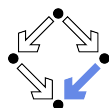


Turing Complete Computational Models

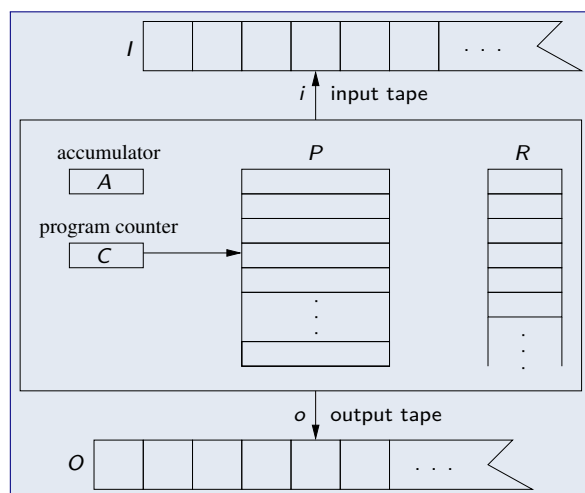
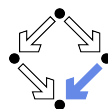
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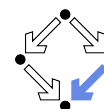
1. Random Access Machines
2. Loop and While Programs
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A Random Access Machine



A model closer to a real computer.

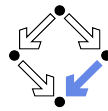
A Random Access Machine



- A **random access machine (RAM)**:
 - an infinite input tape I (whose cells can hold natural numbers of arbitrary size) with a read head position $i \in \mathbb{N}$,
 - an infinite output tape O (whose cells can hold natural numbers of arbitrary size) with a write head position $o \in \mathbb{N}$,
 - an accumulator A which can hold a natural number of arbitrary size,
 - a program counter C which can hold an arbitrary natural number,
 - a program consisting of a finite number of instructions $P[1], \dots, P[m]$,
 - a memory consisting of a countably infinite sequence of registers $R[1], R[2], \dots$, each of which can hold an arbitrary natural number.
- **Execution**:
 - Initially, $i = 0$, $o = 0$, $A = 0$, $C = 1$, $R[1] = R[2] = \dots = 0$.
 - In every step, the RAM reads $P[C]$, increments C by 1, and then performs the action indicated by the instruction.
 - Execution terminates when $C = 0$.

Program is a sequence of machine instructions.

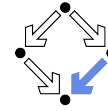
RAM Instructions



Instruction	Description	Action
IN	Read value from input tape into accumulator	$A := I[i]; i := i + 1$
OUT	Write value from accumulator to output tape	$O[o] := A; o := o + 1$
LOAD # n	Load constant n into accumulator	$A := n$
LOAD n	Load content of register n into accumulator	$A := R[n]$
LOAD (n)	Load content of register referenced by reg. n	$A := R[R[n]]$
STORE n	Store content of accumulator into register n	$R[n] := A$
STORE (n)	Store content into register referenced by reg. n	$R[R[n]] := A$
ADD # n	Increment content of accumulator by constant	$A := A + n$
SUB # n	Decrement content of accumulator by constant	$A := \max\{0, A - n\}$
JUMP n	Unconditional jump to instruction n	$C := n$
BEQ i, n	Conditional jump to instruction n	if $A = i$ then $C := n$

Immediate addressing, direct addressing, indirect addressing.

Example

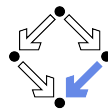


```

START:  LOAD #1    A := 1
        STORE 1    R[1] := A
READ:   LOAD 1     A := R[1]
        ADD #1     A := A + 1
        STORE 1    R[1] := A
        IN        A := I[i]; i := i + 1
        BEQ 0, WRITE if A = 0 then C := WRITE
        STORE (1)  R[R[1]] := A
        JUMP READ  C := READ
WRITE:  LOAD 1     A := R[1]
        SUB #1     A := A - 1
        STORE 1    R[1] := A
        BEQ 1, HALT if A = 1 then C := HALT
        LOAD (1)   A := R[R[1]]
        OUT        O[o] := A; o := o + 1
        JUMP WRITE C := WRITE
HALT:   JUMP 0     C := 0
    
```

Reads $x_1, \dots, x_n, 0$ and writes x_n, \dots, x_1 using stack $R[2], \dots, R[n+1]$.

RAMs versus Turing Machines

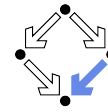


Theorem: Every Turing machine can be simulated by a RAM.

- RAM uses registers $R[1], \dots, R[c-1]$ for its own purposes,
- stores in $R[c]$ the position of the tape head of the Turing machine,
- uses $R[c+1], R[c+2], \dots$ as a virtual Turing machine tape.
 - Using "indirect addressing" operations $\text{LOAD}(n)$ and $\text{STORE}(n)$.
- RAM copies the input from the input tape into its virtual tape, then it mimics the execution of the Turing machine on the virtual tape.
- When the simulated Turing machine terminates, the content of the virtual tape is copied to the output tape.

RAMs represent a Turing complete computational model.

RAMs versus Turing Machines

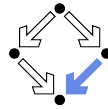


Theorem: Every RAM can be simulated by a Turing machine.

- The Turing machine uses 5 tapes to simulate the RAM:
 - Tape 1 represents the input tape of the RAM.
 - Tape 2 represents the output tape of the RAM.
 - Tape 3 holds a representation of that part of the memory that has been written by the simulation of the RAM.
 - Tape 4 holds a representation of the accumulator of the RAM.
 - Tape 5 serves as a working tape.
- Tape 3 holds a sequence of (address, contents) pairs that represent those registers of the RAM that have been written during the simulation (the contents of all other registers hold 0).
- Every instruction of the RAM is simulated by a sequence of steps of the Turing machine which reads respectively writes Tape 1 and 2 and updates on Tape 3 and 4 the tape representations of the contents of the memory and the accumulator.

RAMs are not more powerful than Turing machines.

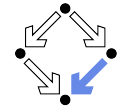
Random Access Stored Program Machine



The program of a RAM is “read-only”.

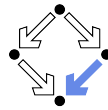
- **Random Access Stored Program Machine (RASP).**
 - A RAM variant where the program is stored in memory R (there is no separate program store P).
- **Every RASP can be simulated by a RAM.**
 - RAM is interpreter for RASP instructions (like a *microprogram* in a processor interprets machine instructions).
- **Every RAM can be simulated by a RASP.**
 - Even if indirect addressing is removed from RASP.
 - RAM instructions $\text{LOAD}(n)$ and $\text{STORE}(n)$ can be interpreted by self-modifying RASP code.

