

Efficiency: Continued

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Announcements:

1. Term test 2 marks has been released, good class average and median
2. Lab 8 has been released
 - a. Infer big-oh from timing
 - b. Almost no-coding
3. Assignment 2: technical glitch for autotesting, should be fixed soon

Agenda:

Worst case performance of Quick Sort

Python recursion limit

Merge Sort performance

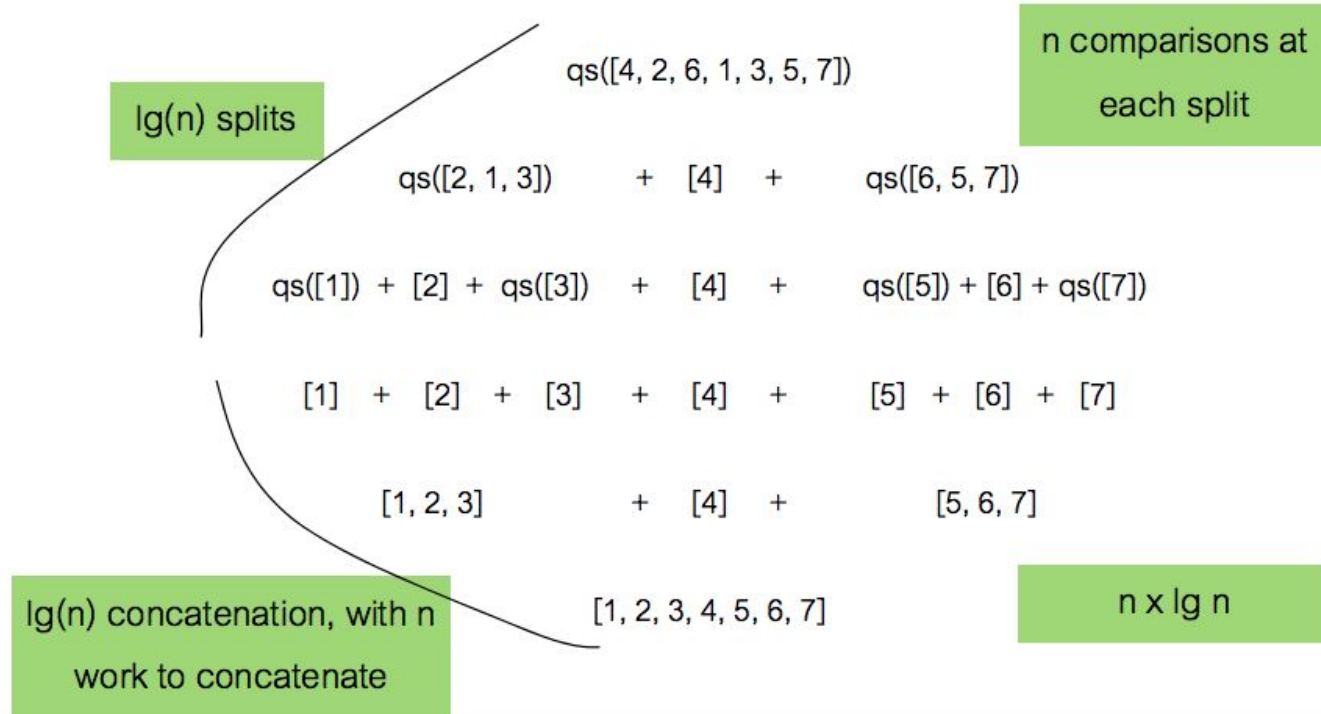
In-class exercise for Big-oh assessment

Quick Sort: Recap

What is the average case performance of Quick sort?

What is the worst case performance?

