COT5310: Formal Languages and Automata Theory

Lecture Notes #1: Computability

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1. Introduction to computability

2. Basic models of computation
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1. Introduction to Computability

- Questions and problems
- Computation (What-) problems and decision problems
- Problems and algorithms
- Hilbert’s problems
- Hilbert’s program
- Algorithms and functions
- The theory of computation
Questions:

- For what real number \( x \) is \( 2x^2 + 36x + 25 = 0 \)?
- What is the smallest prime number greater than 20?
- Is 257 a prime number?

A problem is a class of questions. Each question in the class is an instance of the problem. Examples of problems:

- Find the real roots of the equation \( ax^2 + bx + c = 0 \), where \( a, b, \) and \( c \) are real numbers. If we substitute \( a, b, \) and \( c \) by real values, we get an instance of this problem;
- Does the equation \( ax^2 + bx + c = 0 \), where \( a, b, \) and \( c \) are real numbers, have positive (real) roots? If we substitute \( a, b, \) and \( c \) by real values, we get an instance of this problem.
Most problems in mathematical sciences are of two kinds:

- **computation (what-) problems** - such a problem is to obtain the value of a function for a given argument.

  Example: What is the square root of a given positive integer $x$ to nearest thousandth?

- **decision problems (yes/no problems)** - these are problems whose answer is yes or no.

  Example: Does a given vector satisfy a given set of linear equations?

Decision problems could be considered computation problems but it is worthwhile to make the distinction.
An algorithm for a problem is an organized set of commands for answering on demand any question that is an instance of the problem.

A problem is solvable if there exists an algorithm for it. Otherwise it is called unsolvable.

A solvable decision problem is also called decidable; an unsolvable decision problem is also called undecidable.

An algorithm for a decision problem is usually called a decision algorithm (procedure).
Hilbert’s problems are a set of (originally) unsolved problems in mathematics proposed by Hilbert. Of the 23 total appearing in the printed address, 10 were actually presented at the Second International Congress in Paris on August 8, 1900. In particular, the problems presented by Hilbert were 1, 2, 6, 7, 8, 13, 16, 19, 21, and 22 (Derbyshire 2004, p. 377). Furthermore, the final list of 23 problems omitted one additional problem on proof theory (Thiele 2001).

Hilbert’s problems were designed to serve as examples for the kinds of problems whose solutions would lead to the furthering of disciplines in mathematics. As such, some were areas for investigation and therefore not strictly “problems”.
The **tenth problem** in Hilbert’s list was that of finding a decision procedure to tell whether a **diophantine equation** has a solution. Such an equation is of the form

\[ P(x_1, \ldots, x_n) = 0, \]

where \( P \) is a polynomial in the integer unknowns \( x_1, \ldots, x_n \) with integer coefficients.

At the time Hilbert posed this problem, mathematicians did not consider the possibility of proving it unsolvable. Nevertheless, when undecidability results began to appear in the 1930s and early 1940s, many researchers began to suspect that the problem was unsolvable. Much work has been done, but it was not until 1972 that the proof of the undecidability of Hilbert’s tenth problem was completed by Y. Matiyasevic.
In the early 1920s, David Hilbert put forward a new proposal for the foundation of classical mathematics which has come to be known as **Hilbert's Program**. It calls for a formalization of all of mathematics in axiomatic form, together with a proof that this axiomatization of mathematics is consistent. The consistency proof itself was to be carried out using only what Hilbert called “finitary” methods. Although Hilbert proposed his program in this form only in 1921, various facets of it are rooted in foundational work of his going back until around 1900.

In September 1930, Kurt Gödel announced his first incompleteness theorem at a conference in Königsberg, showing that **Hilbert’s program is not realizable**.
Every algorithm computes a function:

\[ \text{input } (x) \rightarrow \text{algorithm} \rightarrow \text{output } (y) \]

\[ f(x) = y \]

A function computable by some algorithm is called a \textit{computable function}.

The theory of computation is concerned with the following main question: \textbf{What functions are computable?} Equivalently, \textbf{What problems are solvable?}
• The *theory of computation* originated in the 1930s, in the work of the logicians Church, Gödel, Kleene, Post, and Turing;

• The theory of computation provides *models of computation* together with their basic properties;

• The concept of *computation* is related to that of an *algorithm*;

• Computation: numerical or non-numerical.
2. Basic Models of Computation

2.1. Recursive functions

2.2. Turing machines

2.3. Properties of recursive functions
2.1. Recursive Functions

2.1.1. Notation and terminology
2.1.2. Primitive recursive functions
   2.1.2.1. Definitions and examples
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2.1.3. Recursive functions
   2.1.3.1. Definitions and examples
   2.1.3.2. The Ackermann-Peter function is recursive
   2.1.3.3. Limits of Algorithmic Computability
   2.1.3.4. Recursive Sets. Rice’s Theorem
Recursive functions emerge from logic, and so are very useful for formalizing algorithms of which we have intuitively natural descriptions.

A bit of history:

- Richard Dedekind, in 1888, and Giuseppe Peano, in 1889, were the first who used inductive definitions to define addition, multiplication, and exponentiation on natural numbers;
- Thoralf Skolem, in 1923, and Kurt Gödel, in 1931, developed the theory of recursive functions.
2.1.1. Notation and Terminology

- $f : A \leadsto B$ means that $f$ is a **partial function** from $A$ into $B$. That is, $f$ might not be defined for some input values $a \in A$. We will write $f(a) = \bot$ or $f(a) \uparrow$ whenever $f$ is not defined in $a$ ($\bot$ means **undefined**);

- When a function $f : A \leadsto B$ is defined for all $a \in A$, we say that $f$ is a **total function** and also write $f : A \rightarrow B$;

- Let $g_i : A \leadsto B_i$, $1 \leq i \leq n$, and $h : B_1 \times \cdots \times B_n \leadsto C$. Then, applying **function composition** to the functions $h, g_1, \ldots, g_n$ yields the function $f : A \leadsto C$ given by

$$f(a) = h(g_1(a), \ldots, g_n(a)),$$

for all $a \in A$. We usually write $f = h \circ (g_1, \ldots, g_n)$.

$f$ is undefined for all values $a$ for which $g_i(a) \uparrow$ for some $i$, or $h(g_1(a), \ldots, g_n(a)) \uparrow$. 
The theory of recursive functions is primarily concerned with arithmetic functions, i.e., functions from $\mathbb{N}^n$ into $\mathbb{N}$, for some $n \geq 1$.

We adopt the following notation:

- $AF^{(n)} = \{f | f : \mathbb{N}^n \rightarrow \mathbb{N}\}$, $n \geq 1$;
- $AF = \bigcup_{n \geq 1} AF^{(n)}$;
- $TAF^{(n)} = \{f | f : \mathbb{N}^n \rightarrow \mathbb{N}\}$, $n \geq 1$;
- $TAF = \bigcup_{n \geq 1} TAF^{(n)}$;
- $S : \mathbb{N} \rightarrow \mathbb{N}$, $S(x) = x + 1$ for all $x \in \mathbb{N}$ (the successor function).
2.1.2. Primitive Recursive Functions

2.1.2.1. Definitions and examples

2.1.2.2. Algebraic properties

2.1.2.3. Primitive recursive enumerations

2.1.2.4. The Ackermann-Peter function
2.1.2.1. Definitions and Examples

Let $g \in AF^{(n)}$ and $h \in AF^{(n+2)}$, where $n \geq 0$. We say that a function $f \in AF^{(n+1)}$ is obtained from $g$ and $h$ by primitive recursion if the following hold true:

- $f(x_1, \ldots, x_n, 0) = g(x_1, \ldots, x_n)$;
- $f(x_1, \ldots, x_n, y + 1) = h(x_1, \ldots, x_n, y, f(x_1, \ldots, x_n, y))$,

for all $x_1, \ldots, x_n, y \in \mathbb{N}$.

Remark: For $n = 0$ the primitive recursion scheme is:

- $f(0) = a$, where $a \in \mathbb{N}$;
- $f(y + 1) = h(y, f(y))$,

for all $y \in \mathbb{N}$. 
Remark: If $f$ is obtained by primitive recursion from the total functions $g$ and $h$, then $f$ is a total function.

**Proof**  By mathematical induction on $y$. □
2.1.2.1. Definitions and Examples

In order to define the class of primitive recursive functions we need some functions on which to get started. These will be:

- the successor function \( S : \mathbb{N} \rightarrow \mathbb{N} \);
- the constant 0 function \( C_0^{(1)} : \mathbb{N} \rightarrow \mathbb{N} \) given by
  \[
  C_0^{(1)}(x) = 0,
  \]
  for all \( x \in \mathbb{N} \);
- the projection function \( P_i^{(n)} : \mathbb{N}^n \rightarrow \mathbb{N} \) given by
  \[
  P_i^{(n)}(x_1, \ldots, x_n) = x_i,
  \]
  for all \( x_1, \ldots, x_n \in \mathbb{N}, \ n \geq 1, \) and \( 1 \leq i \leq n \).

The functions \( S, C_0^{(1)} \) and \( P_i^{(n)} \) are called initial functions.
2.1.2.1. Definitions and Examples

The class of **primitive recursive functions** is the smallest class \( PRF \) of arithmetic functions which satisfies:

- **includes all the initial functions**;
- **it is closed under composition** (i.e., \( f \circ (g_1, \ldots, g_n) \in PRF \) whenever \( f, g_1, \ldots, g_n \in PRF \) and the composition is defined);
- **it is closed under primitive recursion** (i.e., \( f \in PRF \) whenever \( f \) is obtained from \( g \) and \( h \) by primitive recursion and \( g, h \in PRF \)).

**Remark:** \( PRF \) contains only total functions.

**Proof**  By structural induction.  \( \square \)
2.1.2.1. Definitions and Examples

Let \( m \in \mathbb{N} \) and \( n \geq 1 \). The \textbf{constant} \( m \) function \( C_m^{(n)} : \mathbb{N}^n \rightarrow \mathbb{N} \) is defined by

\[
C_m^{(n)}(x_1, \ldots, x_n) = m,
\]

for all \( x_1, \ldots, x_n \in \mathbb{N} \).

**Proposition 1** \( C_m^{(n)} \) is a primitive recursive function, for all \( m, n \in \mathbb{N} \).

**Proof**  
Case 1. \( m = 0 \) and \( n = 1 \): from definitions;

Case 2. \( m > 0 \) and \( n = 1 \): \( C_m^{(1)}(x) = \underbrace{S(\cdots S(C_0^{(1)}(x)) \cdots)}_{m \text{ times}} \)

Case 3. \( m > 0 \) and \( n > 1 \): (prove it!) \( \square \)
Example 1  The following functions are primitive recursive:

1. the addition, multiplication, and exponentiation functions;
2. the factorial function;
3. the predecessor function $P_d : \mathbb{N} \to \mathbb{N}$ defined by

$$P_d(x) = \begin{cases} 
    x - 1, & \text{if } x > 0 \\
    0, & \text{otherwise}, 
  \end{cases}$$

for all $x \in \mathbb{N}$;

4. the arithmetic (recursive) difference function $\dot{-}$ defined by

$$x \dot{-} y = \begin{cases} 
    x - y, & \text{if } x \geq y \\
    0, & \text{otherwise}, 
  \end{cases}$$

for all $x, y \in \mathbb{N}$;

5. the absolute difference function $|x - y|$;
6. the **sign** functions $sg$ and $\bar{sg}$ given by

$$sg(x) = \begin{cases} 0, & \text{if } x = 0 \\ 1, & \text{otherwise} \end{cases} \quad \bar{sg}(x) = \begin{cases} 0, & \text{if } x \neq 0 \\ 1, & \text{otherwise} \end{cases}$$

for all $x \in \mathbb{N}$;

7. the **comparison** functions $ls$ (less than), $gr$ (greater than), $eq$ (is equal to);

8. the function $[\sqrt{-}]$;

9. the function $E$ given by $E(x) = x - [\sqrt{x}]^2$, for all $x \in \mathbb{N}$;

10. the **maximum** and **minimum** functions $max$ and $min$ given by:

$$max(x, y) = \begin{cases} x, & \text{if } x \geq y \\ y, & \text{otherwise} \end{cases} \quad min(x, y) = \begin{cases} x, & \text{if } x \leq y \\ y, & \text{otherwise} \end{cases}$$

for all $x, y \in \mathbb{N}$. 
2.1.2.1. Definitions and Examples

Remark: As we can see, many of the functions normally studied in number theory, and approximations to real-valued functions, are primitive recursive.

