
Some heuristics for deriving invariant
LN sections 6.7 , 6.8

Overview

- Heuristic 1: “Replace constant with counter” heuristic
- Heuristic 2: incrementally building the invariant
- Heuristic 3: tail invariant
- Other: an example of using invariant to do data refinement

Loop Reduction rule

$P \Rightarrow I$

$\{ * g \wedge I * \} \quad S \quad \{ * I * \}$

$I \wedge \neg g \Rightarrow Q$

$\{ * I \wedge g * \} \quad C := m; S \quad \{ * m < C * \}$

$I \wedge g \Rightarrow m > 0$

$\{ * P * \} \quad \mathbf{while} \ g \ \mathbf{do} \ S \quad \{ * Q * \}$

// setting up I (Init)

// invariance (IC)

// exit cond (EC)

// m decreasing (TC1)

// m bounded below (TC2)

Heuristic “replace constant with counter”

$\{^* 0 \leq n \ ^*\}$

```
i := 0 ; r := true ;  
while i < n do {  
    r := r ∧ (a[i]=0) ; i++ }
```

$\{^* r = (\forall k : 0 \leq k < n : a[k]=0) \ ^*\}$

(1) The loop iterates on the counter up to some upper bound or lower bound.

- Identify this upper/lower bound. Suppose this is **n**.
- Identify the counter. Suppose this is **i**.

(2) If the post-cond is of the form **Q(n)**, we try **Q(i)** ∧ “range” as the invariant, where “range” is a formula stating the valid range of the counter **i** during the loop. Notice with this choice of invariant you get $Q(i) \wedge i=n \Rightarrow Q(n)$ for free.

Heuristic “replace constant with counter”

$\{^* 0 \leq n \text{ }^*\}$

$i := 0 ; r := \text{true} ;$

while $i < n$ **do** {
 $r := r \wedge (a[i]=0) ; i++$ }

$\{^* r = (\forall k : 0 \leq k < n : a[k]=0) \text{ }^*\}$



$(r = (\forall k : 0 \leq k < i : a[k]=0)) \wedge 0 \leq i \leq n$

(we already did this example, now you know what inspired the choice of this invariant)

Let's take an example, now with multi vars

- Consider this well known recursive function to calculate the n^{th} Fibonacci number:

fib 0 = 0

fib 1 = 1

fib (n + 2) = **fib** (n+1) + **fib** n

- The definition is nice but it has bad execution time. Next, we will consider an alternative, more efficient, implementation of the calculation.

Fibonacci's original problem

- Fibonacci's original problem is to calculate the number of rabbits given a certain growth model :
 1. We start with the first month with one new pair of rabbits.
 2. A new pair needs 1 month to become adult
 3. Each month each pair of adult rabbits produces a new pair.

We assume no rabbit dies during the experiment.

- This model can be directly mapped to an imperative implementation (next slide).

Iterative Fib, with $O(n)$ runtime

```
RabitSim (n:int) : int {  
  nNow,nAdult,newPairs,t : int ;  
  t:=1 ; nAdult:=0 ; nNow:=1 ;  
  while t < n do {  
    newPairs := nAdult ;  
    nAdult   := nNow ;  
    nNow     := nNow + newPairs ;  
    t := t + 1  
  } ;  
  return nNow }
```

pre : $n > 0$

post : $\text{return} = \text{fib}(n)$

Reducing the spec. to the statement level

$\{ * n > 0 * \}$

```
t:=1 ; nAdult:=0 ; nNow:=1 ;  
while t < n do {  
    newPairs := nAdult ;  
    nAdult   := nNow ;  
    nNow     := nNow + newPairs ;  
    t := t + 1  
};  
return := nNow
```

$\{ * \text{return} = \text{fib } n * \}$

After some wp calculation

$\{ * n > 0 * \}$

```
t:=1 ; nAdult:=0 ; nNow:=1 ;  
while t < n do {  
    newPairs := nAdult ;  
    nAdult   := nNow ;  
    nNow     := nNow + newPairs ;  
    t := t + 1  
};
```

