```
CPSC320 Assignment 2
|Name: Evan Louie |
|CSID: m6d7
|SID:
       72210099
______
1.
======
 The population of the coins follow a binomial distribution; meaning each coin has two
 possiblilities in state:
   i) Real
   ii) Fake
 This means that for n coins their exists 2n possible answers in which a single coin is a fake
 and 2n+1 possibe answers in which the possiblity that all are real exist.
 We can also show this in terms of a decision tree in which, given n coins, each coin has the
 option to be lighter or heavier. Because their are n leaves (coins) which can be lighter, or
 n leaves to be heavier or one case in which they are all real; The amount of leaves is 2n+1
 Because the decision tree is a ternary tree with 3 possiblities at each branch (heavier,
 lighter, equal), the height of the tree, h, can be used to calculate the maximum number of
 leaves the tree has by evaluating 3<sup>h</sup> (by definition of a ternary tree).
 This means that 3^h \ge 2n+1
 If we solve for h:
   h \ge log(3, 2n+1)
   h \ge \text{ceiling}(\log(3, 2n+1)) (because h is an integer)
    [read: ceiling of log base 3 of 2n+1]
   Therefore the algorithm must perform at least ceiling(log(3, 2n+1)) weighings.
 QED
_____
2.
  (a)
  ____
   TRUE;
   Definition Of Lower Bound \Omega:
    ______
     Let T(n), T: N \to R+, be a function that describes the worst-case running time
```

of a given algorithm in terms of steps to be completed per size of input n. We say that T(n) is $\Omega(f(n))$ i.e. we say that (T(n)) is asymptotically lower bounded by f(n), $f: N \to R+$, i.e. if there is a constant $c \in R+$ and $n0 \in N$ such that $T(n) \ge c \cdot f(n)$ for $\forall n \in N$, $n \ge n0$. If T(n) is $\Omega(f(n))$ we also write $T(n) = \Omega(f(n))$

```
Using This definition, then it logically follows that if f \in \Omega(g) is equivelant to f(n) \ge 1
 c \cdot q(n) for some c and n \ge n0.
 We can first prove that f \in \Omega(g):
   Premise 1: f(n) \ge c \cdot q(n)
   Let:
      f(n) = n
     q(n) = n
      c < c
      n0 = 1
    If we then evaluate the function in Premise 1:
      n \ge (c) \cdot (n) \quad \forall n \in N, \quad n \ge n0
    Which holds true and can be proven by taking the first derivative of each side of the
    function
      d/dn(n) = 1
      d/dn(nc) = c
   As the first derivative of n is greater than or equal to nc when 0 \le c \le 1 and the second
    derivative of both is zero:
      We have shown f \in \Omega(g) and Premise 1 holds.
 Now we can prove that f^2 \in \Omega(g^2):
    Since we have proven Premise 1, it is a trivial matter to prove that applying any
    exponential growth function to either side will lead to the same result by applying the
    same method of proving as for Premise 1:
    Let:
      f(n) = n
      q(n) = n
      C = C
     n0 = 1
    If we evalaute the function:
      (f(n))^2 \ge c \cdot (g(n))^2 \quad \forall n \in \mathbb{N}, n \ge n0
                                \forall_n \in N, n \geq n0
      (n)^2 \ge (c) \cdot (n)^2
    The function holds true and can be proven by taking the first and second derivatives of
      derivative(n^2) = 2n
      derivative(2n) = 2
      derivative(c(n)^2) = 2(c)(n)
      derivative(2cn) = 2(c)
    As any value c \ge 1 will make the rate of the rate of growth of c(n)^2 greater than that
    of n^2,
      (f(n))^2 \ge c \cdot (g(n))^2 \forall n \in \mathbb{N}, n \ge n0
   holds true, we have proven that if f \in \Omega(q) is true, f^2 \in \Omega(q^2) is also TRUE.
 QED
____
(b)
____
 FALSE;
 Definition of Upper Bound O:
   Let T(n), T: N \rightarrow R+, be the function that describes the worst-case running
```

time of a given algorithm in terms of steps to be completed per sitesize of input n. We say that T(n) is O(f(n)) (read "T(n) is of order f(n)") if T is asymptotically upper bounded by f(n), $f: N \to R^+$, i.e. if there is a constant $c \in R^+$ and $n0 \in N$ such that $T(n) \le c \cdot f(n)$ for $\forall n \in N$, $n \ge n0$. If T(n) is O(f(n)), we also write T(n) = I(f(n)).

Using the definition of Upper Bound O, $f \in O(g)$ is equivelant to $f(n) \le c \cdot g(n)$ for some c and $n \ge n0$.

This means that g(n) is not necessarily a member of f(n) however a constant c exists that when multiplied together with g(n) makes it so.

With this is mind, we can make an example in which:

```
f(n) = 2n
```

$$g(n) = n$$

We know that:

some c exists such that $f(n) \le c \cdot g(n)$

Example: if
$$(c == 3)$$
 {
 $f(n) < c \cdot g(n)$
}

However this property does not transfer to a function which applies f(n) and g(n) as an exponential value; such as the one asked here;

If we define f(n) and g(n) using the same values as they were ealier when showing that $f(n) \le c \cdot g(n)$ we get:

f(n) = 2n

g(n) = n

This means that:

$$2^{(f(n))} = 2^{(2n)}$$

$$2^{(g(n))} = 2^{(n)}$$

Are question asks to prove/disprove $2^f \in O(2^g)$, so mathimatically speaking, for some c:

```
2^f(n) \le c \cdot 2^q(n)
```

$$2^(2n) \le c \cdot 2^(n)$$

Must be true for all values n greather than some n0.

To state this otherwise is:

```
Limit(2^{(2n)}, infinity) \leq Limit(2^{(n)}, infinity)
```

[Read: Limit of 2 to the power of 2n is lesser than or equal to the limit of 2 to the power of n when n approaches infinity]

We can show this is never true as taking the second derivative of both evaluate to:

```
derivative (derivative (2^{(2 n)}) = 4^{(n+1)} \cdot \log^2(2)
```

```
derivative (derivative (2^n)) = 2^n \cdot \log^2(2)
```

As shown, the rate which the rate of $2^{(2n)}$ increases is greater than the rate of the rate of $2^{(n)}$, therefore, no c exists such that:

```
2^{(2n)} \le c \cdot 2^{(n)}
```

can be true for some n≥n0

We have proved that even if $f \in O(g)$ is true, $2^{\circ}(f) \in O(2^{\circ}g)$ is FALSE.

QED.