to join with other programs P . Use registers R_n s.t. $n > \text{upper}(P)$ and largest input register. Then replace the halt state of P with the start state of the subroutine. In fact we have already done this - if you look carefully at the adding machine, you can see that the middle section is the same as the machine in (c) of the examples - this is the part where we replace the value in R_i from its temporary location in R_n .

Partial Recursive Functions

Partial computable functions have good closure properties.

Theorem 1.3.

1. For $i \leq k$, the **projection function** $(n_1, \dots, n_k) \mapsto n_i$ is computable.
2. The zero function $n \mapsto 0$ and **successor function** $n \mapsto n + 1$ are computable
3. (Composition) If $f : \mathbb{N}^k \rightarrow \mathbb{N}$ and $g_1, \dots, g_k : \mathbb{N}^l \rightarrow \mathbb{N}$ are all partial computable then so is the composition function $h(n_1, \dots, n_l) = f(g_1(n_1, \dots, n_l), \dots, g_k(n_1, \dots, n_l))$ where defined. If f, g_1, \dots, g_k are total functions, so is h .
4. (Recursion) If f on k variables and g on $k + 2$ variables are partial computable, then so is the partial function $h : \mathbb{N}^{k+1} \rightarrow \mathbb{N}$ defined inductively as:

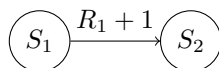
$$\begin{aligned} h(n_1, \dots, n_k, 0) &= f(n_1, \dots, n_k) \\ h(n_1, \dots, n_k, n_{k+1} + 1) &= g(n_1, \dots, n_{k+1}, h(n_1, \dots, n_{k+1})) \end{aligned}$$

Moreover, f, g total $\implies h$ total.

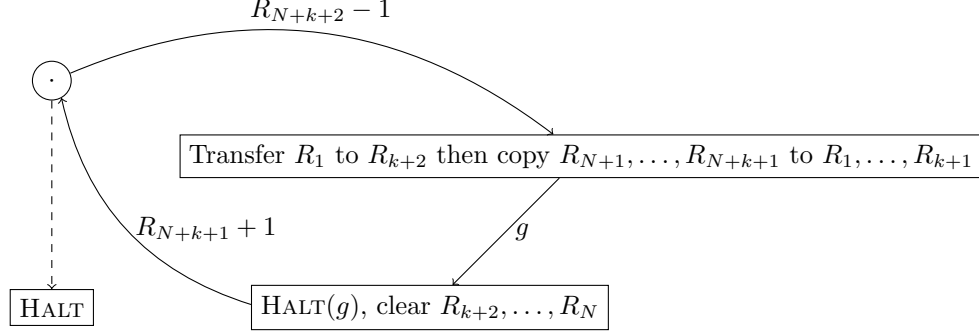
5. (Minimisation) If f on $k + 1$ variables is partial computable then so is the partial function $g : \mathbb{N}^k \rightarrow \mathbb{N}$ defined by $g(n_1, \dots, n_k) = n$ if $f(n_1, \dots, n_k, n) = 0$ and $f(n_1, \dots, n_k, m) > 0$ for all $m < n$, and is undefined if no zero is ever found. Note that f total $\nRightarrow g$ total.

Proof.

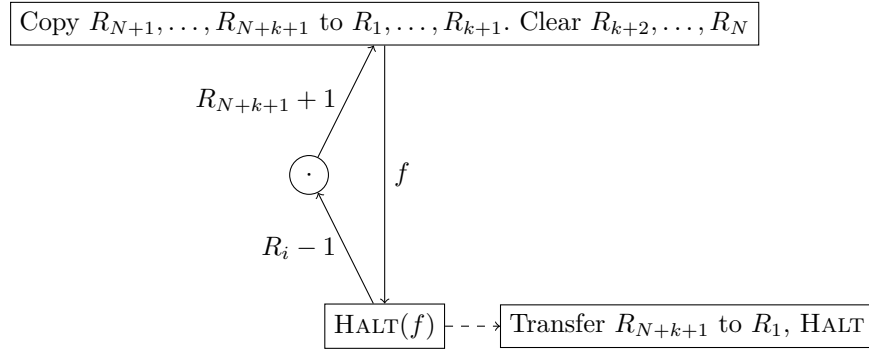
1. We can use the program Transfer R_i to R_1 , HALT.
2. Zero function has already been seen. For successor function, use:



3. First transfer R_1, \dots, R_l to R_{N+1}, \dots, R_{N+l} where N is large enough to not be needed in other subroutines. Then for each $1 \leq i \leq k$ in turn, copy R_{N+1}, \dots, R_{N+l} to R_{k+1}, \dots, R_{k+l} , perform g_i but with all registers shifted up by k and then transfer answer from R_{k+1} to R_i , then clear R_{k+2}, \dots, R_N . Finally, apply f .
4. Copy R_1, \dots, R_k to R_{N+1}, \dots, R_{N+k} , transfer R_{k+1} to R_{N+k+2} ("counts down"), then do f . Then:



5. Copy R_1, \dots, R_k to R_{N+1}, \dots, R_{N+k} . Then



□

The class of **partial recursive functions** is the smallest class of partial functions from \mathbb{N}^k to \mathbb{N} over all $k \geq 1$ closed under the operations **1.3** (1) to (5). That is, f can be constructed from basic functions and applications of (3), (4), (5) a finite number of times.

So **1.3** says that partial recursive \implies partial computable.

A partial function is **primitive recursive** if we never use **1.3** (5) its construction. Note that primitive recursive \implies total recursive, as (5) was the only construction that breaks the totality of the function. [The converse implication is not true: the Ackermann function.]

Example: $+$ and \times are primitive recursive:

$+$: Let $h(m, 0) = m$, $h(m, n+1) = h(m, n) + 1 = g(m, n, h(m, n))$, where $g(x, y, z) = z + 1$.

\times : $H(m, 0) = 0$, $H(m, n+1) = H(m, n) + m = g(m, n, H(m, n))$ for $g(x, y, z) = x + z$.

Example: $(m, n) \mapsto m^n$ is primitive recursive - left as exercise.

We need to be able to “encode” finite sequences of arbitrary length in \mathbb{N} . For $n > 0$ and $i \in \mathbb{N}$, write p_i for the $(i + 1)$ th prime (so $p_0 = 2$). Write $(n)_i$ for the largest power of the prime p_i that divides n .

Lemma 1.4. *For each fixed i , the 1 variable function $(\cdot)_i : \mathbb{N} \rightarrow \mathbb{N}$ is primitive recursive.*

Proof. First note that, for any finite sequence $(m_0, m_1, \dots, m_s) \subseteq \mathbb{N}^{s+1}$, the function

$$f(n) = \begin{cases} m_n & n \leq s \\ 0 & n > s \end{cases} \text{ is primitive recursive.}$$

By induction on s and recursion from **1.3** (4), for $k = 0$ if c constant and $g : \mathbb{N}^2 \rightarrow \mathbb{N}$ is primitive recursive, then so is $h(0) = c, h(n + 1) = g(n, h(n))$.

Thus, given $f : \mathbb{N} \rightarrow \mathbb{N}$ primitive recursive, let $g(n, m) := f(n)$, which is primitive recursive. So $h(0) = c, h(n + 1) = f(n)$ is primitive recursive, and we can repeat this process.

