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## The correctness problem

Programs are written to perform some particular task.

However, it is often very hard to write a program that performs its intended function — as programmers know well.

In practice, most large programs have 'bugs'.

Some bugs are harmless, others merely irritating.

They can cause financial and public relations disasters (e.g. the Pentium FDIV bug).

In some situation bugs can be deadly.

Peter Neumann: 'Computer Related Risks'.



# Testing and verification

One good way to track down bugs is through extensive testing.

But usually there are too many possible situations to try them all exhaustively, so there may still be bugs lying undetected.

Program testing can be very useful for demonstrating the presence of bugs, but it is only in a few unusual cases where it can demonstrate their absence.

An alternative is *verification*, where we try to *prove* that a program behaves as required. Consider ordinary mathematical theorems, like

$$\Sigma_{n=0}^{n=N} n = \frac{N(N+1)}{2}$$

We can *test* this for many particular values of N, but it is easier and more satisfactory simply to *prove* it (e.g. by induction).





It is only the central link that is mathematically precise. The others are still informal — all we can do is try to keep them small.

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# Verifying functional programs

We suggested earlier that functional programs might be easier to reason about formally, because they correspond directly to the mathematical functions that they represent.

This is arguable, but at least we will try to show that reasoning about some simple functional programs is straightforward.

We need to remember that, in general, functional programs are *partial* functions. Sometimes we need a separate argument to establish termination.

Often, the proofs proceed by induction, parallelling the definition of the functions involved by recursion.



Recall the following simple definition of natural number exponentiation:

```
- fun exp x n =
if n = 0 then 1
else x * exp x (n - 1);
```

We will prove that this satisfies the following specification:

For all  $n \ge 0$  and x, exp x n terminates and exp x  $n = x^n$ 

The function is defined by (primitive) recursion.

The proof is by (step-by-step, mathematical) induction.



## Exponentiation (2)

- If n = 0, then by definition exp x n = 1. Since for any integer x, we have x<sup>0</sup> = 1, so the desired fact is established.
- Suppose we know  $\exp x \ n = x^n$ . Because  $n \ge 0$ , we also know  $n + 1 \ne 0$ . Therefore:

$$\exp x (n+1) = x * \exp x ((n+1) - 1)$$
$$= x * \exp x n$$
$$= x * x^{n}$$
$$= x^{n+1}$$

Q.E.D.

Note that we assume  $0^0 = 1$ , an example of how one must state the specification precisely!

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For any integers x and y, gcd x y terminates and returns a gcd of x and y.

Here we need to be even more careful about the specification. What is a gcd of two negative numbers?

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Greatest common divisor (2)

We write x|y, pronounced 'x divides y', to mean that y is an integral multiple of x, i.e. there is some integer d with y = dx.

We say that d is a common divisor of x and y if d|x and d|y.

We say that d is a *greatest* common divisor if:

- We have d|x and d|y
- For any other integer d', if d'|x and d'|y then d'|d.

Note that unless x and y are both zero, we do not specify the sign of the gcd. The specification does not constrain the implementation completely.

#### Greatest common divisor (3)

Now we come to the proof. The gcd function is no longer defined by *primitive* recursion.

In fact, gcd x y is defined in terms of gcd y (x mod y) in the step case.

We do not, therefore, proceed by step-by-step mathematical induction, but by *wellfounded* induction on |y|.

The idea is that this quantity (often called a *measure*) decreases with each call. We can use it to prove termination, and as a handle for wellfounded induction.

In complicated recursions, finding the right wellfounded ordering on the arguments can be tricky. But in many cases one can use this simple 'measure' approach.

### Greatest common divisor (4)

Now we come to the proof. Fix some arbitrary n. We suppose that the theorem is established for all arguments x and y with |y| < n, and we try to prove it for all x and y with |y| = n. There are two cases.

First, suppose that y = 0. Then gcd  $x \ y = x$  by definition. Now trivially x|x and x|0, so it is a common divisor.

Suppose d is another common divisor, i.e. d|x and d|0. Then immediately we get d|x, so x is a greatest common divisor.

This establishes the first part of the induction proof.

#### Greatest common divisor (5)

Now suppose  $y \neq 0$ . We want to apply the inductive hypothesis to gcd  $y \pmod{y}$ .

We will write  $r = x \mod y$  for short. The basic property of the mod function that we use is that, since  $y \neq 0$ , for some integer q we have x = qy + rand |r| < |y|.

Since |r| < |y|, the inductive hypothesis tells us that  $d = \gcd y \ (x \mod y)$  is a gcd of y and r.

We just need to show that it is a gcd of x and y. It is certainly a common divisor, since if d|y and d|r we have d|x, as x = qy + r.

Now suppose d'|x and d'|y. By the same equation, we find that d'|r. Thus d' is a common divisor of y and r, but then by the inductive hypothesis, d'|d as required.

Append (1)

Now consider an example concerning lists rather than numbers. Define:

- fun append [] l = l
  - | append (h::t) l = h::(append t l);

This is supposed to join together two lists. We want to prove that the operation is associative, i.e. for any three lists  $l_1$ ,  $l_2$  and  $l_3$  we have:

append  $l_1$  (append  $l_2$   $l_3$ ) = append (append  $l_1$   $l_2$ )  $l_3$ 

We can proceed by induction on the length of  $l_1$ , but since the function was defined by structural recursion over lists, it is more natural to prove the theorem by *structural induction*.

The principle is: if a property holds for the empty list, and whenever it holds for t it holds for any h::t, then it holds for any list.



We proceed, then, by structural induction on  $l_1$ . There are two cases to consider. First, suppose  $l_1 = []$ . Then we have:

append  $l_1$  (append  $l_2$   $l_3$ )

- = append [] (append  $l_2 l_3$ )
- = append  $l_2 l_3$
- = append (append []  $l_2$ )  $l_3$
- = append (append  $l_1 l_2$ )  $l_3$

As required.

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append t (append  $l_2 l_3$ ) = append (append t  $l_2$ )  $l_3$ Therefore:

append  $l_1$  (append  $l_2$   $l_3$ )

- = append (h :: t) (append  $l_2 l_3$ )
- $= h :: (append t (append l_2 l_3))$
- $= h :: (append (append t l_2) l_3)$
- = append  $(h :: (append t l_2)) l_3)$
- = append (append  $(h :: t) l_2) l_3$ )
- = append (append  $l_1 l_2$ )  $l_3$ )

The theorem is proved.



For a final example, let us define a function to reverse a list:

 $\operatorname{rev}(\operatorname{rev}\,l)=l$ 

This is again a structural induction. However we require two lemmas, which can also be proved by structural induction:

append l [] = lrev(append  $l_1 l_2$ ) = append (rev  $l_2$ ) (rev  $l_1$ )





First suppose that l = []. Then the proof is easy:

$$rev(rev l) = rev(rev [])$$
$$= rev []$$
$$= []$$
$$= l$$

Now suppose that l = h :: t and we know that

$$\operatorname{rev}(\operatorname{rev} t) = t$$

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rev(rev l)

- = rev(rev (h :: t))
- = rev(append (rev t) [h])
- = append (rev [h]) (rev(rev t))
- = append (rev [h]) t
- = append (rev (h :: [])) t
- = append (append [] [h]) t
- = append [h] t
- = append (h :: []) t
- = h :: (append [] t)

$$= h :: t$$

$$= l$$

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Harder cases



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