Remark: \( sg, \overline{sg}, ls, gr, \) and \( eq \) are primitive recursive predicates.

We will denote by \( PRP \) the class of primitive recursive predicates.
2.1.2.1. Definitions and Examples

Primitive recursive functions and programs:

• We want to compute $z = C^{(1)}_{0}(x)$. The corresponding program is:

  ```pascal
  input: x;
  output: z = C^{(1)}_{0}(x);
  begin
    z := 0;
  end.
  ```

• We want to compute $z = S(x)$. The corresponding program is:

  ```pascal
  input: x;
  output: z = S(x);
  begin
    z := x + 1;
  end.
  ```
We want to compute $z = P_i(n)(x_1, \ldots, x_n)$. The corresponding program is:

input: $x = (x_1, \ldots, x_n)$;
output: $z = P_i(n)(x_1, \ldots, x_n)$;
begin
  $z := x[i]$;
end.

We want to compute $z = (h \circ (g_1, \ldots, g_n))(x)$. The corresponding program is:

input: $x$;
output: $z = (h \circ (g_1, \ldots, g_n))(x)$;
begin
  for $i := 1$ to $n$ do $y_i := g_i(x)$;
  $z := h(y_1, \ldots, y_n)$;
end.
We want to compute \( z = f(x, y) \), where \( f \) is obtained from \( g \) and \( h \) by primitive recursion. The corresponding program is:

\[
\text{input: } g, h, \text{ and } x, y \in \mathbb{N}; \\
\text{output: } z = f(x, y); \\
\text{begin} \\
\quad z := g(0); \\
\quad \text{for } i := 1 \text{ to } y \text{ do } z := h(x, i - 1, z); \\
\text{end.}
\]
2.1.2.2. Algebraic Properties

A few basic algebraic properties of primitive recursive functions are in order.

**Proposition 2** If \( f(x_1, \ldots, x_n) \) is primitive recursive and \( \varphi \) is a permutation of the set \( \{1, \ldots, n\} \), then the function \( g(x_1, \ldots, x_n) \) given by

\[
g(x_1, \ldots, x_n) = f(x_{\varphi(1)}, \ldots, x_{\varphi(n)}),
\]

for all \( x_1, \ldots, x_n \in \mathbb{N} \), is primitive recursive.

**Proposition 3** If \( f(x_1, \ldots, x_n) \) is primitive recursive, then the function \( g(x_1, \ldots, x_n, y_1, \ldots, y_m) \) given by

\[
g(x_1, \ldots, x_n, y_1, \ldots, y_m) = f(x_1, \ldots, x_n),
\]

for all \( x_1, \ldots, x_n, y_1, \ldots, y_m \in \mathbb{N} \), is primitive recursive.
Proposition 4 If \( f(x_1, \ldots, x_n) \) is primitive recursive and \( i_1, \ldots, i_k \in \{1, \ldots, n\} \), then the function \( g(x_1, \ldots, x_n) \) given by

\[
g(x_1, \ldots, x_n) = f(y_1, \ldots, y_n),
\]

where

\[
y_i = \begin{cases} x_{i_1}, & \text{if } i \in \{i_1, \ldots, i_k\} \\ x_i, & \text{otherwise} \end{cases}
\]

for all \( i \) and \( x_1, \ldots, x_n \in \mathbb{N} \), is primitive recursive.

We say that the function \( f \in AF(n) \) is 0 almost everywhere if \( f(x_1, \ldots, x_n) \neq 0 \), for finitely many inputs \( (x_1, \ldots, x_n) \in \mathbb{N}^n \).

Proposition 5 Every arithmetic function which is 0 almost everywhere, is primitive recursive.
2.1.2.2. Algebraic Properties

A function $f \in AF^{(n+1)}$ is obtained from $g \in AF^{(n+1)}$ by bounded summation if

$$f(x_1, \ldots, x_n, y) = \sum_{i=0}^{y} g(x_1, \ldots, x_n, i),$$

for all $x_1, \ldots, x_n, y \in \mathbb{N}$.

**Proposition 6** PRF is closed under bounded summation.
A function $f \in AF^{(n)}$ is obtained from $g \in AF^{(n+1)}$ and $u, v \in AF^{(n)}$ by primitive recursive summation if

$$f(x_1, \ldots, x_n) = \sum_{i=u(x_1, \ldots, x_n)}^{v(x_1, \ldots, x_n)} g(x_1, \ldots, x_n, i),$$

for all $x_1, \ldots, x_n \in \mathbb{N}$.

**Proposition 7** PRF is closed under primitive recursive summation.
A function \( f \in AF^{(n+1)} \) is obtained from \( g \in AF^{(n+1)} \) by bounded product if

\[
f(x_1, \ldots, x_n, y) = \prod_{i=0}^{y} g(x_1, \ldots, x_n, i),
\]

for all \( x_1, \ldots, x_n, y \in \mathbb{N} \).

**Proposition 8** PRF is closed under bounded product.
2.1.2.2. Algebraic Properties

A function $f \in AF^{(n)}$ is obtained from $g \in AF^{(n+1)}$ and $u, v \in AF^{(n)}$ by primitive recursive product if

$$f(x_1, \ldots, x_n) = \prod_{i=u(x_1, \ldots, x_n)}^{v(x_1, \ldots, x_n)} g(x_1, \ldots, x_n, i),$$

for all $x_1, \ldots, x_n \in \mathbb{N}$.

**Proposition 9** PRF is closed under primitive recursive product.
2.1.2.2. Algebraic Properties

A function $f \in AF^{(n+1)}$ is obtained from $g \in AF^{(n+1)}$ by bounded minimization if

- $f(x_1, \ldots, x_n, z) = \text{the least } y \leq z \text{ for which } g(x_1, \ldots, x_n, y) = 0$, if such an $y$ exists;
- $f(x_1, \ldots, x_n, z) = z + 1$, otherwise,

for all $x_1, \ldots, x_n, z \in \mathbb{N}$.

If $f$ is obtained from $g$ by bounded minimization then we write

$$f(x_1, \ldots, x_n, z) = \mu(y \leq z)[g(x_1, \ldots, x_n, y) = 0]$$

**Proposition 10** PRF is closed under bounded minimization.
2.1.2.2. Algebraic Properties

A function \( f \in AF^{(n)} \) is obtained from \( g \in AF^{(n+1)} \) and \( u \in AF^{(n)} \) by primitive recursive minimization if

- \( f(x_1, \ldots, x_n) = \text{the least } y \leq u(x_1, \ldots, x_n) \text{ for which } g(x_1, \ldots, x_n, y) = 0 \), if such an \( y \) exists;
- \( f(x_1, \ldots, x_n) = z + 1 \), otherwise,

for all \( x_1, \ldots, x_n \in \mathbb{N} \).

If \( f \) is obtained from \( g \) by primitive recursive minimization then we write

\[
 f(x_1, \ldots, x_n) = \mu(y \leq u(x_1, \ldots, x_n))[g(x_1, \ldots, x_n, y) = 0]
\]

**Proposition 11** PRF is closed under primitive recursive minimization.
A predicate $f \in AP^{n+1}$ is obtained from $g \in AP^{n+1}$ by bounded existential quantification if

- $f(x_1, \ldots, x_n, y) = 1$, if there exists $0 \leq i \leq y$ such that $g(x_1, \ldots, x_n, i) = 1$;
- $f(x_1, \ldots, x_n, y) = 0$, otherwise,

for all $x_1, \ldots, x_n, y \in \mathbb{N}$.

If $f$ is obtained from $g$ by bounded existential quantification then we write

$$f(x_1, \ldots, x_n, y) = \exists_{i=0}^{y} g(x_1, \ldots, x_n, i)$$

**Proposition 12** PRP is closed under bounded existential quantification.
2.1.2.2. Algebraic Properties

A predicate $f \in AP^{(n)}$ is obtained from $g \in AP^{(n+1)}$ and $u, v \in AF^{(n)}$ by primitive recursive existential quantification if

- $f(x_1, \ldots, x_n) = 1$, if there exists $u(x_1, \ldots, x_n) \leq i \leq v(x_1, \ldots, x_n)$ such that $g(x_1, \ldots, x_n, i) = 1$;
- $f(x_1, \ldots, x_n) = 0$, otherwise,

for all $x_1, \ldots, x_n \in \mathbb{N}$.

If $f$ is obtained from $g$ by bounded existential quantification then we write

$$f(x_1, \ldots, x_n) = \exists_{i=u(x_1, \ldots, x_n)}^{v(x_1, \ldots, x_n)} g(x_1, \ldots, x_n, i)$$

**Proposition 13** PRP is closed under primitive recursive existential quantification.
A predicate $f \in AP^{(n+1)}$ is obtained from $g \in AP^{(n+1)}$ by bounded universal quantification if

- $f(x_1, \ldots, x_n, y) = 1$, if $g(x_1, \ldots, x_n, i) = 1$ for all $0 \leq i \leq y$;
- $f(x_1, \ldots, x_n, y) = 0$, otherwise,

for all $x_1, \ldots, x_n \in \mathbb{N}$.

If $f$ is obtained from $g$ by bounded universal quantification then we write

$$f(x_1, \ldots, x_n, y) = \forall_{i=0}^{y} g(x_1, \ldots, x_n, i)$$

**Proposition 14** PRP is closed under bounded universal quantification.
2.1.2.2. Algebraic Properties

A predicate \( f \in AP^{(n)} \) is obtained from \( g \in AP^{(n+1)} \) and \( u, v \in AF^{(n)} \) by primitive recursive universal quantification if

- \( f(x_1, \ldots, x_n) = 1 \), if \( g(x_1, \ldots, x_n, i) = 1 \) for all \( u(x_1, \ldots, x_n) \leq i \leq v(x_1, \ldots, x_n) \);
- \( f(x_1, \ldots, x_n) = 0 \), otherwise,

for all \( x_1, \ldots, x_n, y \in \mathbb{N} \).

If \( f \) is obtained from \( g \) by primitive recursive universal quantification then we write

\[
f(x_1, \ldots, x_n) = \forall_{i=u(x_1, \ldots, x_n)}^{v(x_1, \ldots, x_n)} g(x_1, \ldots, x_n, i)\]

**Proposition 15** PRP is closed under primitive recursive universal quantification.
2.1.2.2. Algebraic Properties

**Proposition 16** PRP is closed under $\neg$, $\lor$, and $\land$.

The properties established so far allow us to prove the primitive recursiveness of the following important functions:

**Example 2** The following functions are primitive recursive:

1. *(integer division)* $\text{quo} : \mathbb{N}^2 \rightarrow \mathbb{N}$ given by
   - $\text{quo}(x, y) = \text{the largest } z \text{ such that } z \cdot y \leq x$, if $y \neq 0$;
   - $\text{quo}(x, y) = 0$, otherwise,
   
   for all $x, y \in \mathbb{N}$;
2.1.2.2. Algebraic Properties

2. (remainder) \( rm : \mathbb{N}^2 \rightarrow \mathbb{N} \) given by
\[
rm(x, y) = x \div y \cdot \text{quo}(x, y),
\]
for all \( x, y \in \mathbb{N} \);

3. (division test) \( \text{div} : \mathbb{N}^2 \rightarrow \mathbb{N} \) given by
\[
\text{div}(x, y) = \begin{cases} 
1, & \text{if } xy > 0 \text{ and } x \mid y \\
0, & \text{otherwise}, 
\end{cases}
\]
for all \( x, y \in \mathbb{N} \);

4. (number of divisors) \( n\text{div} : \mathbb{N} \rightarrow \mathbb{N} \) given by
\[
n\text{div}(x) = \text{number of divisors of } x,
\]
for all \( x \in \mathbb{N} \);
2.1.2.2. Algebraic Properties

5. (primality test) \( pr : \mathbb{N} \rightarrow \mathbb{N} \) given by

\[
pr(x) = \begin{cases} 
1, & \text{if } x \text{ is prime} \\
0, & \text{otherwise}, 
\end{cases}
\]

for all \( x \in \mathbb{N} \);

6. (\( n \)th prime number) \( pn : \mathbb{N} \rightarrow \mathbb{N} \) given by

\[
 pn(n) = \text{the } n\text{th prime number}, 
\]

for all \( n \in \mathbb{N} \) (2 is the 0th prime number).
2.1.2.3. Primitive Recursive Enumerations

Primitive recursive enumerations are primitive recursive encodings of tuples of natural numbers. We will study:

- **Cantor’s enumeration** – used to encode fixed length sequences of natural numbers;
- **Gödel’s numbering** – used to encode arbitrary length sequences of natural numbers.
2.1.2.3. Primitive Recursive Enumerations

Cantor’s enumeration

Cantor’s enumeration, also called Cantor’s diagonal method, is a clever technique used to show that the set $\mathbb{N}^2$ can be put into a bijective correspondence with $\mathbb{N}$.

