Induction, Recursion, Algorithmic Analysis

Problem 1

Prove by induction that

$$1 \cdot 1! + 2 \cdot 2! + \dots + n \cdot n! = (n+1)! - 1$$
 for $n \ge 1$

Solution

Let P(n) be the proposition that $1 \cdot 1! + 2 \cdot 2! + ... + n \cdot n! = (n+1)! - 1$. We will prove that P(n) holds for all $n \ge 1$ by induction on n.

Base case n = 1. 1.1! = 1 = 2! - 1 = (1+1)! - 1 so P(1) holds.

Inductive case. Assume P(k) holds for some $k \in \mathbb{N}_{>0}$. That is $1 \cdot 1! + 2 \cdot 2! + \ldots + k \cdot k! = (k+1)! - 1$. Then

$$1 \cdot 1! + 2 \cdot 2! + \dots + k \cdot k! + (k+1) \cdot (k+1)! = (k+1)! - 1 + (k+1) \cdot (k+1)!$$
 (Induction hypothesis)
= $(1+k+1)(k+1)! - 1$
= $((k+1)+1)(k+1)! - 1$

so P(k+1) holds.

Therefore, by the Principle of Induction, P(n) holds for all $n \ge 1$.

Problem 2

Let $\Sigma = \{1, 2, 3\}$.

- (a) Give a recursive definition for the function sum : $\Sigma^* \to \mathbb{N}$ which, when given a word over Σ returns the sum of the digits. For example $\mathsf{sum}(1232) = 8$, $\mathsf{sum}(222) = 6$, and $\mathsf{sum}(1) = 1$. You should assume $\mathsf{sum}(\lambda) = 0$.
- (b) For $w \in \Sigma^*$, let P(w) be the proposition that for all words $v \in \Sigma^*$, sum(wv) = sum(w) + sum(v). Prove that P(w) holds for all $w \in \Sigma^*$.
- (c) Consder the function rev : $\Sigma^* \to \Sigma^*$ defined recursively as follows:
 - $rev(\lambda) = \lambda$
 - For $w \in \Sigma^*$ and $a \in \Sigma$, rev(aw) = rev(w)a

Prove that for all words $w \in \Sigma^*$, sum(rev(w)) = sum(w)

Solution

(a) We give a definition using the recursive nature of Σ^* :

$$\begin{array}{rcl} \operatorname{sum}(\lambda) & = & 0 \\ \operatorname{sum}(a.w) & = & a + \sum (w). \end{array}$$

(b) We first need the recursive definition of concatenation:

$$\lambda.v = v$$

$$(aw).v = a(w.v)$$

We will now prove P(w) for all $w \in \Sigma^*$ by structural induction on w.

Base case ($w = \lambda$).

$$\begin{array}{ll} \operatorname{sum}(wv) &= \operatorname{sum}(\lambda.v) \\ &= \operatorname{sum}(v) & \operatorname{Definition of concatenation} \\ &= 0 + \operatorname{sum}(v) \\ &= \operatorname{sum}(\lambda) + \operatorname{sum}(v) & \operatorname{Definition of sum} \\ &= \operatorname{sum}(w) + \operatorname{sum}(v) \end{array}$$

So $P(\lambda)$ holds.

Inductive case (w = aw'**).** Assume P(w') holds, that is for all $v \in \Sigma^*$, sum(w'v) = sum(w') + sum(v). Then for all $v \in \Sigma^*$ and all $a \in \Sigma$:

$$\begin{array}{ll} \operatorname{sum}((aw')v) &= \operatorname{sum}(a(w'v)) & \operatorname{Definition of concatenation} \\ &= a + \operatorname{sum}(w'v) & \operatorname{Definition of sum} \\ &= a + \operatorname{sum}(w') + \operatorname{sum}(v) & \operatorname{Inductive hypothesis} \\ &= \operatorname{sum}(aw') + \operatorname{sum}(v) & \operatorname{Definition of sum} \end{array}$$

So P(w') implies P(aw') for all $a \in \Sigma$.

Therefore, by the Principle of Structural Induction, P(w) holds for all $w \in \Sigma^*$.

(c) Let P(w) be the proposition that sum(rev(w)) = sum(w). We will show that P(w) holds for all words $w \in \Sigma^*$ by structural induction on w.

Base case ($w = \lambda$ **).** From the definition of rev we have: sum(rev(λ)) = sum(λ). So $P(\lambda)$ holds.

Inductive case (w = aw'**).** Suppose P(w') holds, that is sum(rev(w')) = sum(w'). For any $a \in \Sigma$ we have:

$$\begin{array}{ll} \operatorname{sum}(\operatorname{rev}(aw')) & = \operatorname{sum}(w'a) & \operatorname{Definition\ of\ rev} \\ & = \operatorname{sum}(w') + \operatorname{sum}(a) & \operatorname{From\ }(b) \\ & = \operatorname{sum}(w') + a + \operatorname{sum}(\lambda) & \operatorname{Definition\ of\ sum} \\ & = a + \operatorname{sum}(w') + 0 & \operatorname{Definition\ of\ sum} \\ & = \operatorname{sum}(aw') & \operatorname{Definition\ of\ sum} \end{array}$$

So P(w') implies P(aw') for all $a \in \Sigma$.