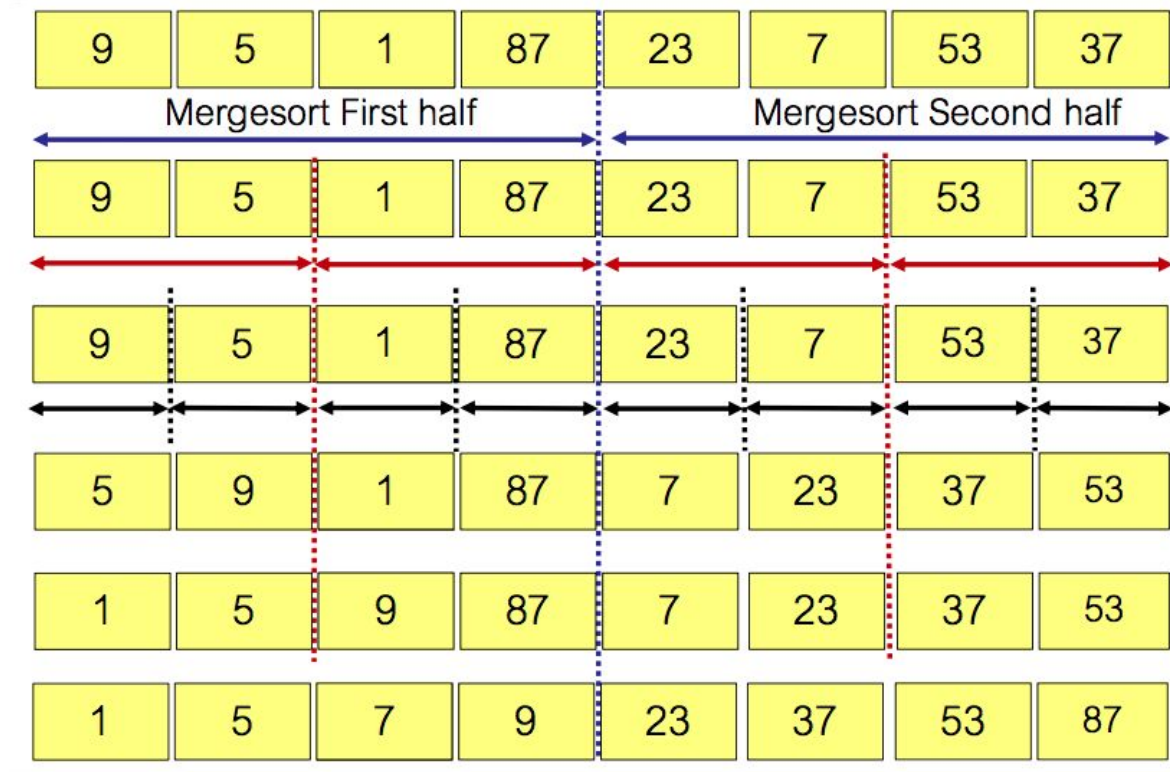
Quick Sort: Average Case



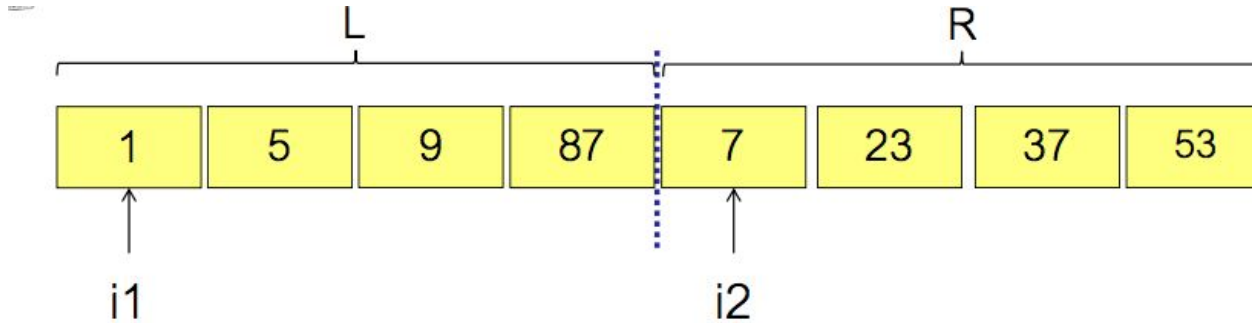
Merge Sort

- Split a list in two halves repeatedly
- Halves with 0 or 1 elements are guaranteed sorted
- Merge the two halves "on the way back"

Merge Sort



Merge Step : merge(L, R)



orted list (different than L or R)



The halves might not be perfectly equal though...

Code for mergesort

idea: break a list up (partition) into two halves, mergesort each half, then recombine (merge) the halves

```
def mergesort(list_):
```

```
    """
```

```
    Produce a copy of list_ in sorted order.
```

```
    """
```

```
    if len(list_) < 2:
```

```
        return list_[:]
```

```
    else:
```

```
        return merge(mergesort(list_[:len(list_) // 2]),
```

```
                      mergesort(list_[len(list_) // 2 :]))
```