Self modifying programs do not add computational power to a RAM.



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Loop Programs



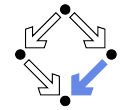
- **Loop Program P :**

$$P ::= x_i := 0 \mid x_i := x_j + 1 \mid x_i := x_j - 1 \mid P; P \\ \mid \text{loop } x_i \text{ do } P \text{ end.}$$

- Set $\{x_0, x_1, x_2, \dots\}$ of program variables.
- Initial value of x_i determines the number of loop iterations.
- Loop must eventually terminate.

Programs with bounded iteration that necessarily terminate.

Semantics



- **Semantics $\llbracket P \rrbracket(m)$** maps the start memory $m : \mathbb{N} \rightarrow \mathbb{N}$ to the final memory after the termination of P :

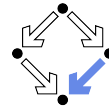
$$\begin{aligned} \llbracket x_i := 0 \rrbracket(m) &:= m[i \leftarrow 0] \\ \llbracket x_i := x_j + 1 \rrbracket(m) &:= m[i \leftarrow m(j) + 1] \\ \llbracket x_i := x_j - 1 \rrbracket(m) &:= m[i \leftarrow \max\{0, m(j) - 1\}] \\ \llbracket P_1; P_2 \rrbracket(m) &:= \llbracket P_2 \rrbracket(\llbracket P_1 \rrbracket(m)) \\ \llbracket \text{loop } x_i \text{ do } P \text{ end} \rrbracket(m) &:= \llbracket P \rrbracket^{m(i)}(m) \end{aligned}$$

- $m[i \leftarrow n]$: memory m after updating the value x_i by value n .
- $\llbracket P \rrbracket^n(m)$: memory m after n times executing P :

$$\begin{aligned} \llbracket P \rrbracket^0(m) &:= m \\ \llbracket P \rrbracket^{n+1}(m) &:= \llbracket P \rrbracket(\llbracket P \rrbracket^n(m)) \end{aligned}$$

A loop program denotes a function over memories.

Syntactic Abbreviations



- $x_i := x_j$

```
x_i := x_j + 1; x_j := x_i - 1
```

- $x_i := n$

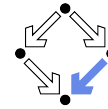
```
x_i := 0; x_i := x_i + 1; x_i := x_i + 1; ...; x_i := x_i + 1
```

- **if** $x_i = 0$ **then** P_t **else** P_e **end**

```
x_t := 1; loop x_i do x_t := 0; end;  
x_e := 1; loop x_t do x_e := 0; end;  
loop x_t do  $P_t$  end; loop x_e do  $P_e$  end;
```

The usual programming language constructs (except for unbounded iteration) can be represented.

Loop Computability



We consider the computability of functions over the natural numbers.

$f : \mathbb{N}^n \rightarrow \mathbb{N}$ is **loop computable**, if there exists a loop program P such that for all $x_1, \dots, x_n \in \mathbb{N}$ and memory $m : \mathbb{N} \rightarrow \mathbb{N}$ defined as

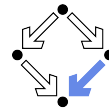
$$m(i) := \begin{cases} x_i & \text{if } 1 \leq i \leq n \\ 0 & \text{else} \end{cases}$$

we have

$$\llbracket P \rrbracket(m)(0) = f(x_1, \dots, x_n)$$

When started in a state where x_1, \dots, x_n contain the arguments of f , the program terminates in a state where x_0 holds the result of f .

Example



- Addition is computable by the program $x_0 := x_1 + x_2$:

```
x_0 := x_1;  
loop x_2 do  
  x_0 := x_0 + 1  
end
```

- Multiplication is computable by the program $x_0 := x_1 \cdot x_2$:

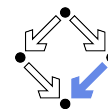
```
x_0 := 0;  
loop x_2 do  
  x_0 := x_0 + x_1  
end
```

- Exponentiation is computable by the program $x_0 := x_1^{x_2}$:

```
x_0 := 1;  
loop x_2 do  
  x_0 := x_0 · x_1  
end
```

Natural number arithmetic is loop computable.

Arithmetic



- $x_0 := x_1 \cdot x_2$:

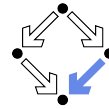
```
x_0 := 0;  
loop x_2 do  
  x_0 := x_0 + x_1  
end
```

~>

```
x_0 := 0;  
loop x_2 do  
  x_0 := x_0;  
  loop x_1 do  
    x_0 := x_0 + 1  
  end  
end
```

Higher arithmetic needs multiply nested loops.

Beyond Exponentiation

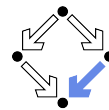


$$a \uparrow^n b := \begin{cases} a^b & \text{if } n = 1 \\ 1 & \text{if } b = 0 \\ a \uparrow^{n-1} (a \uparrow^n (b-1)) & \text{else} \end{cases}$$

- $a \uparrow^1 b = a^b$
 $a \uparrow^1 b = a \cdot a \cdot \dots \cdot a$ (b times)
- $a \uparrow^2 b = a^{a^{\dots^a}}$ (b times)
 $a \uparrow^2 b = a \uparrow^1 a \uparrow^1 \dots \uparrow^1 a$ (b times)
- $a \uparrow^3 b$:
 $a \uparrow^3 b = a \uparrow^2 a \uparrow^2 \dots \uparrow^2 a$ (b times)

The notation allows to define arbitrary “complex” arithmetic functions.

While Programs



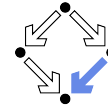
- **While Program** P :

$$P ::= \dots \text{ (as for loop programs)} \\ | \text{ while } x_i \text{ do } P \text{ end.}$$

- Set $\{x_0, x_1, x_2, \dots\}$ of program variables.
- Loop is repeated as long as $x_i \neq 0$.
- If $x_i \neq 0$ forever, loop does not terminate.

Programs with unbounded iteration that may not terminate.

Limits of Loop Computability



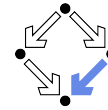
- **Theorem:** for every $n > 0$ and $f(a, b) := a \uparrow^n b$
 - f is loop computable, and
 - every loop program computing f requires at least $n+2$ nested loops.
- **Theorem:** $g : \mathbb{N}^3 \rightarrow \mathbb{N}, g(a, b, n) := a \uparrow^{n+1} b$ is not loop computable.
 - Assume g can be computed by a program P with n loops.
 - Then the computation of $g(a, b, n) = a \uparrow^{n+1} b$ requires $n+3$ loops.
 - Thus P cannot compute g .
- Also the **Ackermann Function** is not loop computable:

$$\begin{aligned} \text{ack}(0, m) &:= m + 1 \\ \text{ack}(n, 0) &:= \text{ack}(n-1, 1) \\ \text{ack}(n, m) &:= \text{ack}(n-1, \text{ack}(n, m-1)), \text{ if } n > 0 \wedge m > 0 \end{aligned}$$

- $\text{ack}(n, m) = 2 \uparrow^{n-2} (m+3) - 3$
- $\text{ack}(4, 2)$ has 20,000 digits.