The technique is quite general and applies in many circumstances.
2.1.2.3. Primitive Recursive Enumerations
Cantor’s enumeration is the function $J : \mathbb{N}^2 \rightarrow \mathbb{N}$ given by:

$$J(x, y) = \frac{(x + y)(x + y + 1)}{2} + x,$$

for all $(x, y) \in \mathbb{N}^2$.

**Lemma 1** Cantor’s enumeration is a primitive recursive bijection.

**Proof**

- $J$ is bijective (*prove it!*);
- $J$ is primitive recursive because it is a composition of primitive recursive functions.

$\square$
The inverse of $J$ is the function $J^{-1} : \mathbb{N} \rightarrow \mathbb{N}^2$ given by:

$$J^{-1}(z) = \left( z \cdot \frac{k(k + 1)}{2}, \frac{k(k + 3)}{2} \cdot z \right),$$

where

$$k = \left\lfloor \frac{\sqrt{8z + 1} - 1}{2} \right\rfloor$$

for all $z \in \mathbb{N}$.

Let $K, L : \mathbb{N} \rightarrow \mathbb{N}$ be the functions given by

$$K(z) = z \cdot \frac{k(k + 1)}{2} \quad \text{and} \quad L(z) = \frac{k(k + 3)}{2} \cdot z,$$

for all $z \in \mathbb{N}$. 

The functions $J$, $K$, and $L$ satisfy:

$$J^{-1}(z) \equiv (K(z), L(z)),$$

for all $z \in \mathbb{N}$.

**Lemma 2** The functions $K$ and $L$ are primitive recursive.

**Proof** $K$ and $L$ are compositions of primitive recursive functions. \(\square\)

The functions $K$ and $L$ are called the inverses of $J$. 
2.1.2.3. Primitive Recursive Enumerations

Cantor’s enumeration can be generalized as follows:

- $J^{(2)} = J$;
- $J^{(n)}(x_1, \ldots, x_n) = J^{(2)}(J^{(n-1)}(x_1, \ldots, x_{n-1}), x_n)$, for all $n > 2$ and $x_1, \ldots, x_n \in \mathbb{N}$.

If we denote by $I^{(n)}_1, \ldots, I^{(n)}_n$ the inverses of $J^{(n)}$, then:

- $I^{(2)}_1 = K$ and $I^{(2)}_2 = L$;
- $I^{(n)}_1 = I^{(n-1)}_1 \circ I^{(2)}_1$,
  \ldots
- $I^{(n)}_{n-1} = I^{(n-1)}_{n-1} \circ I^{(2)}_1$,
  $I^{(n)}_n = I^{(2)}_1$, for all $n > 2$. 

2.1.2.3. Primitive Recursive Enumerations

The functions $J^{(n)}$, $I_1^{(n)}$, ..., $I_n^{(n)}$ satisfy:

$$(J^{(n)})^{-1}(z) = (I_1^{(n)}(z), ..., I_n^{(n)}(z)),$$

for all $z \in \mathbb{N}$.

**Lemma 3** The functions $J^{(n)}$, $I_1^{(n)}$, ..., $I_n^{(n)}$ are primitive recursive.

**Proof** $J^{(n)}$, $I_1^{(n)}$, ..., $I_n^{(n)}$ are compositions of primitive recursive functions. $\square$
2.1.2.3. Primitive Recursive Enumerations

Application: closure under simultaneous recursion

Two functions \( f_1, f_2 \in AF^{(n+1)} \) are obtained from \( g_1, g_2 \in AF^{(n)} \) and \( h_1, h_2 \in AF^{(n+3)} \) by simultaneous recursion if the following hold true:

- \( f_1(x_1, \ldots, x_n, 0) = g_1(x_1, \ldots, x_n) \),
- \( f_2(x_1, \ldots, x_n, 0) = g_2(x_1, \ldots, x_n) \),
- \( f_1(x_1, \ldots, x_n, y + 1) = h_1(x_1, \ldots, x_n, y, f_1(x_1, \ldots, x_n, y), f_2(x_1, \ldots, x_n, y)) \),
- \( f_2(x_1, \ldots, x_n, y + 1) = h_2(x_1, \ldots, x_n, y, f_1(x_1, \ldots, x_n, y), f_2(x_1, \ldots, x_n, y)) \),

for all \( x_1, \ldots, x_n, y \in \mathbb{N} \).
Proposition 17 \( PRF \) is closed under simultaneous recursion.

Proof  Hint: Use Cantor’s enumeration

- assume that \( f_1 \) and \( f_2 \) are functions obtained by simultaneous recursion over primitive recursive functions;
- define
\[
t(x_1, \ldots, x_n, y) = J^{(2)}(f_1(x_1, \ldots, x_n, y), f_2(x_1, \ldots, x_n, y));
\]
- show that \( t \) is primitive recursive;
- obtain \( f_1 = K(t) \) and \( f_2 = L(t) \) and conclude the lemma.
2.1.2.3. Primitive Recursive Enumerations

Gödel’s numbering

The Gödel number of a sequence of natural numbers \((x_0, \ldots, x_{k-1})\), denoted by \(\langle x_0, \ldots, x_{k-1} \rangle\), is

\[ p_{n(0)}^{x_0} \cdots p_{n(k-1)}^{x_{k-1}} \]

By convention, the Gödel number of the empty sequence, denoted by \(\langle \rangle\), is 0.

Example 3

1. \(\langle 3, 1, 2, 1 \rangle = 2^3 \cdot 3^1 \cdot 5^2 \cdot 7^1 = 4200;\)
2. \(\langle 3, 1, 2, 1, 0 \rangle = \langle 3, 1, 2, 1, 0, 0 \rangle = 4200.\)
Remark:

1. Gödel numbers get quite large even when the terms of the sequences are small. For instance:

   \[\langle 1, 1, 1, 1, 1, 1, 1, 1, 1 \rangle = 223,092,870\]

2. We do not have to compute Gödel numbers!

3. Distinct sequences might have the same Gödel number (for instance, (1) and (1,0));

4. Every natural number is the Gödel number of some sequence (this is a consequence of the fundamental theorem of arithmetic). Therefore, the Gödel numbering induces a surjective function from the set of all sequences into the set \( \mathbb{N} \).
2.1.2.3. Primitive Recursive Enumerations

The following two primitive recursive functions are important in order to deal with Gödel numbers:

**Example 4** The following two functions are primitive recursive:

1. $exp : \mathbb{N}^2 \rightarrow \mathbb{N}$ given by
   
   $$exp(i, z) = \begin{cases} 
   \text{the greatest } x \text{ s.t. } pn(i)^x | z, & \text{if } z > 0 \\
   0, & \text{otherwise}
   \end{cases}$$

2. $long : \mathbb{N} \rightarrow \mathbb{N}$ given by
   
   $$long(z) = \begin{cases} 
   \text{the greatest } i \text{ s.t. } exp(i, z) > 0, & \text{if } z > 1 \\
   0, & \text{otherwise}
   \end{cases}$$
Example 5

1. \( \text{exp}(i, 0) = 0; \)
2. \( \text{exp}(i, 1) = 0 \) because \( \text{pn}(i)^0 = 1 \mid 1 \) and \( \text{pn}(i)^1 \nmid 1; \)
3. \( \text{exp}(0, 15) = 0 \) because \( \text{pn}(0) = 2 \nmid 15; \)
4. \( \text{long}(0) = 0 = \text{long}(1); \)
5. \( \text{long}(2) = 0 \) because \( 2 = \text{pn}(0)^1; \)
6. \( \text{long}(15) = 2 \) because \( 15 = \text{pn}(0)^0 \cdot \text{pn}(1)^1 \cdot \text{pn}(2)^1; \)
7. if \( z > 1 \) then

\[
z = \langle \text{exp}(0, z), \ldots, \text{exp}(\text{long}(z), z) \rangle
\]
2.1.2.3. Primitive Recursive Enumerations

Application: closure under hereditary recursion

A function $f \in AF^{(n+1)}$ is obtained from $g \in AF^{(n)}$ and
$h \in AF^{(n+2)}$ by hereditary recursion if the following hold true:

- $f(x_1, \ldots, x_n, 0) = g(x_1, \ldots, x_n)$,
- $f(x_1, \ldots, x_n, y + 1) = h(x_1, \ldots, x_n, y, \langle f(x_1, \ldots, x_n, 0), \ldots, f(x_1, \ldots, x_n, y) \rangle)$,

for all $x_1, \ldots, x_n, y \in \mathbb{N}$. 
Proposition 18 \(PRF\) is closed under hereditary recursion.

Proof\quad\text{Hint: Use simultaneous recursion}

- assume that \(f\) is obtained by hereditary recursion over primitive recursive functions;
- define
  \[ t(x_1, \ldots, x_n, y) = \langle f(x_1, \ldots, x_n, 0), \ldots, f(x_1, \ldots, x_n, y) \rangle; \]
- show that \(t\) and \(f\) can be defined by simultaneous recursion, and conclude the proposition.

\[ \square \]
The Ackermann-Peter function is a simple example of a total function which is computable but not primitive recursive, providing a counterexample to the belief in the early 1900s that every computable function was also primitive recursive.

**History** – In 1928, Wilhelm Ackermann, a mathematician studying the foundations of computation, originally considered the following sequence of operations:
2.1.2.4. The Ackermann-Peter Function

\[
S(x) = x + 1 \\
\begin{cases} 
+ (x, 0) = x \\
+ (x, y + 1) = S(+(x, y)) 
\end{cases} \\
\begin{cases} 
\cdot (x, 0) = 0 \\
\cdot (x, y + 1) = +(x, \cdot (x, y)) 
\end{cases} \\
\begin{cases} 
\overrightarrow{0} = 1 \\
\overrightarrow{(y + 1)} = \cdot (x, \overrightarrow{y}) 
\end{cases} \\
\begin{cases} 
\varphi_4(x, 0) = 1 \\
\varphi_4(x, y + 1) = \varphi_3(x, \varphi_4(x, y)) 
\end{cases}
\]
2.1.2.4. The Ackermann-Peter Function

In general:

\[
\begin{align*}
\varphi_{i+1}(x, 0) & = 1 \\
\varphi_{i+1}(x, y + 1) & = \varphi_i(x, \varphi_{i+1}(x, y))
\end{align*}
\]

for all \( i \geq 3 \) (\( \varphi_0 = S \), \( \varphi_1 = + \), \( \varphi_2 = \cdot \), and \( \varphi_3 = \hat{\cdot} \)).

A few values of \( \varphi_4 \) are in order:

- \( \varphi_4(x, 1) = x \),
- \( \varphi_4(x, 2) = x^x \),
- \( \varphi_4(x, 3) = x^{x^x} \),
- and so on.
2.1.2.4. The Ackermann-Peter Function

In 1928, W. Ackermann considered the function

\[ f(i, x, y) = \varphi_i(x, y), \]

for all \( i, x, y \in \mathbb{N} \), and proved that:

- \( f \) is a recursive function;
- \( h(x) = f(x, x, x) \) grows faster than any primitive recursive function. Therefore, it is not primitive recursive.
2.1.2.4. The Ackermann-Peter Function

In 1935, Rosza Peter, defined a similar function of only two variables. Nowadays, it is called the Ackermann-Peter function.

