7. The Recursion Theorem

- Main result in this section: Kleene's Recursion Theorem.
 - Recursive functions are closed under a very general form of recursion.
- For the proof we will use the **S-m-n-theorem**.
 - Used in many proofs in computability theory.
 - However, both the S-m-n theorem and the proof of the Recursion theorem will be omitted this year.
 Jump to Kleene's Recursion Theorem.

The S-m-n Theorem

$$\begin{split} f &: \mathbb{N}^{m+n} \xrightarrow{\sim} \mathbb{N} \text{ partial rec.} \\ \vec{l} &: \mathbb{N}^m \\ g &: \mathbb{N}^n \xrightarrow{\sim} \mathbb{N} \text{ partial rec.} \\ g(\vec{x}) &\simeq f(\vec{l}, \vec{x}). \end{split}$$

- So there exists a primitive recursive function S_n^m s.t.,
 - if $f = \{e\}^{m+n}$,
 - then $g = {\{S_n^m(e, \vec{l})\}}^n$.
- So $\{S_n^m(e, \vec{l})\}^n(\vec{x}) \simeq \{e\}^{m+n}(\vec{l}, \vec{x}).$

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

The S-m-n Theorem

26 Computability Theory, Michaelmas Term 2008, Sect. 7

- Assume $f : \mathbb{N}^{m+n} \xrightarrow{\sim} \mathbb{N}$ partial recursive.
- Fix the first *m* arguments (say $\vec{l} := l_0, \ldots, l_{m-1}$).
- Then we obtain a partial recursive function

 $g: \mathbb{N}^n \xrightarrow{\sim} \mathbb{N}$, $g(ec{x}) \simeq f(ec{l}, ec{x})$.

- The S-m-n theorem expresses that we can compute a Kleene index of g
 - i.e. an e' s.t. $g = \{e'\}^n$

from a Kleene index of f and \vec{l} primitive recursively.

Notation

$\{{\sf S}_n^m(e,\vec{l})\}^n(\vec{x})\simeq \{e\}^{m+n}(\vec{l},\vec{x}).$

- Assume t is an expression depending on n variables x, s.t. we can compute t from x partial recursively. Then *kxt* is any natural number e s.t. {e}ⁿ(x) ≃ t.
- Then we will have

$$S_n^m(e, \vec{l}) = \lambda \vec{x} \cdot \{e\}^{m+n}(\vec{l}, \vec{x})$$
.

7-1

Theorem 7.1 (S-m-n Theorem)

- **•** Assume $m, n \in \mathbb{N}$.
- There exists a primitive recursive function

$$\mathsf{S}_n^m:\mathbb{N}^{m+1}\to\mathbb{N}$$

s.t. for all $\vec{l} \in \mathbb{N}^m, \vec{x} \in \mathbb{N}^n$

$$\{\mathbf{S}_{n}^{m}(e,\vec{l})\}^{n}(\vec{x}) \simeq \{e\}^{m+n}(\vec{l},\vec{x})$$

Proof of S-m-n Theorem

T is TM for e.

Want to define T' s.t. $T'^{(n)}(\vec{x}) \simeq T^{(n+m)}(\vec{l}, \vec{x})$

T' can be defined as follows:

- 1. The initial configuration is:
 - \vec{x} written on the tape,
 - head pointing to the left most bit:



26 Computability Theory, Michaelmas Term 2008, Sect. 7

7-5

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

Proof of S-m-n Theorem



- 2. T' writes first binary representation of $\vec{l} = l_0, \ldots, l_{n-1}$ in front of this.
 - terminates this step with the head pointing to the most significant bit of $bin(l_0)$.

So configuration after this step is:



Proof of S-m-n Theorem

- Let T be a TM encoded as e.
- A Turing machine T' corresponding to $S_n^m(e, \vec{l})$ should be s.t.

$${\rm T'}^{(n)}(\vec{x}) \simeq {\rm T}^{(n+m)}(\vec{l},\vec{x})$$
 .

Proof of S-m-n Theorem

T is TM for e.