Some arithmetic functions grow “too fast” to be loop computable.

Semantics



- **Semantics** $\llbracket P \rrbracket(m)$ maps start memory $m : \mathbb{N} \rightarrow \mathbb{N}$
 - to the final memory, if P terminates, and
 - to the special value \perp (bottom), if P does not terminate.
- Semantics generalizes that of loop programs:

$$\llbracket P \rrbracket(m) := \begin{cases} \perp & \text{if } m = \perp \\ \llbracket P \rrbracket'(m) & \text{else} \end{cases}$$

$$\llbracket \dots \rrbracket'(m) := \dots \text{ (as for loop programs)}$$

- Semantics of unbounded iteration:

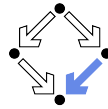
$$\llbracket \text{while } x_i \text{ do } P \text{ end} \rrbracket'(m) := \begin{cases} \perp & \text{if } L_i(P, m) \\ \llbracket P \rrbracket^{T_i(P, m)}(m) & \text{else} \end{cases}$$

$$L_i(P, m) := \Leftrightarrow \forall k \in \mathbb{N} : \llbracket P \rrbracket^k(m)(i) \neq 0$$

$$T_i(P, m) := \min \{ k \in \mathbb{N} \mid \llbracket P \rrbracket^k(m)(i) = 0 \}$$

A while program denotes a function whose result is either a memory or \perp .

Syntactic Abbreviations



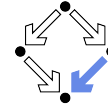
- **while** $x_i < x_j$ **do** P **end**

```

 $x_k := x_j - x_i;$ 
while  $x_k$  do  $P; x_k := x_j - x_i;$  end
    
```

Analogous constructions possible for other termination conditions.

While Computability



$f : \mathbb{N}^n \rightarrow_{\text{p}} \mathbb{N}$ is **while computable**, if there exists a while program P such that for all $x_1, \dots, x_n \in \mathbb{N}$ and memory $m : \mathbb{N} \rightarrow \mathbb{N}$ defined as

$$m(i) := \begin{cases} x_i & \text{if } 1 \leq i \leq n \\ 0 & \text{else} \end{cases}$$

the following holds:

- If $x_1, \dots, x_n \in \text{domain}(f)$, then $\llbracket P \rrbracket(m) : \mathbb{N} \rightarrow \mathbb{N}$ and

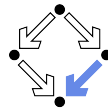
$$\llbracket P \rrbracket(m)(0) = f(x_1, \dots, x_n)$$

- If $x_1, \dots, x_n \notin \text{domain}(f)$, then

$$\llbracket P \rrbracket(m) = \perp$$

For a defined value of $f(x_1, \dots, x_n)$, P terminates with this value in variable x_0 . If $f(x_1, \dots, x_n)$ is undefined, the program does not terminate.

Example



The Ackermann function is while computable with the help of a stack.

```

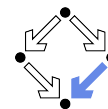
function  $ack(n, m)$ :
  if  $n = 0$  then
    return  $m + 1$ 
  else if  $m = 0$  then
    return  $ack(n - 1, 1)$ 
  end if
  return  $ack(n - 1, ack(n, m - 1))$ 
end function
    
```

```

function  $ack(x_1, x_2)$ :
  push( $x_1$ ); push( $x_2$ )
  while  $size() > 1$  do
     $x_2 \leftarrow pop()$ ;  $x_1 \leftarrow pop()$ 
    if  $x_1 = 0$  then
      push( $x_2 + 1$ )
    else if  $x_2 = 0$  then
      push( $x_1 - 1$ ); push(1);
    else
      push( $x_1 - 1$ );
      push( $x_1$ ); push( $x_2 - 1$ )
    end if
  end while
  return  $pop()$ 
end function
    
```

While programs are computationally more powerful than loop programs.

Normal Form of a While Program



Kleene's Normal Form Theorem: every while computable function can be computed by a while program in Kleene's normal form:

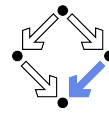
```

 $x_c := 1;$ 
while  $x_c$  do
  if  $x_c = 1$  then  $P_1$ 
  else if  $x_c = 2$  then  $P_2$ 
  ...
  else if  $x_c = n$  then  $P_n$ 
  end if
end while
    
```

- P_1, \dots, P_n do *not* contain while loops.
- Control variable x_c determines which P_i to execute next.

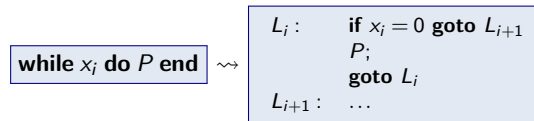
A single while loop is all that is needed.

Normal Form of a While Program

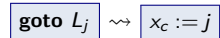


We sketch the proof of Kleene's Normal Form Theorem.

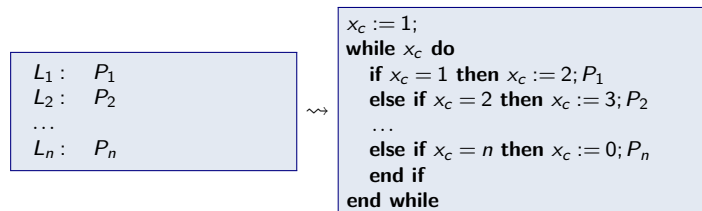
- A while program can be translated into a goto program:



- Gotos can be translated to control variable assignments:

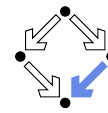


- The resulting program can be translated into normal form:



In essence, the execution loop of a processor.

Turing Machines and While Programs

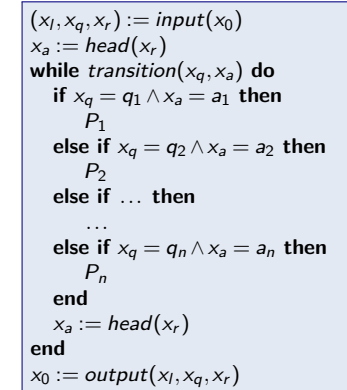


- Theorem:** Every Turing machine can be simulated by a while program and vice versa.