The Ackerman-Peter function is the function $A : \mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$ given by:

(R1) $A(0, y) = y + 1$

(R2) $A(x + 1, 0) = A(x, 1)$

(R3) $A(x + 1, y + 1) = A(x, A(x + 1, y))$

for all $x, y \in \mathbb{N}$. 
Example of computation:

\[
\begin{align*}
A(2, 1) & \quad (R3), \ x \to 1, \ y \to 0 \\
& = A(1, A(2, 0)) \\
& = A(1, A(2, 0)) \\
A(1, A(2, 0)) & \quad (R2), \ x \to 1 \\
& = A(1, A(1, 1)) \\
& = A(1, A(1, 1)) \\
A(1, A(1, 1)) & \quad (R3), \ x \to 0, \ y \to 0 \\
& = A(1, A(0, A(1, 0))) \\
& = A(1, A(0, A(1, 0))) \\
A(1, A(0, A(1, 0))) & \quad (R2), \ x \to 0 \\
& = A(1, A(0, A(0, 1))) \\
& = A(1, A(0, A(0, 1))) \\
A(1, A(0, A(0, 1))) & \quad (R1), \ y \to 1 \\
& = A(1, A(0, 2)) \\
& = A(1, A(0, 2)) \\
A(1, A(0, 2)) & \quad (R1), \ y \to 2 \\
& = A(1, 3) \\
\end{align*}
\]
2.1.2.4. The Ackermann-Peter Function

Conclusions:

- each element of this computation is a term
- each computation step is based on applying one of the equations (R1), (R2) or (R3)
- each equation is used in **one direction only** ("from left to right")
- each equation is based on a substitution ("x→1, y→0") which matches the left hand side of the equation to some subterm of the current term
- the immediate successor of a term t is obtained by replacing a subterm of t by an instance of the right hand side of some equation
2.1.2.4. The Ackermann-Peter Function

Do the equations $R1 - R3$ define a function?

**Lemma 4** For any $x, y \in \mathbb{N}$ there exists an unique $z \in \mathbb{N}$ such that $z = A(x, y)$.

**Proof** by mathematical induction on $x$, followed by mathematical induction on $y$.  

Therefore, the equations $R1 - R3$ do define a function.
2.1.2.4. The Ackermann-Peter Function

The Ackermann-Peter function can be calculated by a simple algorithm based directly on the definition:

Function $A(x, y)$
begin
  if $x = 0$ then return $y + 1$
  else if $x > 0$ and $y = 0$ then return $A(x - 1, 1)$
  else return $A(x - 1, A(x, y - 1))$
end.

At each step either $y$ decreases, or $y$ increases and $x$ decreases. Each time that $y$ reaches zero, $x$ must decrease, so $x$ must eventually reach zero as well. Note, however, that when $x$ decreases there is no upper bound on how much $y$ can increase – and it will often increase greatly.
The Ackermann-Peter Function

<table>
<thead>
<tr>
<th>$x/y$</th>
<th>0</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>$y$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>1</td>
<td>2</td>
<td>3</td>
<td>4</td>
<td>5</td>
<td>$y + 1$</td>
</tr>
<tr>
<td>1</td>
<td>2</td>
<td>3</td>
<td>4</td>
<td>5</td>
<td>6</td>
<td>$y + 2$</td>
</tr>
<tr>
<td>2</td>
<td>3</td>
<td>7</td>
<td>9</td>
<td>11</td>
<td>2$y + 3$</td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>5</td>
<td>13</td>
<td>29</td>
<td>61</td>
<td>125</td>
<td>$2^{y+3} - 3$</td>
</tr>
<tr>
<td>4</td>
<td>13</td>
<td>$65533$</td>
<td>$2^{65533} - 3$</td>
<td>$A(3, 2^{65533} - 3)$</td>
<td>$A(3, A(4, 3))$</td>
<td>$2^{2^{2^{\cdots}} - 3}$</td>
</tr>
</tbody>
</table>

One surprising aspect of the Ackermann-Peter function is that the only arithmetic operations it ever uses are addition and subtraction of 1. Its properties come solely from the power of unlimited recursion. This also implies that its running time is at least proportional to its output, and so is also extremely huge. In actuality, for most cases the running time is far larger than the output.
2.1.2.4. The Ackermann-Peter Function

Some astonishing facts about the Ackermann-Peter function:

1. $A(4, 2)$ is greater than the number of particles in the universe raised to the power 200;
2. $A(5, 2)$ cannot be written as a decimal expansion in the physical universe;
3. Beyond row 4 and column 1, the values can no longer be feasibly written with any standard notation other than the Ackermann-Peter function itself – writing them as decimal expansions, or even as references to rows with lower $x$, is not possible;
4. Despite the inconceivably large values occurring in this early section of the table, some even larger numbers have been defined, such as Graham’s number. This number is constructed with a technique similar to applying the Ackermann-Peter function to itself recursively.
The Ackermann-Peter function enjoy many interesting and important properties.

**Proposition 19** The following properties hold true:

1. $A(x, y) > y$;
2. $A(x, y + 1) > A(x, y)$;
3. $A(x + 1, y) \geq A(x, y + 1)$;
4. $A(x, y) > x$;
5. $A(x + 1, y) > A(x, y)$;
6. $A(x + 2, y) > A(x, 2y)$,

for all $x, y \in \mathbb{N}$. 

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The following theorem is crucial in order to prove that the Ackermann-Peter function is not primitive recursive.

**Theorem 1** For any \( n \geq 1 \) and any primitive recursive function \( f \in PRF^{(n)} \), there exists \( k \in \mathbb{N} \), which depends on \( f \) and \( A \), such that

\[
f(x_1, \ldots, x_n) < A(k, ||(x_1, \ldots, x_n)||),
\]

for all \((x_1, \ldots, x_n) \in \mathbb{N}^n\), where

\[
||(x_1, \ldots, x_n)|| = \max\{x_1, \ldots, x_n\}.
\]

This theorem shows that the Ackermann-Peter function grows faster than any primitive recursive function.
2.1.2.4. The Ackermann-Peter Function

Now we are able to prove the main result regarding the Ackermann-Peter function.

**Corollary 1** The Ackermann-Peter function is not primitive recursive. Moreover, the function \( f(x) = A(x, x) \) is not primitive recursive too.

**Proof** If we assume that \( A(x, y) \) is primitive recursive, then

\[
f(x) = A(x, x)
\]

is a primitive recursive function. By Theorem 1, there exists \( k \) such that \( f(x) < A(k, x) \), for any \( x \). Therefore,

\[
A(x, x) < A(k, x)
\]

for any \( x \), which is a contradiction. \( \square \)
2.1.2.4. The Ackermann-Peter Function

Since the function \( f(x) = A(x, x) \) considered above grows very rapidly, its inverse function, \( f^{-1} \), grows very slowly. In fact, it is less than 5 for any conceivable input size \( x \), since \( A(4, 4) \) has a number of digits that cannot itself be written in binary in the physical universe. For all practical purposes, \( f^{-1}(x) \) can be regarded as being a constant.

This inverse appears in the time complexity of some algorithms, such as the disjoint-set data structure and Chazelle’s algorithm for minimum spanning trees.
A two-parameter variation of the inverse Ackermann-Peter function can be defined as follows:

$$\alpha(x, y) = \min\{i \geq 1 | A(i, \lfloor x/y \rfloor) \geq \log_2 \rho y\}$$

This function arises in more precise analysis of the algorithms mentioned above, and gives a more refined time bound. In the disjoint-set data structure, \(x\) represents the number of operations while \(y\) represents the number of elements; in the minimum spanning tree algorithm, \(x\) represents the number of edges while \(y\) represents the number of vertices.
2.1.2.4. The Ackermann-Peter Function

The Ackermann-Peter function, due to its definition in terms of extremely deep recursion, can be used as a benchmark of a compiler’s ability to optimize recursion. For example, a compiler which, in analyzing the computation of \( A(3,30) \), is able to save intermediate values like the \( A(3, y) \) and \( A(2, y) \) in that calculation rather than recomputing them, can speed up computation of \( A(3,30) \) by a factor of hundreds of thousands. Also, if \( A(2, y) \) is computed directly rather than as a recursive expansion of the form \( A(1, A(1, A(1, ... A(1,0)...)) \)), this will save significant amounts of time.

Computing \( A(1, y) \) takes linear time in \( y \). Computing \( A(2, y) \) requires quadratic time, since it expands to \( O(y) \) nested calls to \( A(1, i) \) for various \( i \). Computing \( A(3, y) \) requires time proportionate to \( 4^{y+1} \).
2.1.3. Recursive Functions

Recursive functions:

- Definitions and examples
- The Ackermann-Peter function is recursive
- Reducibility
- Recursively enumerable sets. Rice’s theorem
Let $f : \mathbb{N} \to \mathbb{N}$ be the function given by

$$f(x) = \begin{cases} \lfloor x/3 \rfloor, & \text{if } 3 \mid x \\ \uparrow, & \text{otherwise,} \end{cases}$$

for all $x \in \mathbb{N}$.

$f$ is not primitive recursive but it is effectively computable:

Function $f(x)$
- input: $x \in \mathbb{N}$;
- output: “$x/3$”, if $3 \mid x$, and “undefined”, otherwise;
- begin
  - if $3 \mid x$ then return $x/3$ else return “undefined”
- end.
2.1.3.1. Definitions and Examples

Remarks:

1. The function $f$, defined as above, is an effectively computable partial function, but it is not primitive recursive;
2. The Ackermann-Peter function is an effectively computable total function, but it is not primitive recursive.

Conclusion: The class of primitive recursive functions needs to be extended!
2.1.3.1. Definitions and Examples

The function $f$ given above can be defined by

$$f(x) = \begin{cases} 
\text{the least } y \text{ s.t. } g(x, y) = 0, & \text{if such an } y \text{ exists} \\
\uparrow, & \text{otherwise},
\end{cases}$$

for all $x \in \mathbb{N}$, where $g(x, y) = |x - 3y|$.

A function $f \in AF^{(n)}$ is obtained from $g \in AF^{(n+1)}$ by minimization if

- $f(x_1, \ldots, x_n) = \text{the least } y \text{ for which } g(x_1, \ldots, x_n, y) = 0$, if such an $y$ exists;

- $f(x_1, \ldots, x_n) = \uparrow$, otherwise,

for all $x_1, \ldots, x_n \in \mathbb{N}$. 

2.1.3.1. Definitions and Examples

If \( f \) is obtained from \( g \) by minimization then we write

\[
f(x_1, \ldots, x_n) = \mu(y)[g(x_1, \ldots, x_n, y) = 0]
\]

The class of **recursive functions** is the smallest class \( RF \) of arithmetic functions which satisfies:

- includes all the initial functions;
- it is closed under composition;
- it is closed under primitive recursion;
- it is closed under minimization.

**Remark:** \( RF \) includes \( PRF \) and contains partial functions too.
2.1.3.1. Definitions and Examples

Example 6

1. The function given by \( f(x) = \mu(y)[|x - 3y| = 0] \), for all \( x \in \mathbb{N} \), is a recursive function (but not primitive recursive);
2. The function given by

\[
\Omega^{(n)}(x_1, \ldots, x_n) = \uparrow,
\]

for all \( x_1, \ldots, x_n \in \mathbb{N} \), is a recursive function (but not primitive recursive), for all \( n \geq 1 \). It is the totally undefined function.