This includes for each fixed k :

1. The step function $\text{Step}_k(n) = \begin{cases} 1 & 0 \leq n \leq k - 1 \\ 0 & \text{otherwise} \end{cases}$
2. The delta function $\delta_k(n) = \begin{cases} 1 & n = k \\ 0 & n \neq k \end{cases}$ Let $\epsilon(n) = \delta_0(\delta_0(n)) = \begin{cases} 0 & n = 0 \\ 1 & n = 1 \end{cases}$ - this is also primitive recursive.
3. The slope function $\text{Slope}_k(n) = \begin{cases} n + 1 & 0 \leq n \leq k - 2 \\ 0 & \text{otherwise} \end{cases}$
4. The remainder function $\text{Rem}_k(n) = n \bmod k$ use recursion in the form $g(n, m) := f(m)$, so $h(0) = 0, h(n + 1) = f(h(n))$ primitive recursive if f is. Here, $\text{Rem}_k(n + 1) = \text{Slope}_k(\text{Rem}_k(n))$
- *5. Floor $_k(n) = \lfloor \frac{n}{k} \rfloor$
- *6. Divide $_k(n) = \begin{cases} n/k & n \equiv 0 \bmod k \\ 0 & \text{otherwise} \end{cases}$
- *7. Division by powers $\text{Power}_k(n, m) = \begin{cases} n/k^m & n \equiv 0 \bmod k^m \\ 0 & \text{otherwise} \end{cases}$
- *8. Maxpower $_k(n) = \begin{cases} 0 & n = 0 \\ \text{largest power of } k \text{ dividing } n & n \neq 0 \end{cases}$

Proofs of *ed function are on example sheet 1.

Now define by recursion $h(n, 0) = 0$ and $h(n, m + 1) = h(n, m) + \epsilon(\text{Power}_k(n, m + 1))$.

$$\epsilon(\text{Power}_k(n, j)) = \begin{cases} 1 & k^j \text{ divides } n > 0 \\ 0 & \text{otherwise} \end{cases}, \text{ so is 0 if } j \geq n$$

So $h(n, n) = \sum_{i=1}^n \epsilon(\text{Power}_k(n, i)) = \text{Maxpower}_k(n)$, so $h(n, n)$ is primitive recursive. \square

Computable = Recursive

We have seen already that partial recursive \implies partial computable.

Theorem 1.5. *Every partial computable function $f : \mathbb{N}^k \rightarrow \mathbb{N}$ is partial recursive.*

Proof. From a program P for f , define $g : \mathbb{N}^{k+2} \rightarrow \mathbb{N}$, “what actually goes on in P ”, to be the function:

$$g(n_1, \dots, n_k, 0, t) \text{ is the state of } P \text{ after time } t \text{ with input } (n_1, \dots, n_k)$$

So $t = 0$ gives 1 and if halt at t_0 then gives 0 for all $t \geq t_0$, and:

$$(n_1, \dots, n_k, i, t) \text{ is the contents of } R_i \text{ at time } t$$

So have N (assume $> k$) such that $g(\dots, i, \cdot) = 0 \forall i > N$. Note that g is a total function.

Suppose that g is recursive and define $q(n_1, \dots, n_k) = \min\{t : g(n_1, \dots, n_k, 0, t) = 0\}$. Then q is partial recursive, and so $f(n_1, \dots, n_k) = g(n_1, \dots, n_k, 1, q(n_1, \dots, n_k))$ is partial recursive.

Proof that g is recursive:

Fix n_1, \dots, n_k and t . For each $0 \leq i \leq N$, g gives $(g_0, \dots, g_N) \in \mathbb{N}^{N+1}$, encode as $c(d_0, \dots, d_N) = 2^{d_0} 3^{d_1} \dots p_N^{d_N} \in \mathbb{N}$ is primitive recursive. Also, $(c(d_0, \dots, d_N))_i = d_i$ is primitive recursive. We will define $h : \mathbb{N}^{k+2} \rightarrow \mathbb{N}$ via recursion where $h(n_0, n_1, \dots, n_k, t)$ is the coded integer of state and registers of P at time t for input n_1, \dots, n_k and start state n_0 (here = 1).

In particular, for $t = 0$, $h = 2^{n_0} 3^{n_1} \dots p_k^{n_k}$. For recursion for h , we need $s : \mathbb{N} \rightarrow \mathbb{N}$, the “transition function”, which computes in coded form the changes at each step. \square