Therefore, by the Principle of Structural Induction, P(w) holds for all $w \in \Sigma^*$.

Problem 3

Define $f: \mathbb{N} \times \mathbb{N} \to \mathbb{N}$ recursively as follows: f(m,0) = 0 for all $m \in \mathbb{N}$ and f(m,n+1) = m + f(m,n).

- (a) Let P(n) be the proposition that f(0,n) = f(n,0). Prove that P(n) holds for all $n \in \mathbb{N}$.
- *(b) Let Q(m) be the proposition $\forall n, f(m,n) = f(n,m)$. Prove that Q(m) holds for all $m \in \mathbb{N}$.

Solution

1. We show that P(n) holds for all $n \in \mathbb{N}$ by induction.

Base case: n = 0. Since f(0,0) = f(0,0), P(0) holds.

Inductive case. Now suppose P(n) holds. Then

$$f(0, n + 1) = 0 + f(0, n)$$
 (Def)
= 0 + f(n, 0) (IH)
= 0 (Def)
= $f(n + 1, 0)$.

So $P(n) \to P(n+1)$, and thus P(n) holds for all $n \in \mathbb{N}$.

2. We will prove by induction that f(m,n) = mn, from which it follows that f(m,n) = mn = nm = f(n,m). Let R(n) be the proposition that: for all m, f(m,n) = mn.

Base case: n = 0. From the definition of f, f(m, 0) = 0 = 0.m for all m. So R(0) holds.

Inductive case. Suppose that R(n) holds. That is, for all m, f(m,n) = mn. Then, for all m,

$$f(m, n+1) = m + f(m, n)$$
 Definition of f
= $m + mn$ Induction hypothesis
= $m(n+1)$.

So R(n+1) holds. Thus, R(n) implies R(n+1), so by the Principle of Induction f(m,n) = mn for all m and n. Therefore f(m,n) = f(n,m).

Problem 4

Analyse the complexity of the following algorithms to compute the *n*-th Fibonacci number

(a) **FibOne**(*n*):

if
$$n \le 2$$
 then return 1 else return **FibOne** $(n-1)$ + **FibOne** $(n-2)$

(b) **FibTwo**(*n*):

$$x = 1, y = 0, i = 1$$
While $i < n$:
$$t = x$$

$$x = x + y$$

$$y = t$$

$$i = i + 1$$
return x

Solution

(a) Let T(n) be the running time of **FibOne**(n). Then in the worst case, there are two recursive calls to smaller instances of **FibOne**, taking time T(n-1) and T(n-2) respectively. All other operations are constant time, so

$$T(n) = O(1) + T(n-1) + T(n-2)$$

 $\leq O(1) + 2.T(n-1).$

From the lectures, this means that $T(n) \in O(2^n)$.

(b) Let T(n) be the running time of **FibTwo**(n). We have a while-loop which runs O(n) times, and within the while loop there are several operations taking O(1) time. All other operations are constant time, so the overall running time is $O(1) + O(n) \times O(1) = O(n)$.

Discussion

NB: It is possible to obtain better bounds for **FibOne**, however because of the O(1) that appears in the recurrence equation, it is not quite as simple as T(n) = Fib(n). A bound of $O(2^n)$ demonstrates a reasonable level of understanding, so would be sufficient in most assessable tasks.

Problem 5

Analyse the complexity of the following recursive algorithm to test whether a number x occurs in an *ordered* list $L = [x_1, x_2, ..., x_n]$ of size n. Take the cost to be the number of list element comparison operations.

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BinarySearch(x, L = [x_1, x_2, ..., x_n]):

if n = 0 then return no
else

if x_{\lceil \frac{n}{2} \rceil} > x then return BinarySearch(x, [x_1, ..., x_{\lceil \frac{n}{2} \rceil - 1}])
else if x_{\lceil \frac{n}{2} \rceil} < x return BinarySearch(x, [x_{\lceil \frac{n}{2} \rceil + 1}, ..., x_n])
else return yes
```

Solution

Let T(n) be the cost of running **BinarySearch** on a list of length n. In the worst case, we make 2 = O(1) element comparisons and recursively call **BinarySearch** on a list of length $\lceil \frac{n}{2} \rceil$. So we have:

$$T(n) = O(1) + T(n/2).$$

The Master Theorem applies to this recurrence: we have a = 1, b = 2, c = 0 and $d = \log_b(a) = 0$, so we are in Case 2. This tells us that $T(n) \in O(n^d \log n) = O(\log n)$.