$\{ * nNow = fib\ n \ * \}$

Applying the previous heuristic

$\{ * n > 0 * \}$

```
t:=1 ; nAdult:=0 ; nNow:=1 ;  
 $\{ * I * \}$   
while t < n do {  
  newPairs := nAdult ;  
  nAdult   := nNow ;  
  nNow     := nNow + newPairs ;  
  t := t + 1 } ;
```

$\{ * nNow = fib\ n * \}$

Inv : $nNow = fib\ t$
 $\wedge 1 \leq t \leq n$
term. metric : $n-t$

Notice that with this choice of invariant, the exit condition " $I \wedge \neg g \Rightarrow Q$ " is quite trivially satisfied.

Let's take a look at the PIC part

- “PIC” : proof of the invariance condition. That is to prove that $\{^* I \wedge g ^*\} S \{^* I ^*\}$ holds, where S is the loop's body. In other words, to prove that $I \wedge g \Rightarrow \mathbf{wp} S I$ is valid.
- Calculating $\mathbf{wp} S I$ gives:

$$nNow + nAdult = \text{fib}(t+1) \quad \wedge \quad 1 \leq t+1 \leq n$$

- The proof structure is as follows:

PROOF PIC

[A1] $nNow = \text{fib } t$

[A2] $1 \leq t \leq n$

[A3] $t < n$

[G1] $nNow + nAdult = \text{fib}(t+1)$

[G2] $1 \leq t+1 \leq n$

The proof of G1

PROOF EQUATIONAL

$$\begin{aligned} & nNow + nAdult \\ &= \{ A1 \} \\ & \quad fib\ t + nAdult \\ & \quad \times = \{ ??\ \text{cannot find the justification} \} \\ & \quad fib\ t + fib\ (t-1) \\ &= \{ \text{def. of fib and } t \geq 1\ (A2) \} \\ & \quad fib\ (t+1) \end{aligned}$$

The proof fails because the invariant has no information about $nAdult$.
Let's extend it then, with the proper information. The new invariant:

$$nNow = fib\ t \wedge nAdult = fib\ (t-1) \wedge 1 \leq t \leq n$$

In general, for every variable involved in a loop, that contributes to its post-condition, we will need to capture its property in the invariant.

The new PIC

- Since you change the invariant, you will have to rework your proofs to reflect the change. For PIC, since you change the invariant by strengthening it with a new conjunct, this means that we can now assume more, though you also have to prove more.
- Re-calculating **wp** S I gives:

$$\begin{aligned} & \text{nNow} + \text{nAdult} = \text{fib } (t+1) \\ & \wedge \text{nNow} = \text{fib } t \\ & \wedge \quad 1 \leq t+1 \leq n \end{aligned}$$

The new PIC

PROOF PIC

[A1] $nNow = fib\ t$

[A1b] $nAdult = fib\ (t-1)$

[A2] $1 \leq t \leq n$

[A3] $t < n$

[G1] $nNow + nAdult = fib\ (t+1)$

[G1b] $nNow = fib\ t$

[G2] $1 \leq t+1 \leq n$

BEGIN

1. { see the subproof below } G1



2. { A1 } G1b

3. { follows from A2 and A3 } G2

END

PROOF EQUATIONAL

$nNow + nAdult$
= { A1 }
 $fib\ t + nAdult$
= { A1b }
 $fib\ t + fib\ (t-1)$
= { def. of fib and $t \geq 1$ (A2) }
 $fib\ (t+1)$

Tail invariant

- Consider again this previous example:

$\{^* 0 \leq n \ ^*\}$

```
i := 0 ; r := true ;  
while i < n do {  
    r := r ∧ (a[i]=0) ; i++ }
```

