1. a)

Big-Omega notation is the lower bound for the running time of an algorithm. If an algorithm g(n) is big-Omega of f(n), this means that there exists constants c > 0 and $n_0 >= 0$ such that |f(n)| >= c|g(n)| for all $n >= n_0$.

t = 0for i=0 to n: for j=0 to n: for k=0 to n: t = t + 1

b) We find the upper bound:

$$\frac{2^{\frac{1}{3}}}{3^{\frac{1}{3}}} \le \frac{2^{\frac{1}{3}}}{3^{\frac{1}{3}}} \le \frac{2^{\frac{1}{3}}}}{3^{\frac{1}{3}}} \le \frac{2^{\frac{1}{3}}}{3^{\frac{1}{3}}} \le \frac{2^{\frac{1}{3}}}}{3^{\frac{1}{3}}} \le \frac{$$

We find the lower bound:

$$\frac{2}{2} \frac{1}{3^{2}} \ge \frac{2}{3^{2}} \frac{1}{3^{2}} = \frac{1}{2} \left(3 - \frac{1}{3^{n}} \right) C \Omega (1)$$

The order of magnitude is \Theta(1).

c)

There exists c1, c2, n1, n2 such that $f(x) \le c1h(x)$ for $n \ge n1$ and $c2g(x) \le h(x)$ for $n \ge n2$. Hence we have

 $af(x) + bf(x) \le (c1a + c2b)h(x)$ for max(n1, n2) Let c3 = c1a + c2b. Then we have found a constant c3 where af(x) + bf(x) <= c3h(x) for max(n1, n2).

9)

We have that $2^n \le 4^n$ for $n \ge 0$. Hence we have that 2^n is $O(4^n)$.

Suppose that 4^n is $O(2^n)$. Then there exists constants c and n0 such that $4^n <= c2^n$ for n >= n0. Then for n >= n0, we have $2^n <= c$. This is a contradiction, since there is no constant where this is true for all n greater than some constant n0. Hence 4^n is not $O(2^n)$.

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A) O(log(log n))

B) O(n)

- C) O(log n)
- D) O(log n)
- E) 2^(h+1) 1
- F) Expected = O(log n) Worst = O(n)
- G) Expected = O(log n) Worst = O(log n)

$$T(n) = 2T(Nz) + \frac{n}{h} = 2T(Nz) + n^2$$

There are two recursive calls, each splitting the array in half. These take T(n/2) time. The last while loop starts at x=n, and each step decrements x by 1/n until it reaches 0, hence it takes $n/(1/n)=n^2$ time.

We prove that $T(n) \le 2n^2-n$ by induction

$$T(n) = 9T(n/3) + n \le 9(\frac{2n^2}{n} - \frac{n}{3}) + n$$

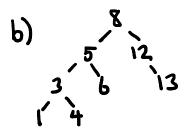
= $2n^2 - 2n \le 2n^2 - n$

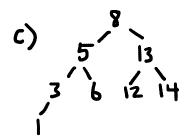
- 4. a) i. 5432169
 - ii. 3621459
 - The optimal ordering of the elements is to order them in non-increasing order of p(e). We prove the contrapositive: if we don't put the items in this order, then there is a different ordering with a smaller expected number of comparisons. Assume we have keys k0,...kn-1, with index(ki) = i, and assume the order of keys is not non-increasing. This means we have p(ki-1) < p(k) for some index 0 < i < n. We define a new order by exchanging ki and ki-1. If we subtract the expected cost of the two orderings, we get p(ki-1) p(k) < 0, hence the new order is more optimal.

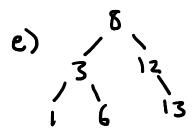
b) The array-based structure uses 15*4 = 60 bytes. The linked structure uses 7*4+7*2*4 = 84 bytes.

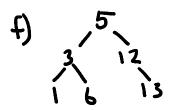
The array-based structure uses less space, since it doesn't need to store references.

- It doesn't matter since they are the same asymptotically.
- **6.** Condition code = balance factor?









- 7. We did not do this lol
- **8.** It takes O(1 + n/m) time, since it takes O(1) time to find the key, and O(average bucket size) time to find the key in the array. The average bucket size is n/m.
 - **b** We did not do coalesced chaining lol