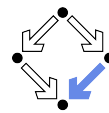
- Consequence: every Turing computable function is while computable and vice versa.

Proof \Rightarrow : construct P to simulate M .

- x_0 holds initial tape content.
 - Determines initial configuration.
- Machine configuration (x_l, x_q, x_r) :
 - x_q : the current state.
 - x_l : the tape left to the tape head.
 - x_r : the tape under/right to head.
- State x_q and symbol x_a under head determine the state transition.
 - If none is possible, final tape content is written to x_0 .

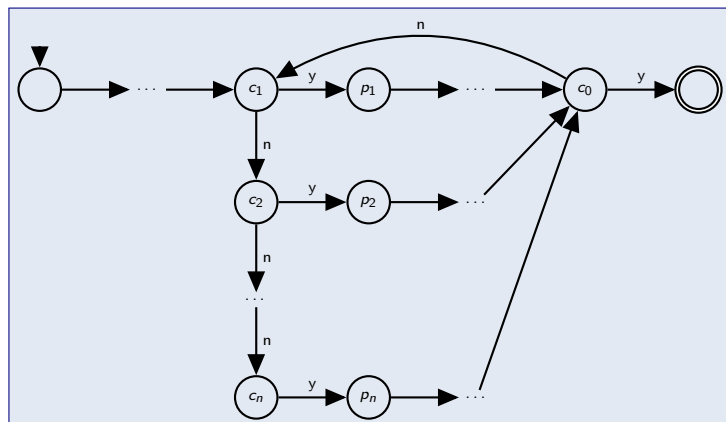


Turing Machines and While Programs



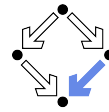
Proof \Leftarrow : construct M to simulate P (given in normal form).

- Each program fragment P_i is translated into a corresponding fragment of the transition function of M with sequence of states c_i, p_i, \dots, c_0 .



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Primitive Recursive Functions



The following functions over the natural numbers are **primitive recursive**:

- The **constant null** function $0 \in \mathbb{N}$.
- The **successor** function $s : \mathbb{N} \rightarrow \mathbb{N}, s(x) := x + 1$.
- The **projection** functions $p_i^n : \mathbb{N}^n \rightarrow \mathbb{N}, p_i^n(x_1, \dots, x_n) := x_i$.
- Every function $h : \mathbb{N}^n \rightarrow \mathbb{N}$ defined by **composition**

$$h(x_1, \dots, x_n) := f(g_1(x_1, \dots, x_n), \dots, g_k(x_1, \dots, x_n))$$

from primitive recursive $f : \mathbb{N}^k \rightarrow \mathbb{N}$ and $g_1, \dots, g_k : \mathbb{N}^n \rightarrow \mathbb{N}$.

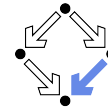
- Every function $h : \mathbb{N}^{n+1} \rightarrow \mathbb{N}$ defined by **primitive recursion**

$$h(y, x_1 \dots x_n) := \begin{cases} f(x_1, \dots, x_n) & \text{if } y = 0 \\ g(y-1, h(y-1, x_1, \dots, x_n), x_1, \dots, x_n) & \text{else} \end{cases}$$

from primitive recursive $f : \mathbb{N}^n \rightarrow \mathbb{N}$ and $g : \mathbb{N}^{n+2} \rightarrow \mathbb{N}$.

Starting with the base functions, by composition and primitive recursion new primitive recursive functions can be defined.

Understanding Primitive Recursion



- Primitive recursion can be defined by a **pattern matching equation**:

$$\begin{aligned} h(0, x_1, \dots, x_n) &:= f(x_1, \dots, x_n) \\ h(y+1, x_1, \dots, x_n) &:= g(y, h(y, x_1, \dots, x_n), x_1, \dots, x_n) \end{aligned}$$

- Primitive recursion can be defined by a **pattern matching construct**:

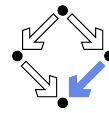
$$\begin{aligned} h(y, x_1 \dots x_n) &:= \\ \text{case } y \text{ of} & \\ 0: & f(x_1, \dots, x_n) \\ z+1: & g(z, h(z, x_1, \dots, x_n), x_1, \dots, x_n) \end{aligned}$$

- $h(y, x)$ denotes the $(y-1)$ -times application of g starting with $f(x)$:

$$\begin{aligned} h(0, x) &= f(x) \\ h(1, x) &= g(0, h(0, x), x) = g(0, f(x), x) \\ h(2, x) &= g(1, h(1, x), x) = g(1, g(0, f(x), x), x) \\ h(3, x) &= g(2, h(2, x), x) = g(2, g(1, g(0, f(x), x), x), x) \\ &\dots \end{aligned}$$

$$h(y, x) = g(y-1, h(y-1, x), x) = g(y-1, g(y-2, \dots, g(0, f(x), x), \dots, x), x)$$

Example



We consider arithmetic on natural numbers.

- **Addition** $y + x$ is primitive recursive:

$$\begin{aligned} 0 + x &:= x \\ (y+1) + x &:= (y+x) + 1 \end{aligned}$$

- **Multiplication** $y \cdot x$ is primitive recursive:

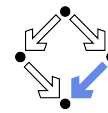
$$\begin{aligned} 0 \cdot x &:= 0 \\ (y+1) \cdot x &:= y \cdot x + x \end{aligned}$$

- **Exponentiation** x^y is primitive recursive:

$$\begin{aligned} x^0 &:= 1 \\ x^{y+1} &:= x^y \cdot x \end{aligned}$$

Natural number arithmetic is primitive recursive.

Primitive Recursion and Loop Computability



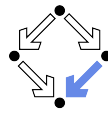
Both the execution of a loop program and the evaluation of a primitive recursive function are bounded; are they equally expressive?

Example: Compute in x_0 the smallest $n < x_1$ for which $p(n) = 1$ holds (respectively $x_0 = x_1$, if $p(n) \neq 1$ for all $n < x_1$).

$x_0 := x_1$	Assume $n = 3$:		
$x_2 := 0$			
loop x_1 do	x_0	x_1	x_2
if $x_0 = x_1 \wedge p(x_2) = 1$ then	5	5	0
$x_0 := x_2$	5	5	1
end	5	5	2
$x_2 := x_2 + 1$	5	5	3
end	3	5	4
	3	5	5

We will construct a primitive recursive function computing the same value.

