

Sections 3.2: Recurrence Relations

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Abstract

Recurrence relations are defined recursively, and solutions can sometimes be given in “closed-form” (that is, without recourse to the recursive definition). We will solve one type of linear recurrence relation to give a general closed-form solution, the solution being verified by induction.

We’ll be getting some practice with summation notation in this section. (Have you seen it before?)

1 Solving Recurrence Relations

Vocabulary:

- **linear recurrence relation:** $S(n)$ depends linearly on previous $S(r)$, $r < n$:

$$S(n) = f_1(n)S(n-1) + \cdots + f_k(n)S(n-k) + g(n)$$

That means no powers on $S(r)$, or any other functions operating on $S(r)$. The relation is called **homogeneous** if $g(n) = 0$. (Both Fibonacci numbers and factorials are defined by homogeneous linear recurrence relations.)

- **first-order:** $S(n)$ depends only on $S(n-1)$, and not previous terms. (Factorials are first-order, while Fibonacci are second-order, depending on the two previous terms.)
- **constant coefficient:** In the linear recurrence relation, when the coefficients of previous terms are constants. (Fibonacci are constant coefficient; factorials are not.)
- **closed-form solution:** $S(n)$ is given by a formula which is simply a function of n , rather than a recursive definition of itself. (Fibonacci numbers have a closed-form solution; I would say factorials, not so much....)

The author suggests an “expand, guess, verify” method for solving recurrence relations.

$$! : 0! = 1$$

$$n! = n \cdot (n-1)!$$

$$S(n) = f_1(n) S(n-1)$$

$$F: F(1) = 1$$

$$F(2) = 1$$

$$F(n) = F(n-1) + F(n-2)$$

Both are
homogeneous:
no $g(n)$

2nd order -
requires two
preceding
values.

Example: The story of T

(a) Practice 1, p. 159 (from the previous section):

$$T(1) = 1$$

$$T(n) = T(n-1) + 3, \text{ for } n \geq 2$$

closed form solution.

(b) Practice 9, p. 168: Here is the recurrence relation for Example 11, p. 130, in lisp:

```
(defun Tee(n)
  (if (integerp n)
      (cond
        ((>= n 2)
         (+ (Tee (- n 1)) 3))
        ((= n 1)
         1)
        (t (error "Tilt! Only positive ints allowed in function tee...")))
      (error "Tilt! Only positive integers allowed in function tee..."))
  )
> (tee 2)
4
> (mapcar #'tee (iseq 1 10))
(1 4 7 10 13 16 19 22 25 28)
```

$P(n): T(n) = 3(n-1) + 1$

$P(1): T(1) = 3(1-1) + 1 = 1 \checkmark$

Assume $P(k)$, consider $P(k+1)$:

$T(k+1) = 3(k+1-1) + 1$. Consider the LHS!

$T(k+1) = T(k) + 3$ (from RR)

$= 3(k-1) + 1 + 3$ (by assumption)

$= 3[(k-1) + 1] + 1$

(c) Practice 11, p. 181: Find a closed-form solution for the recurrence relation for sequence T of part (a).

Example: general linear first-order recurrence relations with constant coefficients.

$$S(1) = a$$

$$S(n) = cS(n-1) + g(n), \quad n \in \{2, 3, 4, \dots\}$$

“Expand, guess, verify” (then prove by induction!):

$$S(n) = c^{n-1}S(1) + \sum_{i=2}^n c^{n-i}g(i) \quad (1)$$

Now check that this formula works for $T(n)$ from above.

$= 3[(k+1)-1] + 1 \checkmark$

(the RHS).

2 Counting Using Recurrence Relations

Algorithm *BinarySearch* (which is discussed in the previous section) is recursive: it calls itself. Starting from a list of length n it makes one comparison and then calls itself with a list of half its initial length. Hence the number of comparisons for the list of length n , $C(n)$, would be (in the worst case)

$$C(n) = C(\text{floor}(n/2)) + 1 :$$

that is, you'd need to check the middle element, then do a binary search of the sorted list to the left or right, of half the length (or so) of the original list. For a list of length 1, we have our base case: $C(1) = 1$.

That floor function in the inductive step is a pain, but is necessary since n may be odd.