Lists of length < 2 are
already sorted

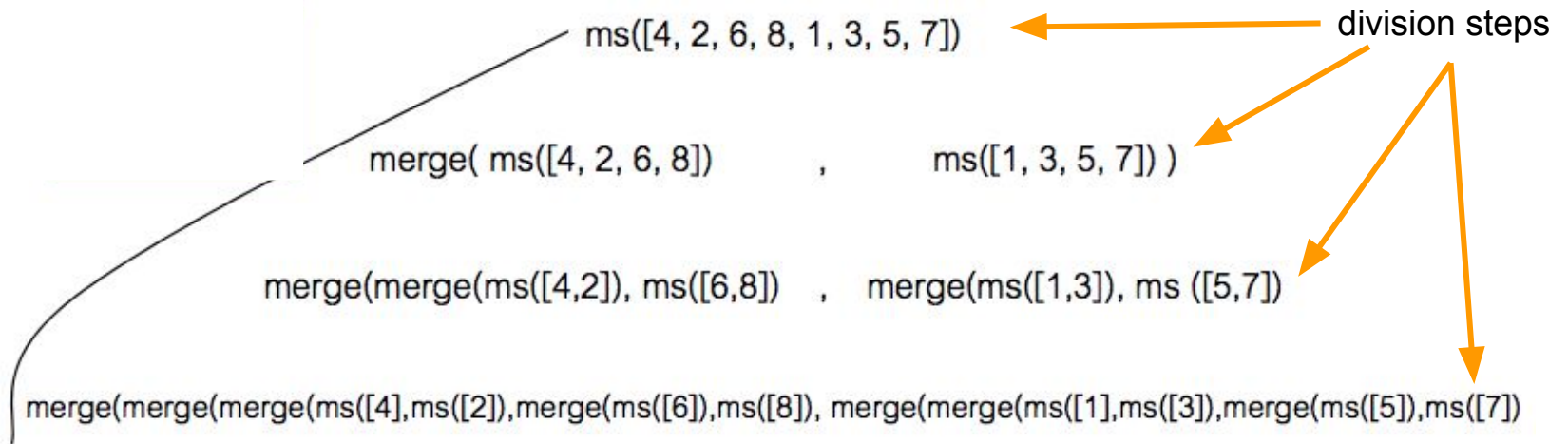
First half

Second half

Merge the two sorted halves,
"on the way back". How?