Primitive Recursion and Loop Computability

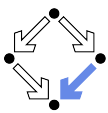


We mimic the execution of the **loop** by a primitive recursive function *loop* whose recursion parameter denotes the number of loop iterations.

$$\begin{aligned} \min(x_1) &:= \text{loop}(x_1, x_1) \\ \text{loop}(x_2, x_1) &:= \begin{cases} x_1 & \text{if } x_2 = 0 \\ \text{if}(x_2 - 1, \text{loop}(x_2 - 1, x_1), x_1) & \text{else} \end{cases} \\ \text{if}(x_2, x_0, x_1) &:= \begin{cases} x_2 & \text{if } x_0 = x_1 \wedge p(x_2) = 1 \\ x_0 & \text{else} \end{cases} \end{aligned}$$

- $\min(x_1) := \text{loop}(x_1, x_1)$ computes the value assigned to x_0 for input x_1 (2nd argument) after x_1 iterations of the **loop** (1st argument).
- $\text{loop}(x_2, x_1)$ computes the value assigned to x_0 for input x_1 after x_2 iterations of the **loop**.
- $\text{if}(x_2, x_0, x_1)$ computes the new value assigned to x_0 from the old value of x_0 for input x_1 after x_2 iterations by the **if** statement.

Primitive Recursion and Loop Computability



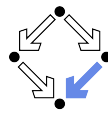
Evaluation of $\min(5) = \text{loop}(5, 5)$.

$$\begin{aligned} \text{loop}(0, 5) &= 5 \\ \text{loop}(1, 5) &= \text{if}(0, \text{loop}(0, 5), 5) = \text{if}(0, 5, 5) = 5 \\ \text{loop}(2, 5) &= \text{if}(1, \text{loop}(1, 5), 5) = \text{if}(1, 5, 5) = 5 \\ \text{loop}(3, 5) &= \text{if}(2, \text{loop}(2, 5), 5) = \text{if}(2, 5, 5) = 5 \\ \text{loop}(4, 5) &= \text{if}(3, \text{loop}(3, 5), 5) = \text{if}(3, 5, 5) = 3 \\ \text{loop}(5, 5) &= \text{if}(4, \text{loop}(4, 5), 5) = \text{if}(4, 3, 5) = 3 \end{aligned}$$

x_0	x_1	x_2
5	5	0
5	5	1
5	5	2
5	5	3
3	5	4
3	5	5

In sequence of evaluations of $\text{loop}(x_2, x_1) = x_0$ the values (x_0, x_1, x_2) correspond to the program trace of the loop program.

Primitive Recursion and Loop Computability



Theorem: every prim. recursive function is loop computable and vice versa.

Proof \Rightarrow : we show that primitive recursive function h is loop computable.

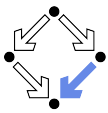
- If h is one of the basic functions, it is clearly loop computable.
- Case $h(x_1, \dots, x_n) := f(g_1(x_1, \dots, x_n), \dots, g_k(x_1, \dots, x_n))$

$$\begin{aligned} y_1 &:= g_1(x_1, \dots, x_n); \\ y_2 &:= g_2(x_1, \dots, x_n); \\ &\dots \\ y_k &:= g_k(x_1, \dots, x_n); \\ x_0 &:= f(y_1, \dots, y_k) \end{aligned}$$

- Case $h(y, x_1 \dots x_n) := \begin{cases} f(x_1, \dots, x_n) & \text{if } y = 0 \\ g(y - 1, h(y, x_1, \dots, x_n), x_1, \dots, x_n) & \text{else} \end{cases}$

$$\begin{aligned} x_0 &:= f(x_1, \dots, x_n); \quad x_y := 0; \\ \text{loop } y \text{ do} \\ &x_0 := g(x_y, x_0, x_1, \dots, x_n); \\ &x_y := x_y + 1 \\ \text{end} \end{aligned}$$

Primitive Recursion and Loop Computability



Proof \Leftarrow : let h be computable by loop program P . Let $f_P : \mathbb{N}^{n+1} \rightarrow \mathbb{N}^{n+1}$ be the function that maps the initial values of the variables used by P to their final values. We show by induction on P that f_P is primitive recursive.

- Case $x_i := k$:

$$f_P(x_0, \dots, x_n) := (x_0, \dots, x_{i-1}, k, x_{i+1}, \dots, x_n)$$

- Case $x_i := x_j \pm 1$:

$$f_P(x_0, \dots, x_n) := (x_0, \dots, x_{i-1}, x_j \pm 1, x_{i+1}, \dots, x_n)$$

- Case $P_1; P_2$:

$$f_P(x_0, \dots, x_n) := f_{P_2}(f_{P_1}(x_0, \dots, x_n))$$

- Case **loop** x_i **do** P' **end**:

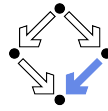
$$f_P(x_0, \dots, x_n) := g(x_i, x_0, \dots, x_n)$$

$$g(0, x_0, \dots, x_n) := (x_0, \dots, x_n)$$

$$g(m+1, x_0, \dots, x_n) := f_{P'}(g(m, x_0, \dots, x_n))$$

Thus the Ackermann function is also not primitive recursive.

μ -Recursive Functions



A partial function over the natural numbers is μ -recursive, if it

- is the constant null, successor, or a projection function,
- can be constructed from other μ -recursive functions by composition or primitive recursion, or
- is a function $h : \mathbb{N}^n \rightarrow_p \mathbb{N}$ defined as

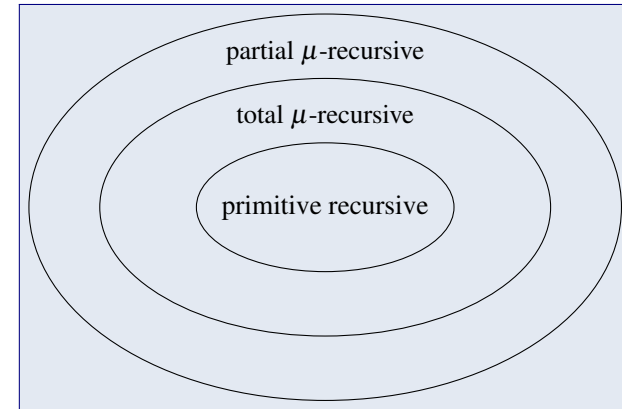
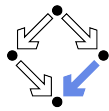
$$h(x_1, \dots, x_n) := (\mu f)(x_1, \dots, x_n)$$

with μ -recursive $f : \mathbb{N}^{n+1} \rightarrow_p \mathbb{N}$ and $(\mu f) : \mathbb{N}^n \rightarrow_p \mathbb{N}$ defined as

$$(\mu f)(x_1, \dots, x_n) := \min \left\{ y \in \mathbb{N} \mid \begin{array}{l} f(y, x_1, \dots, x_n) = 0 \wedge \\ \forall z \leq y : (z, x_1, \dots, x_n) \in \text{domain}(f) \end{array} \right\}$$

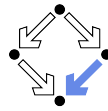
$(\mu f)(x_1, \dots, x_n)$ is the smallest y such that $f(y, x_1, \dots, x_n) = 0$ (and f is defined for all $z \leq y$); the result of h is undefined, if no such y exists.