Function \( \Omega(x_1, \ldots, x_n) \)
- input: \( x_1, \ldots, x_n \in \mathbb{N} \);
- output: “undefined”;
begin
  return “undefined”
end.
Remark: The class $RF$ contains:

- recursive functions which are totally defined, called total recursive functions;
- recursive functions which are partially defined, called partial recursive functions.
2.1.3.2. The Ackermann-Peter Function is Recursive

Inner-most computation of the Ackermann-Peter function:

<table>
<thead>
<tr>
<th>no.</th>
<th>configuration</th>
<th>code</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>$A(x, y)$</td>
<td>$c(x, y, 0)$</td>
</tr>
<tr>
<td>...</td>
<td>$A(x_1, A(x_2, \ldots, A(x_k, y') \ldots))$</td>
<td>$c(x, y, m)$</td>
</tr>
<tr>
<td>$m$</td>
<td>$z$</td>
<td>$c(x, y, m_f)$</td>
</tr>
</tbody>
</table>

Encoding the computation steps:

- $c(x, y, 0) = 2^{23^{x+1}5^{y+1}}$
- $c(x, y, m) = pn(0)^{k+1}pn(1)^{x_1+1} \ldots pn(k)^{x_k+1}pn(k+1)^{y'+1}$
- $c(x, y, m_f) = 2^{13^{z+1}}$
- $c(x, y, k) = c(x, y, k-1)$, if $k > m_f$
Lemma 5  The function $c(x, y, m)$ is primitive recursive.

Proof  (sketch)

Define the predicate $G$ by:

$$G(\alpha) = 1 \iff \alpha \text{ is the code of some configuration}$$

$G$ is primitive recursive:

- $G(\alpha) = eq\left(\exp(0, \alpha), long(\alpha)\right) \cdot \prod_{i=0}^{long(\alpha)} gr\left(\exp(i, \alpha), 0\right)$
2.1.3.2. The Ackermann-Peter Function is Recursive

Define the predicate $R_i(\alpha)$, for $i = 1, 2, 3$, by:

$$R_i(\alpha) = 1 \iff G(\alpha) = 1 \text{ and } (R_i) \text{ can be applied to the configuration encoded by } \alpha$$

$R_i$ is primitive recursive, for all $i = 1, 2, 3$:

- $R_1(\alpha) = G(\alpha) \cdot eq(exp(long(\alpha) \div 1, \alpha), 1)$
- $R_2(\alpha) = G(\alpha) \cdot gr(exp(long(\alpha) \div 1, \alpha), 1) \cdot eq(exp(long(\alpha), \alpha), 1)$
- $R_3(\alpha) = G(\alpha) \cdot gr(exp(long(\alpha) \div 1, \alpha), 1) \cdot gr(exp(long(\alpha), \alpha), 1)$

Define the predicate $R_4$ by:

- $R_4(\alpha) = G(\alpha) \cdot \neg (R_1(\alpha) + R_2(\alpha) + R_3(\alpha))$

$R_4$ is primitive recursive
2.1.3.2. The Ackermann-Peter Function is Recursive

Define the function \( h_i \), \( i = 1, 2, 3, 4 \), by

\[
h_i(\alpha) = \text{the new code obtained by applying } R_i \text{ to } \alpha
\]

The function \( h_i \) is primitive recursive, for all \( i = 1, 2, 3, 4 \):

- \[
h_1(\alpha) = R_1(\alpha) \cdot \left(\frac{\alpha}{pn(0)} \cdot pn\left(long(\alpha) \div 1\right) \cdot \frac{\alpha}{pn\left(long(\alpha)\right)^{exp\left(long(\alpha),\alpha\right)}}\right) \cdot pn\left(long(\alpha) \div 1\right)^{exp\left(long(\alpha),\alpha\right)+1}
\]

- \[
h_2(\alpha) = R_2(\alpha) \cdot \left(\frac{\alpha}{pn\left(long(\alpha) \div 1\right)}\right) \cdot pn\left(long(\alpha)\right)
\]

- \[
h_3(\alpha) = R_3(\alpha) \cdot \left(\frac{\alpha}{pn\left(long(\alpha) \div 1\right)} \cdot pn\left(long(\alpha)\right)^{exp\left(long(\alpha),\alpha\right)}}\right) \cdot pn\left(long(\alpha)\right)^{exp\left(long(\alpha),\alpha\right)+1} \cdot pn\left(long(\alpha) + 1\right)^{exp\left(long(\alpha),\alpha\right) \div 1} \cdot pn(0)
\]

- \[
h_4(\alpha) = \alpha
\]
The function $c$ can be written as:

1. $c(x, y, 0) = 2^{2x+1}3^{y+1}$;
2. $c(x, y, m + 1) = \sum_{i=1}^{4} R_i(c(x, y, m)) \cdot h_i(c(x, y, m))$, for all $x, y, m \in \mathbb{N}$.

Therefore, the function $c$ is primitive recursive, and the proof is completed. \qed
2.1.3.2. The Ackermann-Peter Function is Recursive

**Lemma 6** The function \( t : \mathbb{N}^2 \rightarrow \mathbb{N} \) given by:

\[
t(x, y) = \text{the number of the final step in the inner-most computation of } A(x, y),
\]

for all \( x, y \in \mathbb{N} \), is a total recursive function.

**Proof**

\[
t(x, y) = \mu(m)[\exp(0, c(x, y, m)) = 1]
\]
Theorem 2  The Ackermann-Peter function is recursive.

Proof

\[ A(x, y) = \exp(1, c(x, y, t(x, y))) - 1 \]

Remark: The function \( t(x, y) \) is not primitive recursive (otherwise, \( A \) would be primitive recursive).
2.1.3.2. The Ackermann-Peter Function is Recursive

Notation:

- \( AF \) – arithmetic functions;
- \( TAF \) – total arithmetic functions;
- \( PRF \) – primitive recursive functions;
- \( RF \) – recursive functions;
- \( TRF \) – total recursive functions.

Remark:

- \( AF \) is uncountable: \(|AF| = 2^{\aleph_0}\)
- \( PRF, TRF, \) and \( RF \) is countable:

\[
|PRF| = |TRF| = |RF| = \aleph_0
\]
2.1.3.2. The Ackermann-Peter Function is Recursive

**Theorem 3** The following relationships between these classes of functions hold:

![Diagram showing relationships between classes of functions]

- PRF → TRF
- RF
- AF
- TAF
- AF
- RF
- TRF
- PRF
- TAF
2.1.3.2. The Ackermann-Peter Function is Recursive

Proof  The following inclusions hold:

- $PRF \subset TRF$: the Ackermann-Peter function is a total recursive function but not primitive recursive;
- $TRF \subset RF$: the totally undefined function is recursive but it is not a total function;
- $RF \subset AF$: $|RF| < |AF|$;
- $RF$ and $TAF$ are incomparable: $RF$ contains partial functions, and $|TAF| > |RF|$.

□
Church’s Thesis: formulated by the American logician Alonzo Church in 1935, states that the recursive functions are the only functions that can be mechanically calculated.

Originally, Alonzo Church (1932, 1936, 1941) and Stephen Kleene (1935) introduced lambda-definable functions, and Kurt Gödel (1934) and Jacques Herbrand (1932) introduced recursive functions. These two formalisms describe the same set of functions, as was shown in the case of functions of positive integers by Church (1936) and Kleene (1936).
2.1.3.3. Limits of Algorithmic Computability

Terminology:

- **decision problem**
  - solvable (computable)
    - * decidable = total recursive function
    - * semi-decidable = (partial) recursive function
  - unsolvable (undecidable) = non-recursive function

- **computation problem**
  - solvable (computable) = recursive function
  - unsolvable (non-computable) = non-recursive function

How does a problem (function) that is not computable (recursive) look like?
The class of recursive functions is countable. Therefore, we may enumerate all the \( n \)-ary recursive functions:

\[
f_0^{(n)}, f_1^{(n)}, \ldots
\]

(the arity will be omitted when it is clear from the context).

**Assumption (Wagner-Strong axiom):** There exists a recursive function \( u^{(n+1)} \) such that

\[
u^{(n+1)}(i, x_1, \ldots, x_n) = f_i^{(n)}(x_1, \ldots, x_n),
\]

for all \( i, x_1, \ldots, x_n \in \mathbb{N} \). \( u \) is called an *universal function* (later we will prove that such a function exists but, for the time being we accept it as an axiom).

An universal function compresses an infinite sequence of functions (as the (original) Ackermann function does).
Denote by $d : \mathbb{N}^4 \rightarrow \mathbb{N}$ the function given by:

$$d(x, y, u, v) = \begin{cases} u & \text{if } x = y \\ v & \text{else} \end{cases}$$

for all $x, y, u, v \in \mathbb{N}$.

$d$ is primitive recursive (prove it!)

Let $\text{diag} : \mathbb{N} \rightarrow \mathbb{N}$ be the predicate given by:

$$\text{diag}(x) = \begin{cases} 1, & f_x(x) \downarrow \\ 0, & \text{otherwise} \end{cases}$$

for all $x \in \mathbb{N}$.

This predicate describes, in a functional way, a particular halting test.
Theorem 4 \( \text{diag} \) is not recursive.

Proof (sketch)

- assume, by contradiction, that \( \text{diag} \) is recursive;
- define \( g(x) = u(d(diag(x), 0, i, j), x) \), for some \( i \) and \( j \);
- \( g \) is recursive and, therefore, there exists \( k \) such that \( g = f_k \);
- derive a contradiction.

\( \square \)
2.1.3.3. Limits of Algorithmic Computability

Once some function $f$ has been shown to be non-recursive, we can use that function to give other examples of non-recursive functions by way of the reducibility method.

A function $f$ is recursively reducible (or r-reducible) to $g$ if there exists a recursive function $h$ such that $g \circ h = f$. The function $h$ is called a reducibility function.

If $f$ is recursively reducible to $g$ by means of $h$ then we will write $f \prec_h g$ or even $f \prec g$.

**Remark:** If $f \prec_h g$ and $f$ is not recursive, then $g$ is not recursive.
2.1.3.3. Limits of Algorithmic Computability

Let $\text{halt}^{(n+1)} : N^{(n+1)} \rightarrow N$ be the predicate given by:

$$\text{halt}^{(n+1)}(i, x_1, \ldots, x_n) = \begin{cases} 
1, & f_i(x_1, \ldots, x_n) \downarrow \\
0, & \text{otherwise},
\end{cases}$$

for all $i, x_1, \ldots, x_n \in N$.

**Corollary 2** $\text{halt}^{(2)}$ is not recursive.

**Proof** Hint: use reducibility. \qed
Assumption (Wagner-Strong axiom): For any $m \geq 0$ and $n > 0$ there exists a total recursive function $s_{m,n}^{(m+1)}$ such that

$$f_i^{(m+n)}(x_1, \ldots, x_m, y_1, \ldots, y_n) = f_{s_{m,n}^{(m+1)}}^{(n)}(i, x_1, \ldots, x_m)(y_1, \ldots, y_n),$$

for all $i, x_1, \ldots, x_m, y_1, \ldots, y_n \in \mathbb{N}$. $s_{m,n}$ is called an s-m-n function (later we will prove that such a function exists but, for the time being we accept it as an axiom).

An s-m-n function reduces by $m$ the number of arguments of $(m + n)$-ary functions.
2.1.3.3. Limits of Algorithmic Computability

Let \( t : \mathbb{N} \rightarrow \mathbb{N} \) be the predicate given by:

\[
t(x) = \begin{cases} 
1, & \text{if } f_x \text{ is total} \\
0, & \text{otherwise},
\end{cases}
\]

for all \( x \in \mathbb{N} \).

**Corollary 3** \( t \) is not recursive.

**Proof** Hint: use reducibility and the s-m-n property. \( \square \)

This predicate describes, in a functional way, a **totality test**.
An arithmetic predicate \( f : \mathbb{N}^n \rightarrow \mathbb{N} \) can be regarded as a subset of \( \mathbb{N} \):

\[
A_f = \{ x \in \mathbb{N}^n | f(x) = 1 \}
\]

The characteristic function of a subset \( A \subseteq \mathbb{N}^n \) is the function \( \chi_A : \mathbb{N}^n \rightarrow \mathbb{N} \) given by \( \chi_A(x) = 1 \) iff \( x \in A \), for all \( x \in \mathbb{N}^n \).

Given a predicate \( f \), \( f = \chi_{A_f} \).
A set $A \subseteq \mathbb{N}^n$ is called **primitiv recursive** if its characteristic function $\chi_A$ is a **primitive recursive** function.

A set $A \subseteq \mathbb{N}^n$ is called **recursive** if its characteristic function $\chi_A$ is a **total recursive** function.