Want to define T' s.t. $T'^{(n)}(\vec{x}) \simeq T^{(n+m)}(\vec{l}, \vec{x})$. Configuration after first step:

 $\operatorname{bin}(l_0)$ \square \cdots \square $\operatorname{bin}(l_{m-1})$ \square $\operatorname{bin}(x_0)$ \square \cdots \square

 Then T' runs T, starting in this configuration. It terminates, if T terminates. The result is

$$\simeq \mathrm{T}^{(m+n)}(\vec{l},\vec{x})$$
,

and we get therefore

$$\mathbf{T'}^{(n)}(\vec{x}) \simeq \mathbf{T}^{(m+n)}(\vec{l},\vec{x})$$

as desired

26 Computability Theory, Michaelmas Term 2008, Sect. 7

7-9

 $bin(x_{n-1})$

Proof of the S-m-n Theorem

T is TM for *e*. T' is a TM s.t. $T'^{(n)}(\vec{x}) \simeq T^{(n+m)}(\vec{l}, \vec{x})$

- From a code for T one can now obtain a code for T' in a primitive recursive way.
- S_n^m is the corresponding function.
- The details will not be given in the lecture Jump to Kleene's Recursion Theorem

Proof of the S-m-n Theorem

- A code for ${\rm T}'$ can be obtained from a code for ${\rm T}$ and from \vec{l} as follows:
 - One takes a Turing machine $\mathrm{T}^{\prime\prime},$ which writes the binary representations of

$$\vec{l} = l_0, \dots, l_{m-1}$$

in front of its initial position (separated by a blank and with a blank at the end), and terminates at the left most bit.

 It's a straightforward exercise to write a code for the instructions of such a Turing machine, depending on *l*, and show that the function defining it is primitive recursive.

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

Proof of the S-m-n Theorem

- Assume, the terminating state of T" has Gödel number (i.e. code) s, and that all other states have Gödel numbers < s.
- Then one appends to the instructions of T" the instructions of T, but with the states shifted, so that the new initial state of T is the final state s of T" (i.e. we add s to all the Gödel numbers of states occurring in T).
- This can be done as well primitive recursively.

Proof of the S-m-n Theorem

• So a code for T'' can be defined primitive recursively depending on a code *e* for T and \vec{l} , and S_n^m is the primitive recursive function computing this. With this function it follows now that, if *e* is a code for a TM, then

 $\{\mathbf{S}_{n}^{m}(e,\vec{l})\}^{n}(\vec{x}) \simeq \{e\}^{n+m}(\vec{l},\vec{x})$.

This equation holds, even if e is not a code for a TM: In this case $\{e\}^{m+n}$ interprets e as if it were the code for a valid TM T

Proof of the S-m-n Theorem

• $e' := S_n^m(e, \vec{l})$ will have the same deficiencies as e, but when applying the Kleene-brackets, it will be interpreted as a TM T' obtained from e' in the same way as we obtained T from e, and therefore

 $\{e'\}^n(\vec{x}) \simeq {T'}^{(n)}(\vec{x}) \simeq {T}^{(n+m)}(\vec{l},\vec{x}) \simeq \{e\}^{n+m}(\vec{l},\vec{x})$.

So we obtain the desired result in this case as well.

26 Computability Theory, Michaelmas Term 2008, Sect. 7

Proof of the S-m-n Theorem

- (A code for such a valid TM is obtained by
 - deleting any instructions encode(q, a, q', a', D) in e
 s.t. there exists an instruction encode(q, a, q", a", D')
 occurring before it in the sequence e,
 - and by replacing all directions > 1 by $\lceil R \rceil = 1$.)

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

Kleene's Recursion Theorem

- Assume $f : \mathbb{N}^{n+1} \xrightarrow{\sim} \mathbb{N}$ partial recursive.
- Then there exists an $e \in \mathbb{N}$ s.t.

 $\{e\}^n(\vec{x}) \simeq f(e, \vec{x})$.

(Here $\vec{x} = x_0, \dots, x_{n-1}$).

7-13

Example 1

- Kleene's Rec. Theorem: $\exists e. \forall \vec{x}. \{e\}^n(\vec{x}) \simeq f(e, \vec{x}).$
- There exists an e s.t.

 $\{e\}(x) \simeq e+1$.

For showing this take in the Recursion Theorem $f(e,n):=e+1. \label{eq:fermion}$ Then

$$\{e\}(x) \simeq f(e, x) \simeq e + 1$$
.