Forgetting the floor for the moment, use the “expand, guess, and verify” approach: in the worst-case scenario, the algorithm will find the element (or not) on its last check (when it's down to a list of length 1).

$$C(n) = C(n/2) + 1 = (C(n/4) + 1) + 1 = ((C(n/8) + 1) + 1) + 1 = \dots$$

Obviously this is only going to work easily (in the sense that $C(n/8)$, etc., make sense) if n is a power of 2. Assume therefore that $n = 2^m$, for integer m . This allows us to throw away the floor function, and makes all quotients reasonable.

Before we begin, can you guess how many comparisons we make in the worst case, for $C(n)$ when $n = 2^m$?

Let's consider a change of variable. First of all, we replace n by 2^m :

$$C(2^m) = C(2^m/2) + 1 = C(2^{m-1}) + 1.$$

Then we define $T(m) = C(2^m)$ (think of T as a composition of functions, $C(x)$ and 2^x); hence

$$T(m) = T(m-1) + 1$$

Note that $T(0) = C(1) = 1$. We can solve easily to get a closed-form solution:

$$T(m) = m + 1$$

Let's now re-express that in terms of C and n . Since $n = 2^m$, we can equally well write $m = \log_2(n)$. Hence, $C(n) = C(2^m) = T(m) = m + 1 = \log_2(n) + 1$. This compares quite favorably with the worst-case estimate from *SequentialSearch*, which would be n (linear in n).

(For those of you who've forgotten, the log function grows much more slowly than a linear function does.)

Let's look at the general recurrence relation of the “divide and conquer” variety: given

$$\begin{aligned} S(1) &= a \\ S(n) &= cS(n/2) + g(n) \end{aligned} \quad (2)$$

Handwritten notes illustrating the recurrence relation for $C(n)$ when n is a power of 2.

Top diagram: $n = 5$ (with a squiggle over the 5). Below it, a sequence of numbers: $1, 3, 7, 9, 11$, each underlined. An arrow points from the 7 to the text $8:7$.

Bottom diagram: A sequence of numbers: $1, 3, 7, 9$, each underlined. An arrow points from the 7 to the text $8:3$.

$$T(m-1) = C(2^{m-1})$$

$$n = 2^m \Leftrightarrow m = \log_2 n$$

when $n=1$
 $m=0$

$$\begin{aligned} C(n) &= C(2^m) \\ &= T(m) \\ &= m + 1 \end{aligned}$$

$$C(n) = \log_2 n + 1$$

Assume $n = 2^m$ for some integer m . Then

~~$$S(2^0) = a$$~~

$$S(2^m) = cS(2^{m-1}) + g(2^m)$$

Now let's perform a change of variables: let $T(m) = S(2^m)$, so that

$$T(0) = a$$

$$T(m) = cT(m-1) + g(2^m)$$

Using formula (1) (formula 8, p. 183), we get

$$T(m) = c^{m-1}T(1) + \sum_{i=2}^m c^{m-i}g(2^i)$$

Then reindexing, since we start with 0 rather than 1, we get

$$T(m) = c^mT(0) + \sum_{i=1}^m c^{m-i}g(2^i)$$

Finally, substituting back in S and n , we get

$$S(n) = c^{\log_2 n}a + \sum_{i=1}^{\log_2 n} c^{\log_2 n - i}g(2^i)$$

Whew! This is the general solution for the divide-and-conquer algorithm of type (2).

Example: Exercise #46, p. 202

$$S(1) = 3$$

$$S(n) = S(\frac{n}{2}) + n \quad \text{for } n \geq 2, n = 2^m$$

$$S(2^m) = cS(2^{m-1}) + g(2^m)$$

$$T(m) = cT(m-1) + g(2^m)$$

$$T(1) = cT(0) + g(2^1)$$

$$T(m) = c^{m-1}(cT(0) + g(2^1))$$

$$+ \sum_{i=2}^m c^{m-i}g(2^i)$$

$$= c^mT(0) +$$

$$c^{m-1}g(2^1) +$$

$$\sum_{i=2}^m c^{m-i}g(2^i)$$

$$= c^mT(0) +$$

$$\sum_{i=1}^m c^{m-i}g(2^i)$$