μ -Recursive Functions



Every primitive recursive function is a total μ -recursive function; a μ -recursive function may or may not be total.

μ -Recursion and While Computability



Theorem: every μ -recursive function is while computable and vice versa.

Proof \Rightarrow : we show that μ -recursive h is while computable.

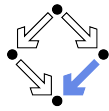
- If h is one of the basic functions or defined by composition or primitive recursion, it is clearly while computable.
- Case $h(x_1, \dots, x_n) := (\mu f)(x_1, \dots, x_n)$

```

x0 := 0;
y := f(x0, x1, ..., xn);
while y do
  x0 := x0 + 1;
  y := f(x0, x1, ..., xn)
end
    
```

μ -recursion denotes unbounded iterative search.

μ -Recursion and While Computability



Proof \Leftarrow : let $h : \mathbb{N}^k \rightarrow_p \mathbb{N}$ be computable by while program P with variables x_0, \dots, x_n . Then $h(x_1, \dots, x_k) := \text{var}_0(f_P(0, x_1, \dots, x_k, 0, \dots, 0))$ where $\text{var}_i(x_0, \dots, x_n) := x_i$. We show that $f_P : \mathbb{N}^{n+1} \rightarrow_p \mathbb{N}^{n+1}$ is μ -recursive by induction on P .

- If P is an assignment, a sequence, or a bounded loop, then f_P is clearly μ -recursive.
- Case **while x_i do P' end**:

$$f_P(x_0, \dots, x_n) := g((\mu g_i)(x_0, \dots, x_n), x_0, \dots, x_n)$$

$$g_i : \mathbb{N}^{n+1} \rightarrow \mathbb{N}$$

$$g_i(m, x_0, \dots, x_n) := \text{var}_i(g(m, x_0, \dots, x_n))$$

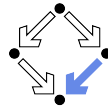
$$g(0, x_0, \dots, x_n) := (x_0, \dots, x_n)$$

$$g(m+1, x_0, \dots, x_n) := f_{P'}(g(m, x_0, \dots, x_n))$$

- $g_i(m, x_0, \dots, x_n)$: the value of program variable i after m iterations
- $g(m, x_0, \dots, x_n)$: the values of all variables after m iterations.

Thus the Ackermann function is also μ -recursive.

Normal Form of a μ -Recursive Function



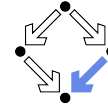
Kleene's Normal Form Theorem: every μ -recursive function h can be defined in Kleene's normal form:

$$h(x_1, \dots, x_k) := f_2(x_1, \dots, x_k, (\mu g)(f_1(x_1, \dots, x_k)))$$

- f_1, f_2, g are **primitive** recursive functions.

A single application of μ is all that is needed.

Normal Form of a μ -Recursive Function



We sketch the proof of Kleene's Normal Form Theorem.

Since h is μ -recursive, it is computable by a while program in normal form

```
x_c := 1; while x_c do ... end
```

with memory function

$$f_P(x_0, \dots, x_n) := g((\mu g_c)(init(x_0, \dots, x_n)), init(x_0, \dots, x_n))$$

with primitive recursive g and g_c and $init(x_0, \dots, x_c, \dots, x_n) := (x_0, \dots, 1, \dots, x_n)$.

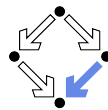
Thus we can define

$$\begin{aligned} h(x_1, \dots, x_k) &:= var_0(f_P(0, x_1, \dots, x_k, 0, \dots, 0)) \\ &= var_0(g((\mu g_c)(init(0, x_1, \dots, x_k, 0, \dots, 0)), init(0, x_1, \dots, x_k, 0, \dots, 0))) \\ &= f_2(x_1, \dots, x_k, (\mu g_c)(f_1(x_1, \dots, x_k))) \end{aligned}$$

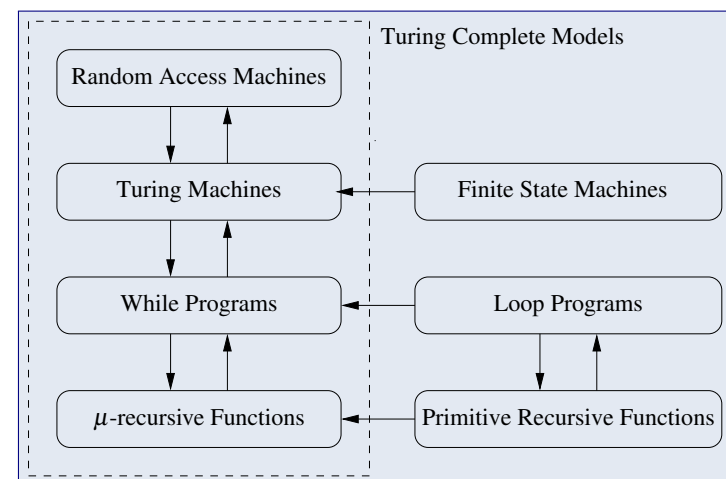
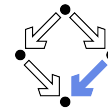
with primitive recursive

$$\begin{aligned} f_1(x_1, \dots, x_k) &:= init(0, x_1, \dots, x_k, 0, \dots, 0) \\ f_2(x_1, \dots, x_k, r) &:= var_0(g(r, init(0, x_1, \dots, x_k, 0, \dots, 0))) \end{aligned}$$

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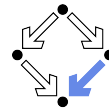


The Big Picture So Far



We are going to sketch some more Turing complete models.