Example 2

- The function computing the Fibonacci-numbers fib is recursive.
 - (This is a weaker result than what we obtained above –
 - above we showed that it is even prim. rec.)

26 Computability Theory, Michaelmas Term 2008, Sect. 7

7-17

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

Remark

Kleene's Rec. Theorem: $\exists e. \forall \vec{x}. \{e\}^n(\vec{x}) \simeq f(e, \vec{x}).$

- Applications as Example 1 are usually not very useful.
- Usually, when using the Rec. Theorem, one
 - doesn't use the index e directly,
 - but only the application of $\{e\}$ to arguments.

Fibonacci Numbers

Remember the defining equations for fib:

$$\begin{split} & \operatorname{fib}(0) &= \operatorname{fib}(1) = 1 \ , \\ & \operatorname{fib}(n+2) &= \operatorname{fib}(n) + \operatorname{fib}(n+1) \ . \end{split}$$

From these equations we obtain

$$\mathsf{fib}(n) = \begin{cases} 1, & \text{if } n = 0 \text{ or } n = 1, \\ \mathsf{fib}(n \doteq 2) + \mathsf{fib}(n \doteq 1), & \text{otherwise.} \end{cases}$$

We show that there exists a recursive function $g:\mathbb{N}\to\mathbb{N},$ s.t.

$$g(n) \simeq \left\{ \begin{array}{ll} 1, & \mbox{if } n=0 \mbox{ or } n=1 \mbox{,} \\ g(n \dot{-} 2) + g(n \dot{-} 1), & \mbox{otherwise.} \end{array} \right.$$

Fibonacci Numbers

Show: Exists *g* rec.

s.t. $g(n) \simeq \begin{cases} 1, & \text{if } n = 0 \text{ or } n = 1, \\ g(n - 2) + g(n - 1), & \text{otherwise.} \end{cases}$ Shown as follows: Define a recursive $f : \mathbb{N}^2 \to \mathbb{N}$ s.t.

$$f(e,n) \simeq \begin{cases} 1, & \text{if } n = 0 \text{ or } n = 1, \\ \{e\}(n \doteq 2) + \{e\}(n \doteq 1), & \text{otherwise.} \end{cases}$$

Now let e be s.t.

$$\{e\}(n)\simeq f(e,n)$$

Then e fulfils the equations

$$\{e\}(n) \simeq \begin{cases} 1, & \text{if } n = 0 \text{ or } n = 1, \\ \{e\}(n - 2) + \{e\}(n - 1), & \text{otherwise.} \end{cases}$$

26 Computability Theory, Michaelmas Term 2008, Sect. 7

7-21

Fibonacci Numbers

$$\{e\}(n) \simeq \begin{cases} 1, & \text{if } n = 0 \text{ or } n = 1, \\ \{e\}(n \div 2) + \{e\}(n \div 1), & \text{otherwise.} \end{cases}$$

Let $g = \{e\}$. Then we get

$$g(n) \simeq \left\{ \begin{array}{ll} 1, & \mbox{if } n=0 \mbox{ or } n=1 \mbox{,} \\ g(n \dot{-} 2) + g(n \dot{-} 1), & \mbox{otherwise.} \end{array} \right.$$

These are the defining equations for fib. One can show by induction on *n* that g(n) = fib(n) for all $n \in \mathbb{N}$.

Therefore fib is recursive.

General Applic. of Rec. Theorem

- Similarly, one can introduce arbitrary partial recursive functions g, where
 - $g(\vec{n})$ refers to arbitrary other values $g(\vec{m})$.
- So, instead of arguing as before that fib is partial recursive, it suffices to say the following
 - By the recursion theorem, there exists a partial recursive function fib : N → N, s.t.

$$\mathsf{fib}(n) \simeq \left\{ \begin{array}{ll} 1, & \text{if } n = 0 \text{ or } n = 1, \\ \mathsf{fib}(n \dot{-} 2) + \mathsf{fib}(n \dot{-} 1), & \text{otherwise.} \end{array} \right.$$

- We can prove by induction on n that $\forall n : \mathbb{N}.\mathsf{fib}(n) \downarrow$ holds.
- Therefore fib is total and therefore recursive.