$\{^* r = (\forall k : 0 \leq k < n : a[k]=0) \ ^*\}$

- We proved this with this as the invariant:

$r = (\forall k : 0 \leq k < i : a[k]=0) \ \wedge \ 0 \leq i \leq n$

Tail invariant

- The invariant we used:

$$r = (\forall k : 0 \leq k < i : a[k]=0) \quad \wedge \quad 0 \leq i \leq n$$

capturing the work that **has been done**

- An alternative invariant :

capturing the work that **still to be done**

$$r \wedge (\forall k : i \leq k < n : a[k]=0) = (\forall k : 0 \leq k < n : a[k]=0) \\ \wedge \quad 0 \leq i \leq n$$

A “tail invariant” expresses the invariant in terms of the remaining work that is still to be done. The loop works on shrinking this to-be-done part until it disappears, or become small enough it can be computed directly without a loop.

Example: iterative impl. of tail recursion

- Example of “typical recursion” :

g x = **if** x ≤ 0 **then** x **else** 1+**g**(x-1)

- A tail recursive function does not combine the result of its recursion:

f x = **if** x ≤ 0 **then** x **else** **f**(x-1)

Examples

- Get the last element of a list:

```
last [x]      = x  
last (x:y:s) = last (y:s)
```

- Summing the elements of a list:

```
lsum a []      = a  
lsum a (x:s)   = lsum (x+a) s
```

- Reversing a list (more efficient in Haskell) :

```
rv t []      = t  
rv t (x:s)   = rv (x:t) s
```

Tail Recursion, general form

- A tail recursive function $F:A \rightarrow B$ has this general form:

$$\begin{array}{rcl} F\ x & = & g\ x \rightarrow \text{base } x \\ & & | \quad F\ (\Delta\ x) \end{array} \quad \begin{array}{l} // \text{ base case} \\ // \text{ recursion} \end{array}$$

- Such a function can be optimized by implementing the recursion as loop-iterations, as the latter does not use stack space to pass around parameters.
- **Termination.** F terminates if we can find a function $m:A \rightarrow \text{Int}$ such that:

$$\begin{array}{ll} 1. \neg g\ x & \Rightarrow m\ (\Delta\ x) < m\ x \\ 2. \neg g\ x & \Rightarrow m\ x > 0 \end{array}$$

Iterative implementation

- **Problem:** given α , calculate $F \alpha$. Recall the def. of F :

$$F x = \begin{array}{l} g x \rightarrow \text{base } x \\ | \quad F (\Delta x) \end{array}$$

- Consider this simple loop to calculate $F \alpha$:

{* true *}

$x := \alpha ;$

while $\neg g x$ **do** $x := \Delta x ;$

$r := \text{base } x$

{* $r = F \alpha$ *}

To prove that this works, we will use the following tail invariant:

$$F x = F \alpha$$

For termination we use $m x$ as the termination metric where m is the function you used to prove the termination of F (prev. slide)

Proof sketch

$\{^* \text{ true } ^*\}$

$x := \alpha ;$

while $\neg g\ x$ **do** $x := \Delta x ;$

$r := \text{base } x$

Invariant: $F\ x = F\ \alpha$

$\{^* r = F\ \alpha ^*\}$

- When the loop terminates, $g\ x$ holds. So by its definition $F\ x = \text{base } x$. Then the invariant implies that $\text{base } x = F\ \alpha$.
- For the proof of invariance: **wp** $(x := \Delta x)$ inv gives $F\ (\Delta x) = F\ \alpha$. But since $\neg g\ x$ holds, $F\ (\Delta x)$ is also equal to $F\ x$, by the def. of F . So the wp can be reduced to $F\ x = F\ \alpha$, which is exactly the invariant itself.

Corollary

If a problem can be expressed as computing the result of a tail recursive function, the previous implementation scheme, using the previously given invariant, provides a correct iterative program to compute the solution of the problem.

