CS325 HW1

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1.

After simplification, we can just compute the function $n=8\log_2(n)$, and we got the two results, $n\approx 1.1$

 $n \approx 43.5593$

And we found that while n is bigger than 1.1 but smaller than 43.5593, the insertion running time is smaller than merge, which is faster. So, the answer is [1.1 - 43.5593].

2.

- a. O(g(n)), because the limit as n goes to infinity of f(n) / g(n) is 0.
- b. $\Omega(g(n))$, because the limit as n goes to infinity of f(n) / g(n) is infinity.
- c. $\Theta(g(n))$, because the limit as n goes to infinity of f(n) / g(n) is a constant.
- d. $\Theta(g(n))$, because the limit as n goes to infinity of f(n) / g(n) is a constant.
- e. O(g(n)), because the limit as n goes to infinity of f(n) / g(n) is 0.
- f. O(g(n)), because the limit as n goes to infinity of f(n) / g(n) is 0.
- g. $\Theta(g(n))$, because the limit as n goes to infinity of f(n) / g(n) is a constant.
- h. O(g(n)), because the limit as n goes to infinity of f(n) / g(n) is 0.
- i. $\Omega(g(n))$, because the limit as n goes to infinity of f(n) / g(n) is infinity.
- j. O(g(n)), because the limit as n goes to infinity of f(n) / g(n) is 0.

3.

- a. Proves, because $f1(n) = \Theta(g(n))$ so, f1(n) = c1g(n) and $f2(n) = \Theta(g(n)) = c2g(n)$ we can get f1(n)/f2(n) is equally a Constant c so $f1(n) = \Theta f2(n)$ is prove
- b. disproves, if $f1(n)=f2(n)=2^n g1(n)=g2(n)=20^n$ so, $\lim n\to \inf$ inity, (f1(n)+f2(n))/(g1(n)+g2(n)) is 0. It's O(g(n)) instead of O(g(n)), there is a contradiction.

4.

See teach files

5.

a.

insertTime.py

```
from random import randint
from time import time
MIN\_INT = 0
MAX_INT = 10000
def insertsort(array):
  #go through each element in the array
  for j in range (1,len(array)):
    key = array[j];
    i = j - 1
    while i \ge 0 and array[i] > key:
       array[i+1] = array[i]
       i = i - 1
    array[i+1] = key
  return array
if __name__ == "__main___":
  begin = 10000
  increment = 5000
  values = [begin + i * increment for i in range(7)]
  list = [[randint(MIN_INT,MAX_INT) for _ in range(value)] for value in values]
```

print ("value

for array in list:

array_size = len(array)

running time")

```
start_time = time()
     insertsort(array)
     print("{}
                    {}".format(array_size,time()-start_time))
mergeTime.py
from random import randint
from time import time
MIN_INT = 0
MAX_INT = 10000
def merge(left,right):
  #create empty array to hold sorted values
  array = []
  #keep merge until one side has no element left
  while len(left) != 0 and len(right) != 0:
     #if the first element of left array is smaller merge it into empty array
     #and remove the merged element
     if left[0] < right[0]:
       array.append(left[0])
       left.remove(left[0])
     # the situation that is equal or bigger than right side
     else:
       array.append(right[0])
       right.remove(right[0])
```

```
#if the left side array is empty, then append the right side
  if len(left) == 0:
     array += right;
  else:
     array += left;
  return array
#define mergesort algorithm
def mergesort(array):
  array\_size = len(array)
  if array_size <= 1:
     return array
  else:
     array_split = array_size//2
     left_array = mergesort(array[:array_split])
     right_array = mergesort(array[array_split:])
     return merge(left_array,right_array)
if __name__ == "__main___":
  begin = 10000
  increment = 5000
  values = [begin + i * increment for i in range(7)]
  list = [[randint(MIN_INT,MAX_INT) for _ in range(value)] for value in values]
  print ("value
                      running time")
  for array in list:
     array\_size = len(array)
```

b.

Here is the table result from my algorithms, I took three tables and got the average running time. insertTime table

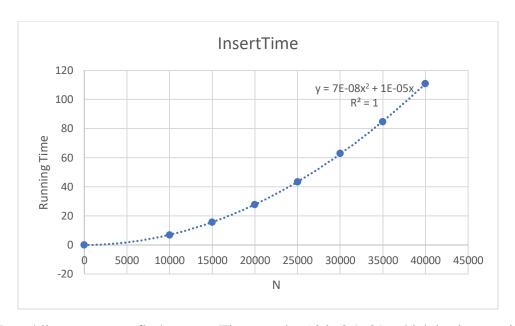
Ν		averageT
	0	0
	10000	6.9
	15000	15.69
	20000	27.78
	25000	43.33
	30000	62.93
	35000	84.89
	40000	110.91

mergeTime table

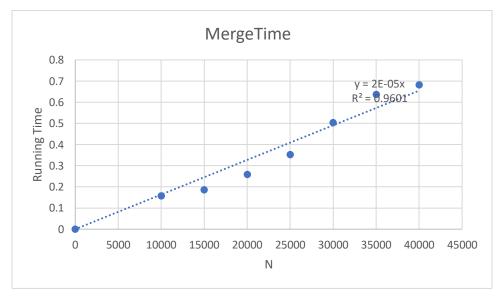
N	averageT
0	0
10000	0.1578
15000	0.1857
20000	0.2593
25000	0.3531
30000	0.5042
35000	0.6367
40000	0.682

c.

I used the polynomial curve to fit the insertTime graph, r^2 is 1, which means it fits perfectly.

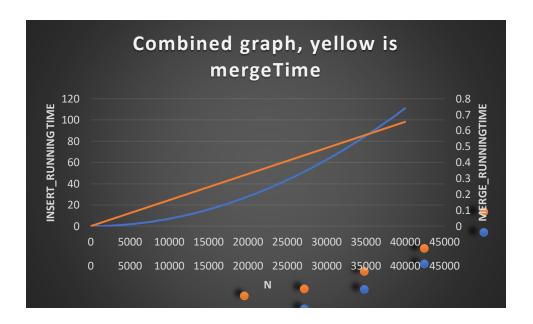


And I used linear curve to fit the mergeTime graph, r^2 is 0.9601, which is also considered as extreme good.



d.

The combined graph, the yellow line is mergeTime, the blue one is insertTime, and there are two Y-aixs.



e.

For the Insert Sort algorithm, when using the worst-case input, the experimental running times appeared to be very close to the theoretical worst case, $O(n^2)$, as expected. The curve fit for the above graph points to an equation of the $f(n) = n^2$ nature. When the best-case input was use, the experimental running times appeared to be very close to the theoretical best case, O(n), as expected.

For the Merge Sort algorithm, when using the worst-case input, the experimental running times appeared to be very close to $O(n \lg n)$, as expected. There technically isn't a worst case for the Merge Sort algorithm, but it was discovered an array setup that would be the hardest to sort. See link at the bottom of mergesort_worst.py The curve fit for the above graph points to an equation of the f(n) = n nature. When the best-case input was use, the experimental running times appeared to be very close to $O(n \lg n)$, as expected. Again, there technically isn't a best case for the Merge Sort algorithm but the results of this "best case" were faster than the "worst case".