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

General Applic. of Rec. Theorem

- This use of the the recursion theorem corresponds to the recursive definition of functions in programming.
- E.g. in Java one defines

```
public static int fib(int n){
    if (n == 0 || n == 1){
        return 1;}
    else{
        return fib(n-1) + fib(n-2);
    }
};
```

Example 3

- As in general programming, recursively defined functions need not be total:
- There exists a partial recursive function $g: \mathbb{N} \xrightarrow{\sim} \mathbb{N}$ s.t.

$$g(x)\simeq g(x)+1$$
 .

- We get $g(x)\uparrow$.
- The definition of g corresponds to the following Java definition: public static int g(int n){ return g(n) + 1; };
- When executing g(x), Java loops.

26 Computability Theory, Michaelmas Term 2008, Sect. 7

7-25

$\mathsf{Ack}(0,y) \ = \ y+1 \ ,$

Ackermann Function

The Ackermann function is recursive:

Remember the defining equations:

From this we obtain

$$\mathsf{Ack}(x,y) = \left\{ \begin{array}{ll} y+1, & \text{if } x=0, \\ \mathsf{Ack}(x \doteq 1,1), & \text{if } x>0 \text{ and } y=0, \\ \mathsf{Ack}(x \doteq 1,\mathsf{Ack}(x,y \doteq 1)), & \text{otherwise.} \end{array} \right.$$

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

Example 4

• There exists a partial recursive function $g: \mathbb{N} \xrightarrow{\sim} \mathbb{N}$ s.t.

 $g(x) \simeq g(x+1) + 1$.

Note that that's a "black hole recursion", which is not solvable by a total function.

- It is solved by $g(x)\uparrow$.
- Note that a recursion equation for a function f cannot always be solved by setting $f(x)\uparrow$.
 - E.g. the recursion equation for fib can't be solved by setting fib(n)↑.

Ackermann Function

$$\mathsf{Ack}(x,y) = \left\{ \begin{array}{ll} y+1, & \text{if } x=0, \\ \mathsf{Ack}(x \doteq 1,1), & \text{if } x>0 \text{ and } y=0, \\ \mathsf{Ack}(x \doteq 1,\mathsf{Ack}(x,y \doteq 1)), & \text{otherwise.} \end{array} \right.$$

Define g partial recursive s.t.

$$g(x,y) \simeq \begin{cases} y+1, & \text{if } x = 0, \\ g(x - 1, 1), & \text{if } x > 0 \land y = 0, \\ g(x - 1, g(x, y - 1)), & \text{if } x > 0 \land y > 0. \end{cases}$$

- g fulfils the defining equations of Ack.
- Proof that $g(x, y) \simeq Ack(x, y)$ follows by main induction on x, side-induction on y. The details will not be given in the lecture. Jump over remaining slides.

26 Computability Theory, Michaelmas Term 2008, Sect. 7

Proof of Correctness of Ack

- We show by induction on x that g(x, y) is defined and equal to Ack(x, y) for all $x, y \in \mathbb{N}$:
 - Base case x = 0.

$$g(0,y)=y+1=\mathsf{Ack}(0,y)$$
 .

• Induction Step $x \rightarrow x + 1$. Assume

$$g(x,y) = \operatorname{Ack}(x,y)$$
 .

We show

$$g(x+1,y) = \mathsf{Ack}(x+1,y)$$

by side-induction on y:

26 Computability Theory, Michaelmas Term 2008, Sect. 7

7-29

7-30

Proof of Correctness of Ack

Show g(x + 1, y) = Ack(x + 1, y)

• Base case y = 0:

$$g(x+1,0) \simeq g(x,1) \stackrel{\mbox{Main-IH}}{=} {\rm Ack}(x,1) = {\rm Ack}(x+1,0) \ . \label{eq:gamma}$$

• Induction Step $y \rightarrow y + 1$:

$$\begin{array}{rcl} g(x+1,y+1) &\simeq & g(x,g(x+1,y)) \\ & \underset{\simeq}{\text{Main-IH}} \\ & g(x,\operatorname{Ack}(x+1,y)) \\ & \underset{\simeq}{\text{Side-IH}} \\ & \cong & \operatorname{Ack}(x,\operatorname{Ack}(x+1,y)) \\ & = & \operatorname{Ack}(x+1,y+1) \ . \end{array}$$

Jump over remaining slides (Proof of the Recursion Theorem)