Goto Programs



- A **goto program** has form

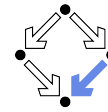
$$L_1 : P_1; L_2 : P_2; \dots; P_n : A_n$$
 where L_k denotes a label and P_k an action:

$$P ::= x_i := 0 \mid x_j := x_j + 1 \mid x_j := x_j - 1 \mid \text{if } x_i \text{ goto } L_j$$
- **Semantics** $\llbracket P \rrbracket(k, m)$:
 - A partial function which maps the initial state (k, m) of P , consisting of program counter $k \in \mathbb{N}$ and memory $m : \mathbb{N} \rightarrow \mathbb{N}$, to its final state (unless the program does not terminate).

$$\begin{aligned} \llbracket P \rrbracket(0, m) &:= m \\ \llbracket P = \dots; P_k : x_i := 0; \dots \rrbracket(k, m) &:= \llbracket P \rrbracket(k+1, m[i \leftarrow 0]) \\ \llbracket P = \dots; P_k : x_i := x_j + 1; \dots \rrbracket(k, m) &:= \llbracket P \rrbracket(k+1, m[i \leftarrow m[j] + 1]) \\ \llbracket P = \dots; P_k : x_i := x_j - 1; \dots \rrbracket(k, m) &:= \llbracket P \rrbracket(k+1, m[i \leftarrow \max\{0, m[j] - 1\}]) \\ \llbracket P = \dots; P_k : \text{if } x_i \text{ goto } L_j; \dots \rrbracket(k, m) &:= \begin{cases} \llbracket P \rrbracket(k+1, m), & \text{if } m(i) = 0 \\ \llbracket P \rrbracket(j, m), & \text{if } m(i) \neq 0 \end{cases} \end{aligned}$$

We have already seen how goto programs can be translated to while programs and vice versa; goto programs are therefore Turing complete.

λ -Calculus



- A **λ -term** T :

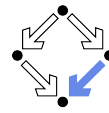
$$T ::= x_i \mid (T T) \mid (\lambda x_i. T)$$
 - x_i : a variable.
 - $(T T)$: an **application**.
 - $(\lambda x_i. T)$: an **abstraction**.
- **Reduction relation** \rightarrow :

$$((\lambda x_i. T_1) T_2) \rightarrow (T_1[x_i \leftarrow T_2])$$
 - The result of the application of a function to an argument.
- **Reduction sequence** $T_1 \rightarrow^* T_2$

$$T_1 \rightarrow \dots \rightarrow T_2$$
 - T_2 is in **normal form**, if no further reduction is possible.
- **Church-Rosser Theorem**: If $T_1 \rightarrow^* T_2$ and $T_1 \rightarrow^* T'_2$ such that both T_2 and T'_2 are in normal form, then $T_2 = T'_2$.

Every computable function can be represented by a λ -term.

λ -Calculus



How can we represent unbounded iteration (recursion)?

- Can define **fixpoint operator** Y :

$$(YF) \rightarrow^* (F(YF))$$
 - $Y := (\lambda f. (\lambda x. (f(xx)))) (\lambda x. (f(xx)))$
- Can translate **recursive function definition to λ -term**:

$$f(x) := \dots f(g(x)) \dots \rightsquigarrow f := YF$$

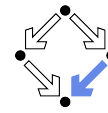
$$F := \lambda h. \lambda x. \dots h(g(x)) \dots$$

- λ -term behaves like recursive function.

$$fa = (YF)a \rightarrow^* F(YF)a \rightarrow^* \dots (YF)(g(a)) \dots = \dots f(g(a)) \dots$$

Formal basis of functional programming languages.

Rewriting Systems



- A **term rewriting system** is a set of rules of form

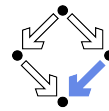
$$L \rightarrow R$$

- L, R : terms such that L is not a variable and every variable that appears in R must also appear in L .
- **Rewriting Step** $T \rightarrow T'$:
 - There is some rule $L \rightarrow R$ and a substitution σ (a mapping of variables to terms) such that
 - some subterm U of T matches the left hand side L of the rule under the substitution σ , i.e., $U = L\sigma$,
 - T' is derived from T by replacing U with $R\sigma$, i.e with the right hand side of the rule after applying the variable replacement.
- **Rewriting Sequence** $T_1 \rightarrow^* T_2$

$$T_1 \rightarrow \dots \rightarrow T_2$$
 - T_2 is in **normal form**, if no further reduction is possible.

Every computable function can be represented by a term rewriting system.

Rewriting Systems



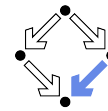
- Term rewriting system:

$$\begin{aligned}f(x, f(y, z)) &\rightarrow f(f(x, y), z) \\ f(x, e) &\rightarrow x \\ f(x, i(x)) &\rightarrow e\end{aligned}$$

- Rewriting sequence:

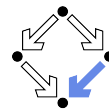
$$\begin{aligned}f(a, f(i(a), e)) &\rightarrow f(f(a, i(a)), e) \rightarrow f(e, e) \rightarrow e \\ f(a, f(i(a), e)) &\rightarrow f(a, i(a)) \rightarrow e\end{aligned}$$

Rewriting systems can be also defined over strings and graphs; the later form the basis of tools for model driven architectures.



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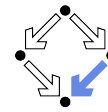
Languages and Machines



- Regular languages:
 - Representable by regular expressions.
 - Recognizable by finite state machines.
- Recursively enumerable languages:
 - Representable by ... ?
 - Recognizable by Turing machines.
- Relationship:
 - Every regular language is recursively enumerable.
 - Every finite state machine can be simulated by a Turing machine.
But not vice versa.

Are there any other interesting classes of languages and associated machine models and how do they relate to those above?

Grammars

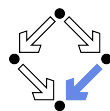


Grammar $G = (N, \Sigma, P, S)$:

- N : a finite set of **nonterminal symbols**,
- Σ : a finite set of **terminal symbols** disjoint from N .
 $N \cap \Sigma = \emptyset$
- P : a finite set of **production rules** of form $l \rightarrow r$ such that
 $l \in (N \cup \Sigma)^* \circ N \circ (N \cup \Sigma)^*$
 $r \in (N \cup \Sigma)^*$
 - l and r consist of nonterminal and/or terminal symbols.
 - l must contain at least one nonterminal symbol.
 - Multiple rules $l \rightarrow r_1, l \rightarrow r_2, \dots, l \rightarrow r_n$ can be abbreviated:
 $l \rightarrow r_1 \mid r_2 \mid \dots \mid r_n$
- S : the **start symbol**.
 $S \in N$

Grammar G describes a language over alphabet Σ .

Grammars and Recursively Enum. Lang.