**Proposition 20**

(1) $\emptyset$ and $\mathbb{N}$ are primitive recursive sets.

(2) If $A, B \subseteq \mathbb{N}^n$ are (primitive) recursive sets, then $A \cup B$, $A \cap B$, and $A - B$ are (primitive) recursive sets too.
Example 7  Not every set $A \subseteq \mathbb{N}^n$ is recursive.

1. The set $A = \{x \in \mathbb{N} \mid f_x(x) \downarrow \}$ is not recursive because its characteristic set is $\chi_A = \text{diag}$;
2. The set $B = \{x \in \mathbb{N} \mid f_x \text{ total} \}$ is not recursive because its characteristic set is $\chi_B = t$.

The sets in the example above are referring to the class of all unary recursive functions. There is a general result due to Rice (1953).
Theorem 5 (Rice’s Theorem)
Let $A \subseteq \{f_i^{(1)}\mid i \geq 0\}$ and $\text{Ind}(A) = \{i \mid f_i \in A\}$. Then, $\text{Ind}(A)$ is recursive iff $\text{Ind}(A) = \emptyset$ or $\text{Ind}(A) = \mathbb{N}$.

Proof $\Rightarrow$: (sketch)
- Assume that $\text{Ind}(A)$ is recursive but $\text{Ind}(A) \neq \emptyset$ and $\text{Ind}(A) \neq \mathbb{N}$;
- reduce $\text{diag}$ to $\chi_{\text{Ind}(A)}$ and get a contradiction.

Proof $\Leftarrow$: straightforward.  

Rice’s theorem states that, no non-trivial property of unary recursive functions is decidable.
Corollary 4 The following sets are not recursive:

1. \( \{ i \in \mathbb{N} | f_i \text{ total} \} \);
2. \( \{ i \in \mathbb{N} | \text{dom}(f_i) = \emptyset \} \);
3. \( \{ i \in \mathbb{N} | \text{dom}(f_i) \text{ infinite} \} \);
4. \( \{ i \in \mathbb{N} | f_i(x_0) = y_0 \} \), where \( x_0 \) and \( y_0 \) are given;
5. \( \{ i \in \mathbb{N} | f_i = f_m \} \), where \( m \) is given;
6. \( \{ i \in \mathbb{N} | y_0 \in \text{cod}(f_i) \} \), where \( y_0 \) is given.

Proof Use Rice’s Theorem. \( \square \)
2.2. Turing Machines

2.2.1. Turing machines

2.2.2. Techniques for Turing machine construction

2.2.3. Recursive functions are Turing-computability

2.2.4. Turing-computable functions are recursive

2.2.5. The Church-Turing Thesis
2.2.1. Turing Machines

A formal model for an effective procedure should possess certain properties:

- it should be finitely describable;
- the procedure should consist of discrete steps, each of which can be carried out mechanically.

In 1936, Alan Turing proposed such a model. His model, called today Turing machine, has become the accepted formalization of an effective procedure.
2.2.1. Turing Machines

The basic model has a **finite control**, an **input tape** that is divided into cells, and a **tape head** that scans one cell of the tape at a time.

```
a b a c B B ...
```

The tape has a **leftmost cell** but it is infinite to the right. Each cell may hold **exactly one** of a finite number of tape symbols. Initially, the $n$ leftmost cells, for some $n \geq 0$, hold the input; the remaining infinity of cells each hold the **blank**, which is a special tape symbol that is not an input symbol. The finite control can be in one of a finite number of **internal states**.
2.2.1. Turing Machines

In one move, the machine, depending upon the symbol scanned by the tape head and the state of the finite control, does the followings:

- changes state,
- replaces the current symbol on the tape cell by a new symbol, and
- moves its tape head one cell to the left or to the right.
2.2.1. Turing Machines

A deterministic Turing machine, abbreviated $DTM$, is a 7-tuple $M = (Q, \Sigma, \Gamma, \delta, q_0, B, F)$, where

- $Q$ is a finite non-empty set of states;
- $\Sigma$ is a finite non-empty set of input symbols, called the input alphabet of $M$;
- $\Gamma$ is a finite non-empty set of tape symbols, called the tape alphabet of $M$. It is assumed that $\Sigma \subseteq \Gamma$;
- $\delta : Q \times \Gamma \sim Q \times \Gamma \times \{L, R\}$ is the transition relation/function of $M$;
- $q_0 \in Q$ is the initial/start state;
- $B \in \Gamma - \Sigma$ is the blank symbol;
- $F \subseteq Q$ is the set of final states.
2.2.1. Turing Machines

Let $M = (Q, \Sigma, \Gamma, \delta, q_0, B, F)$ be a DTM.

- A configuration (instantaneous description) of $M$ is any pair $(q, u|av) \in Q \times \Gamma^*\{|\}\Gamma^+$. Its meaning is:
  - $q$ denotes the current state;
  - $u$ consists of the symbols from the first cell up to, but not including, the symbol under the tape head;
  - the vertical bar “|” denotes the position of the tape head;
  - $a$ is the symbol under the tape head;
  - $v$ is either the empty string, if all symbols from the tape head, including $a$, are $B$, or the string of tape symbols for which all the other symbols on the rest of the tape are blank;
2.2.1. Turing Machines

- A configuration \((q, u|av)\) for which \(q = q_0\), \(u = \lambda\), and 
\(v = \lambda\) whenever \(a = B\), is called an **initial configuration**;
- \(\text{Config}(M)\) stands for the set of all configurations of \(M\);
- \(C_0(w)\) stands for \((q_0, |w)\), if \(w \neq \lambda\), and \((q_0, |B)\), otherwise.

**Example 8** The following pairs are configurations of some Turing machine:

- \((q, aaBa|aBBBBaab)\)
- \((q, |BaBBab)\)
- \((q, |BBaBbab)\)
2.2.1. Turing Machines

Let $M$ be a DTM. Define the transition relation of $M$,

$$\vdash_M \subseteq \text{Config}(M) \times \text{Config}(M),$$

as follows:

$$(q, u|av) \vdash_M (q', u'|a'v')$$

iff one of the following two properties holds:

- (move/step to the right)
  $$\delta(q, a) = (q', \bar{a}, R), u' = u\bar{a},$$
  and
  
  if $v = \lambda$ then $a' = B$ and $v' = \lambda$ else $v = a'v'$;

- (move/step to the left)
  $$\delta(q, a) = (q', \bar{a}, L), u \neq \lambda, u = u'a', v' = \bar{a}v.$$
Let $M$ be a $DTM$. 

- (computation step/move/transition)

$$(q, u|av) \vdash_M (q', u'|a'v')$$

- (computation)

$$(q_0, u_0|a_0v_0) \vdash_M \cdots \vdash_M (q_n, u_n|a_nv_n) ,$$

where $(q_0, u_0|a_0v_0)$ is an initial configuration.

- a configuration $C$ is called a blocking configuration if there is no configuration $C'$ such that $C \vdash_M C'$;

- a configuration $C'$ is called reachable from $C$ if $C \vdash_M^* C'$;
2.2.1. Turing Machines

**Remark:** Let $M$ be a $DTM$. Then:

1. In any configuration, the machine can perform at most one move (because the machine is deterministic);
2. Blocking configurations are not necessarily final configurations, and vice-versa;
3. A configuration $(q, u|av)$ is a blocking configuration if either $\delta(q, a)^\uparrow$ or $u = \lambda$ and $\delta(q, a) = (q', \bar{a}, L)$. 
2.2.1. Turing Machines

Let $M$ be a DTM and $w$ and input for $M$.

- $M$ halts on $w$ if there exists a blocking or final configuration $C$ such that $(q_0, |w|)^* \vdash C$. Two cases are to be considered:
  - $M$ accepts $w$ if $C$ is a final configuration. Denote by $\text{accept}(M)$ the set of all $w$ accepted by $M$;
  - $M$ rejects $w$ if $C$ is a blocking configuration. Denote by $\text{reject}(M)$ the set of all $w$ rejected by $M$;
- $M$ loops on $w$ if it does not halt on $w$.

$$\text{accept}(M) \cup \text{reject}(M) \cup \text{loop}(M) = \Sigma^*$$

$\text{accept}(M)$ is also called the language accepted by $M$, denoted $L(M)$. 

2.2.1. Turing Machines

Example 9 The following Turing machine accepts

\[ L = \{0^n1^n|n \geq 1\}. \]

<table>
<thead>
<tr>
<th></th>
<th>0</th>
<th>1</th>
<th>X</th>
<th>Y</th>
<th>B</th>
</tr>
</thead>
<tbody>
<tr>
<td>(q_0)</td>
<td>((q_1, X, R))</td>
<td>((q_3, Y, R))</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(q_1)</td>
<td>((q_1, 0, R))</td>
<td>((q_2, Y, L))</td>
<td>((q_1, Y, R))</td>
<td></td>
<td></td>
</tr>
<tr>
<td>(q_2)</td>
<td>((q_2, 0, L))</td>
<td>((q_0, X, R))</td>
<td>((q_2, Y, L))</td>
<td></td>
<td></td>
</tr>
<tr>
<td>(q_3)</td>
<td></td>
<td></td>
<td></td>
<td>((q_3, Y, R))</td>
<td>((q_4, B, R))</td>
</tr>
<tr>
<td>(q_4)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

\(q_4\) is the only final state.
2.2.1. Turing Machines

Turing machines may be viewed as computers of arithmetic functions. The traditional approach is to represent integers in unary. The integer $x \geq 0$ is represented by $0^x$.

$$(x_1, \ldots, x_k) \sim 0^{x_1}1 \cdots 10^{x_k}$$

Example 10

- $(1, 2, 1) \sim 010010$
- $(2, 0, 0, 3) \sim 00111000$

Remark:

- Three tape symbols are needed: 0, 1, and $B$;
- 0 and 1 may be replaced by any two distinct symbols $a$ and $b$. 
2.2.1. Turing Machines

Let $M$ be a $DTM$ and $f : \mathbb{N}^k \rightarrow \mathbb{N}$. $M$ computes $f$ if, for any $x_1, \ldots, x_k \in \mathbb{N}$, one of the following two properties holds true:

- if $f(x_1, \ldots, x_k) \downarrow$ then $M$ halts on $(q_0, |0^{x_1}1 \cdots 10^{x_k})$ with a tape consisting of $0^f(x_1, \ldots, x_k)$;
- if $f(x_1, \ldots, x_k) \uparrow$ then $M$ does not halt on $(q_0, |0^{x_1}1 \cdots 10^{x_k})$.

An arithmetic function is Turing-computable if there exists a Turing machine which computes it.

We denote by $TCF$ the set of all Turing-computable functions.
Remark:

- All natural numbers are Turing-computable;
- If the unary notation for natural numbers is adopted as being

\[ x \mapsto 0^x + 1 \]

then Turing machines with only two tape symbols are sufficient to compute arithmetic functions (prove it!).
### Example 11

The following Turing machine computes $x \div y$.

<table>
<thead>
<tr>
<th></th>
<th>0</th>
<th>1</th>
<th>$B$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$q_0$</td>
<td>$(q_1, B, R)$</td>
<td>$(q_5, B, R)$</td>
<td></td>
</tr>
<tr>
<td>$q_1$</td>
<td>$(q_1, 0, R)$</td>
<td>$(q_2, 1, R)$</td>
<td></td>
</tr>
<tr>
<td>$q_2$</td>
<td>$(q_3, 1, L)$</td>
<td>$(q_2, 1, R)$</td>
<td>$(q_4, B, L)$</td>
</tr>
<tr>
<td>$q_3$</td>
<td>$(q_3, 0, L)$</td>
<td>$(q_3, 1, L)$</td>
<td>$(q_0, B, R)$</td>
</tr>
<tr>
<td>$q_4$</td>
<td>$(q_4, 0, L)$</td>
<td>$(q_4, B, L)$</td>
<td>$(q_6, 0, R)$</td>
</tr>
<tr>
<td>$q_5$</td>
<td>$(q_5, B, R)$</td>
<td>$(q_5, B, R)$</td>
<td>$(q_6, B, R)$</td>
</tr>
<tr>
<td>$q_6$</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

The machine has no final state.
**2.2.2. Techniques for Turing Machine Construction**

**Storage in the finite control**

The finite control can be used to hold a finite amount of information. To do so, the state is written as a pair of elements, one exercising control and the other one storing a symbol.