Example : GCD

- Given $X, Y > 0$, compute $\text{gcd } X \ Y$.
- Straight forward solution: $O(X \text{ min } Y)$.
- We can do better than that (Euclides, 320 BC)
- Approach: try to cast the problem into tail recursion.
- We will need to explore some properties of common divisors.

Some properties of gcd

- We'll assume **positive integers**! Define :

$$d \text{ divides } x \quad = \quad \exists k : k > 0 : x = k * d$$

$$d \text{ comdiv } x, y \quad = \quad d \text{ divides } x \wedge d \text{ divides } y$$

- Theorem. If $x > y$ we have:

$$d \text{ comdiv } x, y \quad \Leftrightarrow \quad d \text{ comdiv } (x-y), y$$

\\ hmm ... looks like a “split”

Some properties of gcd

- We can characterize **gcd** via this equation :

$$\alpha = \mathbf{gcd} \ x \ y = \alpha \text{ comdiv } x,y \\ \wedge \\ (\forall e : e \text{ comdiv } x,y : e \leq \alpha)$$

- Theorem-1. for positive integers x and y, if $x > y$ we have:
 $\mathbf{gcd} \ x \ y = \mathbf{gcd} \ (x-y) \ y$
- Theorem-2. for positive integer x: $\mathbf{gcd} \ x \ x = x$

Gcd in tail recursion

- Those properties of **gcd** leads to the following recursive relation for **gcd** . Notice that the function is tail recursive:

$$\begin{array}{lcl} \text{gcd } x \ y & = & \begin{array}{l} x = y \rightarrow x \\ | \quad x > y \rightarrow \text{gcd } (x-y) \ y \\ | \quad /* \ x < y \ */ \text{gcd } x \ (y - x) \end{array} \end{array}$$

- Consider the problem of calculating **gcd** X Y, given some positive integers X and Y. The previous implementation scheme gives us a solution for this.

Iterative implementation of Gcd

- Recall the recursive definition:

$$\begin{array}{lcl} \text{gcd } x \ y & = & x = y \rightarrow x \\ & | & x > y \rightarrow \text{gcd } (x-y) \ y \\ & | & /* \ x < y \ */ \ \text{gcd } x \ (y-x) \end{array}$$

- Problem:** given positive integers X, Y calculate $\text{gcd } X \ Y$.
Implemented by this loop:

$\{ * \ X > 0 \ \wedge \ Y > 0 \ * \}$

$x, y := X, Y ;$

while $x \neq y$ **do**

if $x > y$ **then** $x := x - y$
 else $y := y - x$

$\{ * \ x = \text{gcd } X \ Y \ * \}$

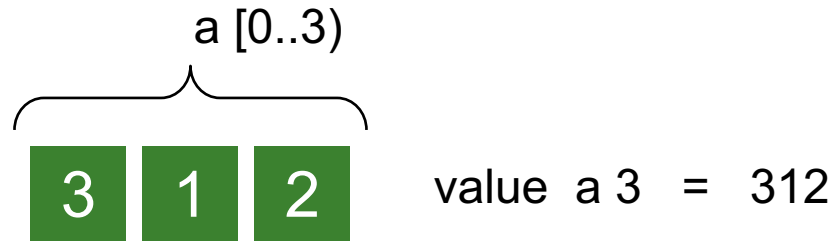
inv : $\text{gcd } x \ y = \text{gcd } X \ Y$

term. metric : $x + y$

Note: to prove that this termination metric has a lower bound 0, we need to extend the invariant with $x + y \geq 0$

Another example

- Consider a list of non-negative integers representing a number.



Write a program that computes the “value” of the number represented by the list.

- A recursive solution:

```
val x s = s=[] → x
           | val (10*x + hd s) (tail s)
```

The value of a list *s* can be calculated by **val** 0 *s*.

