

1. (a) There are two steps.
  - i. Perform a linear scan to find the index into the array that contains the (time, name) pair with the correct name, say (\*,N).  
*Running time:  $O(n)$*
  - ii. Bubble item (\*,N) up to the root by repeatedly interchanging with its parent node, and then delete the root node using the same recipe as for `deleteMin`.  
*Running time:  $O(\log n)$*

The overall running time is  $O(n)$ .

- (b) Same as part (a), except modify step i. to search based on the time. The running time is still  $O(n)$ .
  - (c) Apply step ii. from part (a). The running time is now  $O(\log n)$ .
2. (a) `i := 0;`  
 (b) `while i < n do`  
 (c)     `if A[i]=i then`  
 (d)         `i := i+1`  
 (e)     `else`  
 (f)         `A[A[i]],A[i] := A[i],A[A[i]]`  
 (g)     `fi`  
 (h) `od`

At the beginning of each loop iteration  $A[0], A[1], \dots, A[i-1] = 0, 1, \dots, i-1$ . Each iteration of the loop either increments  $i$  or increases by at least one the number of array elements that are in correct position. Thus, the number of iterations is between  $n$  and  $2n$ , giving a running time of  $\Theta(n)$ .

*More details (not required for full marks):* The only possible modifications made to  $A$  are to swap two elements. Thus,  $A[0 \dots n-1]$  remains a permutation of  $0 \dots n-1$  throughout execution.

The first time the loop iterates we have  $i = 0$ , so the loop invariant

$$A[0], A[1], \dots, A[i-1] = 0, 1, \dots, i-1 \quad (1)$$

holds trivially. Induction on the number of times the loop iterates shows that (1) holds throughout execution. On the one hand, if line (d) is executed, then  $A[i] = i$  and upon incrementing  $i$  invariant (1) still holds. On the other hand, considering invariant (1), line (f) is executed only if  $A[i] > i$  and thus  $A[A[i]] \neq A[i]$ . But then upon completion of line (f) we have  $A[A[i]] = A[i]$ , thus increasing by at least one the number of array elements in correct position.

3. There are three steps.
  - i. Perform an in place heapify so that  $A[0 \dots n-1]$  is a min-heap.  
*Running time:  $O(n)$*

- ii. Perform  $k$  **deleteMin** operations to obtain the  $k$  smallest integers in the array. Similar to heapsort, for  $i = 1, 2, \dots, k$ , place the  $i$ th integer extracted from the heap into position  $A[n - i]$  of the array. Now  $A[n - 1], A[n - 2], \dots, A[n - k]$  are the  $k$  smallest integers in increasing order.  
*Running time:*  $O(k \log n)$ , which is  $O(n)$  if  $k \in O(n / \log n)$ .
- iii. Reverse the order of elements in the array by interchanging  $A[i]$  and  $A[n - i - 1]$  for  $i = 0, 1, \dots, \lfloor (n - 1) / 2 \rfloor$ .  
*Running time:*  $O(n)$ .
4. (a) Each coin is genuine or counterfeit, but the two cases where all coins are genuine or all are counterfeit are excluded; the total number of possible answers is thus  $2^n - 2$ . Each weighing has exactly 3 possible outcomes, so if an algorithm performs at most  $k$  weighings, the number of different answers the algorithm could return is at most  $3^k$ . So we must have  $3^k \geq 2^n - 2$ ; solving for the integer  $k$  yields the lower bound of  $\lceil \log_3(2^n - 2) \rceil$  weighings.
- (b) Using the algorithm from part (c), we need 3 weighings when  $n = 4$ .  
 Notice that  $\lceil \log_3(2^4 - 2) \rceil = \lceil \log_3(14) \rceil = \lceil 2.402173503 \rceil = 3$ .
- (c) Perform exactly  $n - 1$  weighings between the pairs of coins  $C_1$  and  $C_i$  for  $2 \leq i \leq n$ .  $C_1$  is counterfeit if and only if all weighings determine that  $C_1$  weighs at most as much as the any of the other coins, with  $C_1$  weighing less than at least one other coin since there is at least one genuine coin; the weak players are those that weigh the same as  $C_1$  and the genuine coins are those that weigh more than  $C_1$ .  
 The case where  $C_1$  is a genuine coin is similar.  
 This algorithm is  $\Theta(n)$ . Observe that  $\lceil \log_3(2^n - 2) \rceil > \log_3(2^{n-1})$  for  $n \geq 2$ , and this is equal to  $\log_3(2) \cdot (n - 1)$ , so the lower bound from part (a) is  $\Omega(n)$ . Therefore is algorithm is asymptotically optimal in the number of contests.
5. (a) Choose  $i$  uniformly and randomly from the range  $0, \dots, n - 1$  and return  $A[i]$ . This will be the dominating element with probability at least  $p = (\lfloor n/2 \rfloor + 1)/n > 1/2$ .
- (b) Use the algorithm from part a) to find an element  $x$  which will be the dominating element with probability at least  $p$ . Perform a linear scan of the array to assay if  $x$  occurs at least  $\lfloor n/2 \rfloor + 1$  times; if so then report yes, and if no repeat by calling anew the algorithm from part a).
- The expected-case running time is  $O(n)$  because each iteration has cost  $O(n)$ , and the expected number of iterations is bounded by  $\sum_{i=1}^{\infty} i(1 - p)^{i-1}p = 1/p < 2$ .  
 An alternative analysis would be to observe that, with probability at least  $1/2$ ,  $T(n) \leq cn$  for some constant  $c > 0$ , and otherwise  $T(n) \leq cn + T(n)$ . So the expected cost is given by  $T(n) \leq cn + \frac{1}{2}T(n)$ , which solves to  $T(n) \leq 2cn$ .