**Example 12** Design a Turing machine that looks at the first input symbol, records it in its finite control, accepts if the first symbol does not appear elsewhere on its input, and rejects, otherwise.
Shifting over

A Turing machine can make space on its tape by shifting blocks of symbols a finite number $k$ of cells to the right. To do so,

- the machine uses its finite control to store $k$ symbols;
- initially, the state is $(q, B, \ldots, B)$;
- if the current state is $(q, X_1, \ldots, X_k)$ and the symbol scanned on the tape is $X$, the machine prints $X_1$, and enters the state $(q, X_2, \ldots, X_k, X)$. This process is repeated until the block in question is exhausted.

If space is available, it can push blocks of symbols left in a similar manner.
Remark:

- If $k$ is large, the number of states could became very large. A different solution will be proposed after the “checking off symbols” technique is discussed;
- by this technique, blocks of symbols can be surrounded by two special symbols, $\$ \text{ and } \#$, for instance. The block in question is shifted one cell to the right by printing $\$ \text{ first and } \# \text{ at the end.}
2.2.2. Techniques for Turing Machine Construction

Multiple tracks

We can view the tape of the Turing machine as being divided into $k$ tracks, for any $k \geq 2$. To do so, the tape symbols are considered $k$-tuples, one component for each track.

**Example 13** Design a Turing machine that takes a binary input greater than 2 and determines whether it is a prime.
Checking off symbols

Turing machines can check off symbols on their tapes. To do so, the tape is divided into 2 tracks, one holding the input and the other one being used to check symbols.

**Example 14** Design a Turing machine that accepts the language

\[ L = \{wcw|w \in \{a, b\}^*\}, \]

where \( c \notin \{a, b\} \).
2.2.2. Techniques for Turing Machine Construction

Using this technique, a Turing machine can shift blocks of symbols a finite number $k$ of cells to the right, as follows:

- check (if it is necessarily) the first symbol of the block;
- repeatedly do
  - store and erase the rightmost symbol of the block;
  - move right $k$ cells and print the symbol;
  
until the checked symbol is encounter (and shift it too).

If $k$ is large, this solution might be more convenient than the previous one.
2.2.2. Techniques for Turing Machine Construction

Subroutines

As with programs, Turing machine can be designed in a “to-down” manner, by using “subroutines”. The general idea is to write part of a Turing machine program to serve as a subroutine, with an initial state and several return states. The call of this subroutine is effected by entering in its initial state, and the return is effected by the move from a return state.

Example 15 Design a Turing machine which computes the multiplication function.
Lemma 7 Initial functions are Turing-computable.

Proof

- \( S(x) = x + 1 \)
  start off with \( 0^x \) on the tape, skip over 0's until the leftmost \( B \) is encountered, change \( B \) to 0, and halt;

- \( C_m^{(n)}(x_1, \ldots, x_n) = m \)
  start off with \( 0^{x_1}1 \cdots 10^{x_n} \) on the tape, change all 0s and 1s to \( B \)'s and, on encountering \( B \), print \( m \) 0s;

- \( P_i^{(n)}(x_1, \ldots, x_i, \ldots, x_n) = x_i \)
  start off with \( 0^{x_1}1 \cdots 10^{x_i}1 \cdots 10^{x_n} \) on the tape, change all 0s and 1s to \( B \)'s until the \( (i - 1) \)st 1 is encountered, skip over 0's until the leftmost 1 or \( B \) is encountered, and erase the rest of the tape.

\[\square\]
Lemma 8 Function composition of Turing-computable functions is Turing-computable.

Proof Let $g_1, \ldots, g_m \in AF^{(n)}$ and $f \in AF^{(m)}$ be Turing-computable functions. Assume that $M_i$ computes $g_i$, $1 \leq i \leq m$, and $M_0$ computes $f$. Define $M$ as follows:

- start off with $0^{x_1}1 \cdots 10^{x_n}$;
- surround the input by $\$ and #,

$$0^{x_1}1 \cdots 10^{x_n}\#$$

- copy the input to the right of the first #,

$$0^{x_1}1 \cdots 10^{x_n}\#0^{x_1}1 \cdots 10^{x_n}$$
2.2.3. Recursive Functions are Turing-computable

- call $M_1$ on the second input's copy
  $$0^{x_1} \cdots 10^{x_n} \# 0^{g_1(x_1, \ldots, x_n)}$$

- repeat the cycle and call $M_2, \ldots, M_m$
  $$0^{x_1} \cdots 10^{x_n} \# 0^{g_1(x_1, \ldots, x_n)} \# \cdots \# 0^{g_m(x_1, \ldots, x_n)}$$

- change all #’s, except for the first one, to 1’s
  $$0^{x_1} \cdots 10^{x_n} \# 0^{g_1(x_1, \ldots, x_n)} 1 \cdots 10^{g_m(x_1, \ldots, x_n)}$$

- call $M_0$
  $$0^{x_1} \cdots 10^{x_n} \# 0^{f(g_1(x_1, \ldots, x_n), \ldots, g_m(x_1, \ldots, x_n))}$$

- erase the block $0^{x_1} \cdots 10^{x_n} \#$ and halt.
Lemma 9 If \( f \) is obtained by primitive recursion over \( g \) and \( h \), and \( g \) and \( h \) are total Turing-computable functions, then \( f \) is Turing-computable.

Proof Assume that \( M_1 \) computes \( g \) and \( M_2 \) computes \( h \). For the sake of simplicity we will consider that \( f \) is a 2-ary function. Define \( M \) as follows:

- start off with \( 0^x10^y \)
- generates

\[
0^x10^y0^x10^{y-1} \cdots 0^x100^x
\]
- call \( M_1 \) on the last block and compute

\[
f(x,0) = g(x)
\]
2.2.3. Recursive Functions are Turing-computable

- replace the last block 0$ by 1

\[ 0^x 10^y 0^x 10^{y-1} \ldots 0^x 110^f(x,0) \]

- call \( M_2 \) on the last block and compute \( f(x, 1) = h(x, 0, f(x, 0)) \)

\[ 0^x 10^y 0^x 10^{y-1} \ldots 0^f(x,1) \]

- repeat the cycle \( y - 1 \) times and get

\[ 0^f(x,y) \]

- erase the symbol $ and halt.

\( \Box \)
Lemma 10 If $f$ is obtained by minimization over a total Turing-computable function $g$, then $f$ is Turing-computable.

Proof For the sake of simplicity we assume that $f$ is a unary function. Let $M$ be a Turing machine which computes $g$. Define $M'$ as follows:

1. start off with $0^x$
2. generate $0^x1$
3. copy the first block after the second $\$, $0^x10^x1$, and call $M$ on the second block in order to compute $g(x, 0)$
4. if $g(x, 0) = 0$ then halt; otherwise, increment the second argument, $0^x10$, and go to step 3.

$\square$
Theorem 6 Recursive functions are Turing-computable.

Proof The class of Turing-computable functions includes the initial functions and it is closed under function composition, primitive recursion, and minimization.
2.2.4. Turing-computable Functions are Recursive

Let $M = (Q, \Sigma, \Gamma, \delta, q_1, B, F)$ be a Turing machine. If it is used to compute functions then we may assume that:

- $\Sigma = \{0, 1\}$;
- $F = \emptyset$;
- $Q = \{q_1, \ldots, q_n\}$, $n \geq 1$;
- $\Gamma = \{X_1, \ldots, X_m\}$, where $X_1 = 0$, $X_2 = 1$, $X_3 = B$, and $m \geq 3$.

Therefore, $M$ halts only if a blocking configuration is reached.
2.2.4. Turing-computable Functions are Recursive

Configurations \((q, u|av)\) of \(M\) are encoded as follows:

- \((q, u|av) \sim (i, |u|, G(uav)) \sim J(3)(i, |u|, G(uav)) \in \mathbb{N};\)
- \(G(uav)\) is the Gödel number of the sequence \(uav\), i.e.,

\[
G(uav) = p^n(0)^{i_1} \cdots p^n(k-1)^{i_k}
\]

assuming \(uav = X_{i_1} \cdots X_{i_k}\).

**Example 16**

- \((q_3, |010) \sim (3, 0, 2^1 \cdot 3^2 \cdot 5^1) \sim J(3)(3, 0, 90) = 4959\)
- \((q_2, 100|B01) \sim (2, 3, 2^2 \cdot 3^1 \cdot 5^1 \cdot 7^3 \cdot 11^1 \cdot 13^2) \sim J(3)(2, 3, 38258220) = J(2)(17, 38258220) = 12,441,388,261,154,451\)
2.2.4. Turing-computable Functions are Recursive

Given a configuration $C$, denote by $JG(C)$ its code (defined as above).

Define $c_M : \mathbb{N}^2 \rightarrow \mathbb{N}$ by

$$c_M(x, n) = \begin{cases} 
JG(C_n), & \text{if there exists a computation } C_0(x), \ldots, C_n \\
JG(C_{n_0}), & \text{if there exists a computation } C_0(x), \ldots, C_{n_0}, \text{ } C_{n_0} \text{ is a halting configuration, and } n > n_0,
\end{cases}$$

for all $x, n \in \mathbb{N}$.

**Lemma 11** The function $c_M$ is primitive recursive.
Define $nr_M : \mathbb{N} \rightarrow \mathbb{N}$ by

$$nr_M(x) = \begin{cases} 
  n, & \text{if there exists a computation} \\
  C_0(x), \ldots, C_n, \text{ and } C_n \text{ is a} \\
  \text{halting configuration} \\
  \uparrow, & \text{otherwise},
\end{cases}$$

for all $x \in \mathbb{N}$.

**Lemma 12** The function $nr_M$ is recursive.

**Proof** $nr_M(x) = \mu(y)[c_M(x, y) = c_M(x, y + 1)]$. \qed
2.2.4. Turing-computable Functions are Recursive

Without loss of generality we may assume that any blocking configuration of $M$ is of the form $(q, |0^m)$, for some $m \geq 0$ (assume $0^m = B$ whenever $m = 0$).

Define $\text{decode} : \mathbb{N} \rightarrow \mathbb{N}$ by

$$\text{decode}(z) = \left( 1 - \text{eq}(I_3^{(3)}(z), pn(0)^3) \right)$$

$$\left( \mu(y \leq z)[I_3^{(3)}(z) = \prod_{i=0}^{y-1} pn(i)] \right),$$

for all $z \in \mathbb{N}$.

**Lemma 13** $\text{decode}$ is primitive recursive.
2.2.4. Turing-computable Functions are Recursive

**Lemma 14** The unary function $f : \mathbb{N} \to \mathbb{N}$ computed by a Turing machine $M$ is given by

$$f(x) = \text{decode}(c_M(x, nr_M(x))),$$

for all $x \in \mathbb{N}$.

**Theorem 7** Any Turing-computable function is recursive.

**Corollary 5** $RF = TCF$. 
2.3. Properties of Recursive Functions

2.3.1. Universal functions
2.3.2. The s-m-n theorem
2.3.3. Recursively enumerable sets
2.3.1. Universal Functions

Recall that, to compute functions, Turing machines can be considered in the form \( M = (Q, \Sigma, \Gamma, \delta, q_1, B, F) \), where:

- \( Q = \{q_1, \ldots, q_n\}, \ n \geq 1 \);
- \( \Sigma = \{0, 1\} \);
- \( \Gamma = \{0, 1, B\} \);
- \( F = \emptyset \).