Iterative implementation

- Recall again the tail recursive function `val`:

```
val x s = s=[] → x  
        | val (10*x + hd s) (tail s)
```

- Problem:** given an `S`, calculate `val 0 S`. Iterative impl.:

```
{* true *
```

```
x , s := 0 , S ;  
while s≠[] do {  
  x := 10*x + hd s ;  
  s := tail s }
```

```
{* x = val 0 S *
```

```
inv : val x s = val 0 S
```

```
term. metric : #s
```

Data refinement, an example

- Consider again the previous program:

```
{* true *
```

```
x , s := 0 , S ;  
while s≠[] do {  
    x := 10*x + hd s ;  
    s := tail s }  
}
```

```
{* x = val 0 S *
```

- What if the input list “S” is actually an array a[0..n) ? You might foresee that hd s and tail s can then be implemented more efficiently by simply shifting a cursor/counter into the array. How to justify the correctness of such a transformation?

Transformation 1

The same program, but with “S” instantiated to $a[0..n]$:

$\{ * 0 \leq n * \}$

$x, s := 0, a[0..n]$

$i := 0$

while $s \neq []$ **do** {
 $x := 10 * x + \text{hd } s$
 $s := \text{tail } s$
 $i := i + 1$

}

$\{ * x = \text{val } 0 (a[0..n]) * \}$

$\text{inv}_1 : \text{val } x \text{ } s = \text{val } 0 (a[0..n])$

We will introduce a new variable i and program it so that we can maintain the following “data invariant” that captures the relation between the two data structures (the array a and the list s):

$\text{inv}_2 : s = a[i..n) \wedge 0 \leq i \leq n$

Notice that the new code does not change the behavior of the base program. So it does not break its correctness argument. This is also called “superposition” of a new variable (in this example: i).

Transformation 2

$\{^* 0 \leq n \ ^*\}$

$x, s := 0, a[0..n)$

$i := 0$

while $s \neq []$ **do** {
 $x := 10 * x + \text{hd } s$
 $s := \text{tail } s$
 $i := i + 1$
}

$\{^* x = \text{val } 0 (a[0..n]) \ ^*\}$

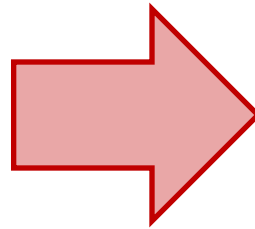
$\{^* 0 \leq n \ ^*\}$

$x, s := 0, a[0..n)$

$i := 0$

while $i \neq n$ **do** {
 $x := 10 * x + a[i]$
 $s := \text{tail } s$
 $i := i + 1$
}

$\{^* x = \text{val } 0 (a[0..n]) \ ^*\}$



inv₁: $\text{val } x \ s = \text{val } 0 (a[0..n))$

inv₂: $s = a[i..n) \wedge 0 \leq i \leq n$

The data invariant implies that $s \neq []$ in the original program is equivalent to $i \neq n$, and can thus be replaced by the latter. Similarly, $\text{hd } s$ can be replaced with $a[i]$.

Transformation 3

$\{^* 0 \leq n \ ^*\}$

~~$x, s := 0, a[0..n]$~~

$i := 0$

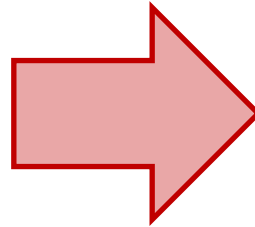
while $i \neq n$ **do** {
 $x := 10 * x + a[i]$

~~$s := \text{tail } s$~~

$i := i + 1$

}

$\{^* x = \text{val } 0 (a[0..n]) \ ^*\}$



$\{^* 0 \leq n \ ^*\}$

$x := 0$

$i := 0$

while $i \neq n$ **do** {
 $x := 10 * x + a[i]$

$i := i + 1$

}

$\{^* x = \text{val } 0 (a[0..n]) \ ^*\}$

Finally, notice that we don't actually need s anymore (it has no influence towards the result x), so we can just as well drop it.