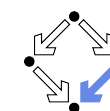


Theorem: The languages of (unrestricted) grammars are exactly the recursively enumerable languages.

- Proof \Rightarrow : construct 2-tape nondeterministic M with $L(M) = L(G)$.
 M uses the second tape to construct some sentence of $L(G)$: it starts by writing S on the tape and then nondeterministically chooses some rule $l \rightarrow r$ and applies it to some occurrence of l on the tape, replacing it by r . Then M checks whether the result equals the word on the first tape. If yes, M accepts the word, otherwise, it continues with another production rule.
- Proof \Leftarrow : construct grammar G with $L(G) = L(M)$.
 Sentential forms encode pairs (w, c) of input w and configuration c of M ; every form contains a non-terminal symbol such that by a rule application the current configuration is replaced by the successor configuration. The rules ensure that
 - from the start symbol, every matching pair (w, c) of M can be derived;
 - for every transition that moves c to c' , a rule is constructed that allows a derivation $(w, c) \Rightarrow (w, c')$;
 - if configuration c describes a final state from which no further transition is possible, the derivation $(w, c) \Rightarrow w$ is possible.

Unrestricted grammars represent another Turing complete model.

The Chomsky Hierarchy



Noam Chomsky, 1959.

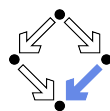
Type i	Grammar $G(i)$	Language $L(i)$	Machine $M(i)$
0	unrestricted	recursively enumerable	Turing machine
1	context-sensitive	context-sensitive	linear bounded automaton
2	context-free	context-free	push down automaton
3	right linear	regular	finite state machine

$L(i)$ is the set of languages of grammars $G(i)$ and machines $M(i)$.

- For $i > 0$, the set of languages of type $L(i)$ is a proper subset of the set of languages $L(i-1)$, i.e. $L(i) \subset L(i-1)$.
- For $i > 0$, every machine in $M(i)$ can be simulated by a machine in $M(i-1)$ (but not vice versa).

Grammars correspond to machine models.

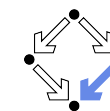
Context-Free Languages (Type 2)



- **Context-free grammar G :** every rule has form $A \rightarrow r$ with $A \in N$.
 - Independent of the context, any occurrence of A can be replaced.
- **Example:** $L := \{a^i b^i \mid i \in \mathbb{N}\}$
 $S \rightarrow \varepsilon \mid aSb$
 $S \Rightarrow aSb \Rightarrow aaSbb \Rightarrow aaaSbbb \Rightarrow aaabbb$
- **Pushdown automaton M :** nondeterministic FSM with unbounded stack of symbols as “working memory”:
 - in every transition $\delta(q, a, b) = (q', w)$,
 - M reads the next input symbol a (a may be ε , i.e., M may not read a symbol) and the symbol b on the top of the stack, and
 - replaces b by a (possibly empty) sequence w of symbols.

Most languages in computer science are context-free.

Generation of Syntax Analyzers



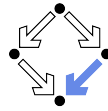
“Compiler generators” for the generation of syntax analyzers (parsers).

- Input: a (deterministic) context free grammar.

```
statement: assignment | conditional | whileloop | ... ;
whileloop: 'while' '(' valexp ')' statement ;
```
- Output: a (deterministic) push down automaton (as a program)

```
public final LoopStatement whileloop() throws ... {
    ...
    pushFollow(FOLLOW_valexp_in_whileloop1457);
    valexp();
    state._fsp--;
    if (state.failed) return value;
    ...
    pushFollow(FOLLOW_statement_in_whileloop1484);
    statement();
    state._fsp--;
    if (state.failed) return value;
    ...
}
```

Context-Sensitive Languages (Type 1)



Context-sensitive grammar G :

- in every rule $l \rightarrow r$, we have $|l| \leq |r|$, i.e., the length of left side l is less than or equal the length of right side r ,
- the rule $S \rightarrow \varepsilon$ is only allowed, if the start symbol S does not appear on the right hand side of any rule.

Example: $L := \{a^i b^j c^i \mid i \in \mathbb{N}\}$

$S \rightarrow \varepsilon \mid T, T \rightarrow ABC \mid TABC$

$BA \rightarrow AB, CB \rightarrow BC, CA \rightarrow AC$

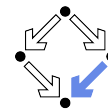
$AB \rightarrow ab, bC \rightarrow bc, Aa \rightarrow aa, bB \rightarrow bb, cC \rightarrow cc$

$S \Rightarrow T \Rightarrow TABC \Rightarrow ABCABC \Rightarrow ABACBC \Rightarrow AABCBC \Rightarrow AABBC$
 $\Rightarrow AabBCC \Rightarrow aabBCC \Rightarrow aabbCC \Rightarrow aabbcC \Rightarrow aabbcc$

- Linear bounded automaton M :** nondeterministic Turing machine with k tapes (for some k).
 - For input of length n , only the first n cells of each tape are used.
 - The “space” used is a fixed multiple of the length of the input word.

Less practical importance.

Summary



We have seen examples of each type of language.

- Type 3:** $\{(ab)^n \mid n \in \mathbb{N}\}$
 - Language is regular.
- Type 2:** $\{a^n b^n \mid n \in \mathbb{N}\}$
 - Language is context-free.
- Type 1:** $\{a^n b^n c^n \mid n \in \mathbb{N}\}$
 - Language is context-sensitive.
- Type 0:** $\{a^i b^j c^k \mid k = ack(i, j)\}$
 - Language is recursively enumerable (also recursive).

None of these languages of type i is also of type $i + 1$.

1. Random Access Machines

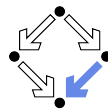
2. Loop and While Programs

3. Primitive Recursive and μ -recursive Functions

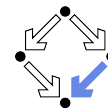
4. Further Turing Complete Models

5. The Chomsky Hierarchy

6. Real Computers



Real Computers



Are real computers Turing complete?

- Hardware view:**
 - Finite number of digital elements and thus a finite number of states.
 - Cannot simulate the infinite Turing machine tape.
 - Cannot perform unbounded arithmetic.
 - A computer is thus a **finite state machine** (i.e., not Turing complete).
View taken by model checkers.
- Algorithm theory view:**
 - On demand, arbitrary much (e.g., virtual) memory may be added.
 - Can thus simulate arbitrary large portion of the Turing machine tape.
 - Can thus perform unbounded arithmetic.
 - A computer is **Turing complete**.
View taken by algorithm design.

A matter of the point of view respectively the goal of the modeling.