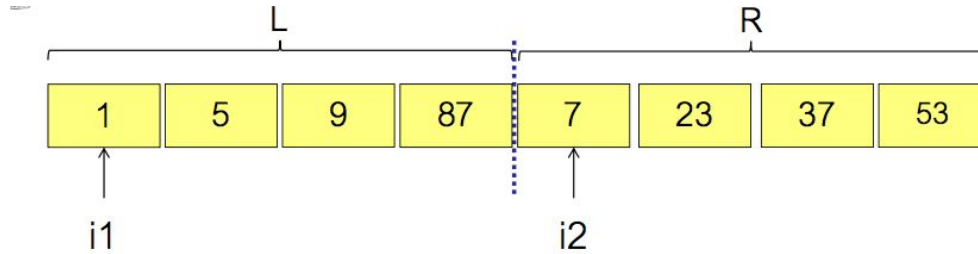
Divide and Conquer

Division step analysis

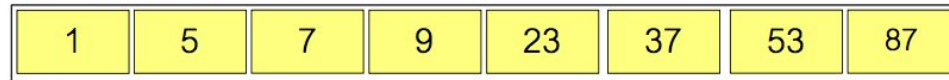


Divide and Conquer

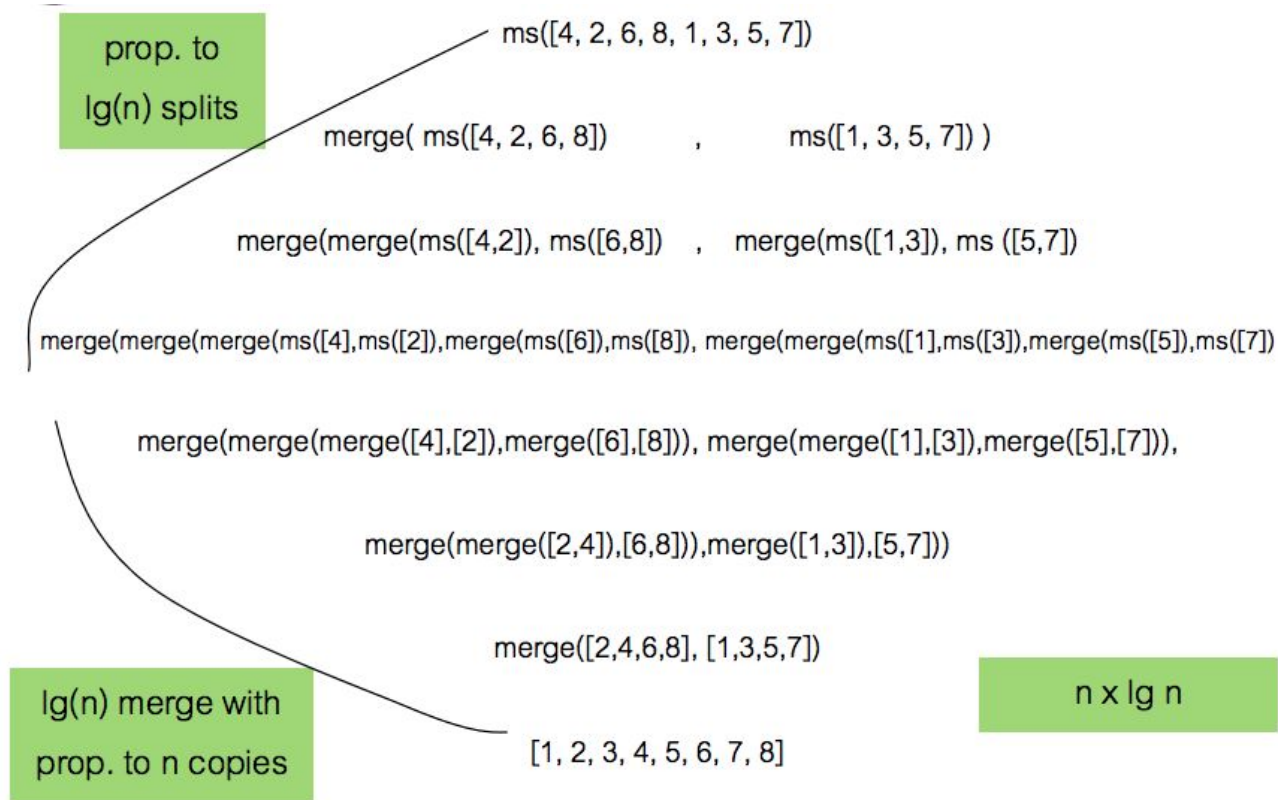
Conquer (merge step)



orted list (different than L or R)



Counting mergesort: $n = 8$



Python Recursion Limits

```
import sys
```

```
def fact(n):  
    if n == 0 or n == 1:  
        return 1  
    else:  
        return n * fact(n - 1)
```

```
if __name__ == '__main__':  
    fact(999)
```



```
return n * fact(n - 1)  
File "/Users/macbook15/PycharmProjects/CSC148H1S/week10/limit.py", line 8, in fact  
    return n * fact(n - 1)  
[Previous line repeated 994 more times]  
File "/Users/macbook15/PycharmProjects/CSC148H1S/week10/limit.py", line 5, in fact  
    if n == 0 or n == 1:
```

RecursionError: maximum recursion depth exceeded in comparison

```
Process finished with exit code 1
```

Efficiency: A few tips

1. Two algorithms to solve the same problem -- two Big Oh classes ($O(n^2)$ and $O(n \lg n)$), which one to choose?
2. Timing tools might be noisy, take averages of multiple runs
3. A good computer scientist should perform analytical (pen and paper) AND empirical (measuring and plotting) analysis

Big Oh*

Suppose the number of “steps” (operations that don’t depend on n , the input size) can be expressed as $t(n)$. We say that $t \in \mathcal{O}(g)$ if:

there are positive constants c and B so that for every natural number n no smaller than B ,
$$t(n) \leq cg(n)$$

*Taken from Danny’s slides

Empirical study

$$t(n) = 7n^2$$

$$t(n) = n^2 + 396$$

$$t(n) = 3960n + 4000$$

Which one will have the largest growth?

Onto gnuplot

Exercise: Analytical determination of Time Complexity

Exercise 1:

```
def silly(n):  
    n = 17 * n**(1/2)  
    n = n + 3  
    print("n is: {}".format(n))  
  
    if n > 97:  
        print('big!')  
    else:  
        print('not so big!')
```

How does the running time of **silly** depend on **n**?

Exercise 2:

How does the running time of this code fragment depend on n ?

```
sum = 0
for i in range(n):
    sum += i
```

Exercise 3:

How does the running of this code fragment depend on n ?

```
i, sum = 0, 0
while i**2 < n:
    j = 0
    while j**2 < n:
        sum += i * j
        j += 1
    i += 1
```

Empirical verification

Onto PyCharm