Idea of Proof of the Rec. Theorem

Assume

$$f: \mathbb{N}^{n+1} \xrightarrow{\sim} \mathbb{N}$$

We have to find an e s.t.

$$\forall \vec{x} \in \mathbb{N}. \{e\}^n(\vec{x}) \simeq f(e, \vec{x})$$

- We set $e = \lambda \vec{x} \cdot \{e_1\}^{n+1}(e_1, \vec{x})$ for some e_1 to be determined.
- Then the left and right hand side of the equation of the recursion theorem reads

$$\{e\}^{n}(\vec{x}) \simeq \{\lambda \vec{x}. \{e_{1}\}^{n+1}(e_{1}, \vec{x})\}^{n}(\vec{x}) \\ \simeq \{e_{1}\}^{n+1}(e_{1}, \vec{x}) \\ f(e, \vec{x}) \simeq f(\lambda \vec{x}. \{e_{1}\}^{n+1}(e_{1}, \vec{x}), \vec{x})$$

CS_226 Computability Theory, Michaelmas Term 2008, Sect. 7

Idea Proof of Rec. Theorem

We need to satisfy $\forall \vec{x} \in \mathbb{N}. \{e\}^n(\vec{x}) \simeq f(e, \vec{x}).$ Let $e = \lambda \vec{x}. \{e_1\}^{n+1}(e_1, \vec{x}).$ $\{e\}^n(\vec{x}) \simeq \{e_1\}^{n+1}(e_1, \vec{x}) ,$ $f(e, \vec{x}) \simeq f(\lambda \vec{x}. \{e_1\}^{n+1}(e_1, \vec{x}), \vec{x}) .$

• So e_1 needs to fulfill the following equation:

$$[e_1]^{n+1}(e_1, \vec{x}) \simeq \{e\}^n(\vec{x})$$

$$\stackrel{!}{\simeq} f(e, \vec{x})$$

$$\simeq f(\lambda \vec{x} \cdot \{e_1\}^{n+1}(e_1, \vec{x}), \vec{x})$$

• This can be fulfilled if we define e_1 s.t.

 $\{e_1\}^{n+1}(e_2, \vec{x}) \simeq f(\lambda \vec{x}. \{e_2\}^{n+1}(e_2, \vec{x}), \vec{x})$

Idea of Proof of Rec. Theorem

 $\{e_1\}^{n+1}(e_2, \vec{x}) \simeq f(\lambda \vec{x}. \{e_2\}^{n+1}(e_2, \vec{x}), \vec{x}).$

• By the S-m-n Theorem we can obtain this if we have e_1 s.t.

 $\{e_1\}^{n+1}(e_2, \vec{x}) \simeq f(\mathsf{S}_n^1(e_2, e_2), \vec{x})$

• There exists a partial recursive function $g: \mathbb{N}^n + 1 \xrightarrow{\sim} \mathbb{N}$, s.t.

 $g(e_2, \vec{x}) \simeq f(\mathsf{S}_n^1(e_2, e_2), \vec{x})$

• If e_1 is an index for g we obtain the desired equation.

$$\{e_1\}^{n+1}(e_2, \vec{x}) \simeq f(\mathsf{S}_n^1(e_2, e_2), \vec{x})$$

26 Computability Theory, Michaelmas Term 2008, Sect. 7

7-33

Complete Proof of Rec. Theorem

Let e_1 be s.t.

$$\{e_1\}^{n+1}(y, \vec{x}) \simeq f(\mathsf{S}^1_n(y, y), \vec{x})$$

Let $e := S_n^1(e_1, e_1)$. Then we have

{

$$e \}^{n}(\vec{x}) \qquad e = \mathsf{S}_{n}^{1}(e_{1}, e_{1}) \qquad \{\mathsf{S}_{n}^{1}(e_{1}, e_{1})\}^{n}(\vec{x})$$

S-m-n theorem
$$e_{1} \{e_{1}\}^{n+1}(e_{1}, \vec{x})$$

Def of $e_{1} \qquad f(\mathsf{S}_{n}^{1}(e_{1}, e_{1}), \vec{x})$
$$e = \mathsf{S}_{n}^{1}(e_{1}, e_{1}) \qquad f(e, \vec{x}) .$$

26 Computability Theory, Michaelmas Term 2008, Sect. 7