Moreover,

- denote \( 0 = X_1, \ 1 = X_2, \ B = X_3 \), and
- denote \( L = D_1, \ R = D_2 \), and
- assume that for any state \( q \) there exists at least one transition \( \delta(q_i, X_j) = (q_k, X_l, D_m) \) such that \( q = q_i \) or \( q = q_k \).
Let $M = (Q, \Sigma, \Gamma, \delta, q_1, B, F)$ a Turing machine as given above. We encode $M$ as follows:

- encode each transition $\delta(q_i, X_j) = (q_k, X_l, D_m)$ by $0^i10^j10^k10^l10^m$
- assume that $M$ has $r$ transitions and their codes are $\text{code}_1, \ldots, \text{code}_r$;
- encode $M$ by:

$$111\text{code}_111\text{code}_211 \cdots 11\text{code}_{r-1}11\text{code}_r111$$
Example 17 Let $M = (\{q_1, q_2, q_3\}, \{0, 1\}, \{0, 1, B\}, \delta, q_1, B, \emptyset)$, where:

- $\delta(q_1, 1) = (q_3, 0, R)$,
- $\delta(q_3, 0) = (q_1, 1, R)$,
- $\delta(q_3, 1) = (q_2, 0, R)$, and
- $\delta(q_3, B) = (q_3, 1, L)$.

The following binary sequence is a code of this machine:

$$11101001000101000100010100100100111000101010010010011$$

$$00010010010100110001000100100100111$$
2.3.1. Universal Functions

Remark:

1. Except for the totally undefined Turing machine or Turing machines with exactly one transition, all the other Turing machines have more than one code, but finitely many.

2. If we order lexicographically the transitions of Turing machines, then each Turing machine has exactly one code. The lexicographic order is given by:
   - order first by states, \( q_1 < q_2 < \cdots < q_n \),
   - order then by input symbols, \( 0 < 1 < B \).

   The unique code such obtained is denoted by \( \langle M \rangle \).

3. Distinct Turing machines have distinct codes.
2.3.1. Universal Functions

- The code of a Turing machine is a binary sequence. If we order lexicographically these binary sequences, we get the standard enumeration of Turing machines
  
  \[ M_0, M_1, \ldots \]

- The standard enumeration of Turing machines induces a standard enumeration of all \( n \)-ary recursive functions:
  
  \[ f_0^{(n)}, f_1^{(n)}, \ldots \]

  where \( f_i^{(n)} \) is the \( n \)-ary function computed by \( M_i \);

- As there are infinitely many Turing machines computing the same \( n \)-ary function, each function \( f_i^{(n)} \) appears infinitely many times in the standard enumeration of all \( n \)-ary recursive functions.
2.3.1. Universal Functions

Given $M_i$ and $x_1, \ldots, x_n \in \mathbb{N}$, we denote by $\langle M, x_1, \ldots, x_n \rangle$ the binary string

$$\langle M_i \rangle 0^{x_1}1 \ldots 10^{x_n}$$

**Theorem 8** There exists a Turing machine $M_u$ that, on input $\langle M_i, x_1, \ldots, x_n \rangle$, computes $f_i^{(n)}(x_1, \ldots, x_n)$, for all $i$ and $x_1, \ldots, x_n$.

**Proof** For the sake of simplicity we will consider only the case $n = 1$.

We shall exhibit a 3-track Turing machine $M_u$:

- the first track holds the input $\langle M_i, x \rangle$;
2.3.1. Universal Functions

- the second track is used to simulate the tape of $M_i$;
- the third track holds the current state of $M_i$, with $q_i$ represented by $0^i$. Initially, it holds $0^1$ (i.e., $q_1$).

The behavior of $M_u$ is as follows:

- mark the first cell so the head can find its way back;
- $M_u$ checks the input to see that $\langle M_i \rangle$ is the code of some Turing machine;
- copy $0^x$ on the second track and initialize track 3 to hold $0$ ($q_1$);
2.3.1. Universal Functions

- let \(0^i (q_i)\) be the current content of track 3 and \(X_j\) be the current symbol scanned on track 2.
  - \(M_u\) scans the first track (from left to right) looking for a substring which begins with \(110^i10^j1\);
  - If no such string is found, \(M_u\) halts (in a blocking configuration);
  - If such a code is found, let it be \(0^i10^j10^k10^l10^m\). Then, it puts \(0^k\) on track 3, replaces \(X_j\) by \(X_l\) on the second track, and moves in the direction pointed by \(D_m\);
- When \(M\) halts, \(M_u\) halts.
A Turing machine as the one in the proof above is called an **universal Turing machine**.

**Corollary 6** There exist \( n \)-ary universal functions, for any \( n \geq 2 \).

**Proof** Modify an arbitrary universal Turing machine \( M_u \) as follows:

- start off with \( 0^i10^{x_1}1 \ldots 10^{x_n} \) on the tape;
- generate the code of \( M_i \) (use a new track if needed);
- call \( M_u \) on \( \langle M_i, x_1, \ldots, x_n \rangle \).

The machine such obtained computes an universal function \( u^{(n+1)} \), for all \( n \geq 1 \). \( \square \)
Theorem 9 (s-m-n theorem)
For any recursive function $f_i^{(m+n)}$ there exists a total recursive function $s_{m,n}^{(m+1)}$ such that:

$$f_i^{(m+n)}(x_1, \ldots, x_m, y_1, \ldots, y_n) = f_n^{(m+1)}(s_{m,n}^{(m+1)}(i, x_1, \ldots, x_m) (y_1, \ldots, y_n),$$

for all $x_1, \ldots, x_m, y_1, \ldots, y_n \in \mathbb{N}$.

Proof use Turing machines
A set $A \subseteq \mathbb{N}$ is called recursively enumerable, abbreviated r.e., if $A = \emptyset$ or $A$ is the range of an unary total recursive function.

Remark: If $A \subseteq \mathbb{N}$ is a non-empty r.e. set and $f$ is an unary total recursive function such that $A = \text{cod}(f)$, then we will say that $f$ enumerates $A$. Why?

- $0 \mapsto f(0) \in A$;
- $1 \mapsto f(1) \in A$;
- $\cdots$
- $A = \{f(x)|x \geq 0\}$. 
Remark: \( A \subseteq \mathbb{N} \) is an r.e. set if \( A = \emptyset \) or there exists a Turing machine \( M \) such that:

- \( M \) halts on all inputs \( x \in \mathbb{N} \);
- for each input \( x \), when \( M \) halts on \( x \) then the output \( M(x) \) is in \( A \);
- for any \( a \in A \) there exists \( x \in \mathbb{N} \) such that \( a = M(x) \)

(the machine \( M \) enumerates \( A \)).

We may say that:

\( A \) is an r.e. set if \( A = \emptyset \) or there exists an algorithm \( \mathcal{A} \) that enumerates all the elements of \( A \).
2.3.3. Recursively Enumerable Sets

Notation:

- $PRS = \text{the set of all primitive recursive sets } A \subseteq \mathbb{N};$
- $RS = \text{the set of all recursive sets } A \subseteq \mathbb{N};$
- $RES = \text{the set of all r.e. sets } A \subseteq \mathbb{N};$

Proposition 21

$$PRS \subseteq RS \subseteq RES$$
2.3.3. Recursively Enumerable Sets

Theorem 10

1. A set is r.e. iff it is the range of an unary recursive function.
2. A set is r.e. iff it is the domain of an unary recursive function.

Remark: The above theorem can be used to show that certain sets are/are not r.e.:

- use (1.) to show that a given set is r.e.;
- use (2.) to show that a given set is not r.e.
2.3.3. Recursively Enumerable Sets

Example 18

1. The set \( A = \{ x | f_x(x) \downarrow \} \) is r.e.
2. The set \( N - A \), where \( A \) is the set from (1.), is not r.e.
3. The set \( B = \{ x | f_x \text{ is total} \} \) is not r.e.
4. The set \( N - B \), where \( B \) is the set from (3.), is not r.e.

Proposition 22 The set \( RES \) is closed under union and intersection but not under complementation.

From this proposition follows that \( RS \subset RES \). Moreover,

\[ PRS \subset RS \subset RES \]
Theorem 11 A set $A$ is recursive iff $A$ and $\mathbb{N} - A$ are r.e.

Proof $\Rightarrow$: straightforward.

$\Leftarrow$: If $A = \emptyset$ or $\mathbb{N} - A = \emptyset$, then $A$ is recursive.

Assume that both $A$ and $\mathbb{N} - A$ are non-empty. Let $g_1$ and $g_2$ be two total recursive functions such that $A = \text{cod}(g_1)$ and $\mathbb{N} - A = \text{cod}(g_2)$.

Let $h(x) = \mu(y)[g_1(y) = x \lor g_2(y) = x]$. $h$ is recursive and

$$
\chi_A(x) = \begin{cases} 
1, & g_1(h(x)) = x \\
0, & \text{otherwise}
\end{cases}
$$

Therefore, $A$ is recursive. \qed
2.3.3. Recursively Enumerable Sets

Remark: Given a subset $A \subseteq \mathbb{N}$, define the language of $A$ as being the set $L_A = \{0^x | x \in A\} \subseteq \{0, 1\}^*$. Then:

$A \subseteq \mathbb{N}$ is recursive $\iff$ $\chi_A$ is total recursive
$\iff (\exists$ a Turing machine $M$)
$(M$ computes $\chi_A$)
$\iff (\exists$ a Turing machine $M'$)
$(L_A = \text{accept}(M') \land \overline{L_A} = \text{reject}(M'))$

The second equivalence is based on:

- Given $M$ which computes $\chi_A$, design $M'$ which accepts when $M$ halts in a configuration $(q, |0)$, and rejects when $M$ halts in a configuration $(q, |B)$.
Remark:

$A \subseteq \mathbb{N}$ is r.e. $\iff A = \text{dom}(f)$ for some recursive function
$\iff (\exists$ a Turing machine $M$)
$(L_A = \{0^x|M \text{ halts on } 0^x\})$
$\iff (\exists$ a Turing machine $M'$)
$(L_A = \text{accept}(M'))$

The second equivalence is based on:

- Given $M$ which computes $f$, design $M'$ which accepts $0^x$
  if and only if $M$ halts on $0^x$. 
2.3.3. Recursively Enumerable Sets

**Extension to languages**

Let \( \Sigma \) be an alphabet and \( L \subseteq \Sigma^* \).

We say that \( L \) is **recursive** if there exists a Turing machine \( M \) such that \( L = \text{accept}(M) \) and \( \overline{L} = \text{reject}(M) \).

We say that \( L \) is **recursively enumerable**, abbreviated r.e., if there exists a Turing machine \( M \) such that \( L = \text{accept}(M) \).

Denote:

- \( RS(\Sigma) = \) the set of all recursive sets \( L \subseteq \Sigma^* \);
- \( RES(\Sigma) = \) the set of all r.e. sets \( L \subseteq \Sigma^* \);
- \( \mathcal{L}_{\text{rec}} = \) the class of all recursive languages;
- \( \mathcal{L}_{\text{re}} = \) the class of all r.e. languages.
2.3.3. Recursively Enumerable Sets

Remark: The main difference between recursive and r.e. sets:
- If \( L \) is recursive \( \Rightarrow (\exists M)(L = \text{accept}(M) \land \overline{L} = \text{reject}(M)) \)
  - given \( w \in \Sigma^* \), if \( w \in L \) then \( M \) accepts \( w \) and, therefore, we know that \( w \) is a member of \( L \);
  - given \( w \in \Sigma^* \), if \( w \notin L \) then \( M \) rejects \( w \) and, therefore, we know that \( w \) is not a member of \( L \);
- If \( L \) is r.e. \( \Rightarrow (\exists M)(L = \text{accept}(M)) \)
  - given \( w \in \Sigma^* \), if \( w \in L \) then \( M \) accepts \( w \) and, therefore, we know that \( w \) is a member of \( L \);
  - given \( w \in \Sigma^* \), if \( w \notin L \) then \( M \) either rejects \( w \) or loops. In this case we might not know that \( w \) is not a member of \( L \).

Recursive sets = membership problem is decidable
r.e. sets = membership problem is semi-decidable
### Theorem 12

1. $RS(\Sigma)$ is closed under union, intersection, and complementation;
2. $RES(\Sigma)$ is closed under union and intersection, but not under complementation;
3. $RS(\Sigma) \subseteq RES(\Sigma)$;
4. $L \subseteq \Sigma^*$ is recursive iff $L$ and $\Sigma^* - L$ are r.e.
5. $\mathcal{L}_{rec} \subseteq \mathcal{L}_{re} = \mathcal{L}_0$ ($\mathcal{L}_0$ is the class of type 0 Chomsky languages).