

Week 5 tute solutions:

Q1) Quick select average case

Theorem: Randomised Pivot Selection

Using randomised pivot selection, the Quickselect algorithm has expected run time $O(N)$.

Proof

Assume without loss of generality that the array contains no duplicate elements and denote by T_N the time taken by Quickselect to select the k 'th order statistic of an array of size N . If the pivot selected is the i 'th order statistic of the array, then the time taken by Quickselect will be

$$T_N = N + 1 + \begin{cases} T(N - i) & \text{if } i < k, \\ 1 & \text{if } i = k, \\ T(i - 1) & \text{if } i > k \end{cases}.$$

Averaging out over all possible pivot selections, we obtain an expected run time of

$$T_N = N + 1 + \frac{1}{N} \left(\sum_{i=1}^{k-1} T(N - i) + \sum_{i=k+1}^N T(i - 1) + 1 \right).$$

Multiplying both sides by N , we obtain

$$NT_N = N^2 + N + \sum_{i=1}^{k-1} T(N - i) + \sum_{i=k+1}^N T(i - 1) + \frac{1}{N}.$$

Substituting $N = N - 1$, we find the equation

$$(N - 1)T_{N-1} = (N - 1)^2 + (N - 1) + \sum_{i=2}^k T(N - i) + \sum_{i=k+1}^{N+1} T(i - 1) + \frac{1}{N - 1}.$$

Subtracting this from the earlier equation yields

$$NT_N - (N - 1)T_{N-1} = N^2 + N - (N - 1)^2 - (N - 1) - T(N - 1) + T(N - 1) + \frac{1}{N} - \frac{1}{N - 1}.$$

Cleaning up, we obtain

$$NT_N = (N - 1)T_{N-1} + 2N + \frac{1}{N(N - 1)}.$$

Divide through by N and we find

$$T_N = \frac{N - 1}{N} T_{N-1} + 2 + \frac{1}{N^2(N - 1)}.$$

Finally, we use the facts that $(N - 1)/N < 1$ and $1/(N^2(N - 1)) < 1$ to state the bound

$$T_N < T_{N-1} + 3.$$

Recalling Lecture 2, the solution to this recurrence is the linear equation, hence we have established that the expected run time of Quickselect with randomised pivoting is linear.

$$T_N < 3N = O(N).$$

Q2.

What is the computational complexity of this algorithm? Attempt to prove it formally!

$$T_1 = 1$$

$$T_2 = 1$$

$$T_n = T_{n-1} + T_{n-2}$$

$$T_n < 2 * T_{n-1}, \text{ given that } T_{n-1} > T_{n-2}$$

$$T_n < 2^2 * T_{n-2}$$

$$T_n < 2^3 * T_{n-3}$$

...

$$T_n < 2^N * 1$$

Hence the complexity of this algorithm is bounded by $O(2^n)$

Can you write a more efficient version that is NOT iterative, but instead single-recursive (rather than double-recursive as in the version above)?

```
def fib(n):
    return _fib(n, 0, 1)
```

```
def _fib(n, a, b):
    if n == 0:
        return b
    return _fib(n-1, b, a+b)
```

What is the time complexity of such a single-recursive implementation?

$O(n)$

Q3) How many times is body() executed for the following values?

Lo1=1, Hi1=10, Lo2=i, Hi2=10 --- 55

Lo1=0, Hi1= 9, Lo2=0, Hi2=9 --- 100

Lo1=1, Hi1=n, Lo2=i-2, Hi2=i+2 --- 5n

Lo1=0, Hi1=n, Lo2=i, Hi2=2*i --- See below.

$$T_0 = 1$$

$$T_1 = T_0 + 2 = 3$$

$$T_2 = T_1 + 3 = 6$$

$$T_3 = T_2 + 4 = 10$$

$$T_3 = T_1 + 3 + 4$$

$$T_3 = T_0 + 2 + 3 + 4$$

$$T_3 = 1 + 2 + 3 + 4$$

Just a sum of natural numbers to $N + 1$, so

$$T_N = ((N + 1)(N + 1 + 1)) / 2$$

$$T_N = (N^2 + 3N + 2) / 2$$

Q4)

Given an integer n , $\text{body}()$ will be run n times where $n > 0$, and $\text{base}()$ will always run once. function $r(n)$, given an integer of $n > 0$ will make a single call to $\text{body}()$ before recursively calling itself with the value of $n-1$. As our n value shrinks by 1 every time our function is recursed, n will be greater than zero on n occasions, calling $\text{body}()$ on n occasions. No matter what value n our function r is given (even if a negative number), the "else" condition will yield true on exactly one occasion.

Thus,

$r(10)$, $\text{body}() = 10$, $\text{base}() = 1$

$r(5)$, $\text{body}() = 5$, $\text{base}() = 1$

$r(1)$, $\text{body}() = 1$, $\text{base}() = 1$

Q5)

If M_i is the minimum of coins to form the value $= i$, using the denomination set $\{D_1, \dots, D_N\}$.

Initialize $M_0 = 0$

DP recurrence:

$M_i = \min_{1 \leq j \leq N} (M_{i-D_j} + 1)$ for all $1 \leq i \leq V$