Test I Study Guide

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1 Sorting Algorithms

1.1 Insertion-Sort

```
Insertion-Sort(A)
for j = 2 to A.length
key = A[j]
// Insert A[j] into the sorted sequence A[1..j - 1]
i = j - 1
while i > 0 and A[j] > key
A[i + 1] = A[i]
i = i - 1
A[i + 1] = key
```

1.2 Merge Sort

```
Merge-Sort(A, p, r)
        if p < r
2
             q = \lfloor (q + r) / 2 \rfloor
             Merge-Sort(A, p, q)
             Merge-Sort(A, q + 1, r)
             Merge(A, p, q, r)
    Merge(A, p, q, r)
        n_1 = q - p + 1
9
        n_2 = r - q
10
        let L[1...n_1 + 1] and R[1...n_2] be new arrays
11
        for i = 1 to n_1
             L[i] = A[p + i - 1]
13
        for j = 1 to n_2
14
            R[j] = A[q + j]
15
        R[n_1 + 1] = \infty
        R[n_2 + 1] = \infty
17
        i = 1
        j = 1
19
```

1.3 Heap Sort

```
Max-Heapify(A, i)
        l = Left(i)
        r = Right(i)
        if 1 \le A.heap-size and A[1] > A[i]
            largest = 1
        else largest = i
        if r \leq A.heap-size and A[r] > A[largest]
            largest = r
        if largest \neq i
9
            exchange A[i] with A[largest]
10
            Max-Heaphify(A-largest)
11
12
   Heapsort(A)
13
        Build-Max-Heap(A)
14
        for i = A.length downto 2
15
            exchange A[1] with A[i]
16
            A.heap-size = A.heap-size - 1
17
            Max-Heapify(A, 1)
18
```

1.4 Quicksort

```
Quicksort(A, p, r)
        if p < r
2
            q = Partition(A, p, r)
            Quicksort(A, p, q - 1)
            Quicksort(A, q + 1, r)
   Partition(A, p, r)
        x = A[r]
        i = p - 1
9
        for j = p to r - 1
            if A[j] \leq x
11
                i = i + 1
12
                exchange A[i] with A[j]
13
        exchange A[i + 1] with A[r]
14
        return i + 1
15
```

1.5 Counting Sort

Let A = input array, B = output array, n = size of array, k = maximum number.

```
countingSort(A, B, n, k)
let C[0...k] be the counting array
for i = 0 to k
C[i] = 0
for j = 1 to n
C[A[j]] = C[A[j]] + 1
for i = 1 to k
C[i] = C[i] + C[i - 1]
for j = n down to 1
D[C[A[j]]] = A[j]
C[A[j]] = C[A[j]] - 1
```

1.6 Rate of Growth

```
Insertion Sort Worst: \Theta(n^2), Average: \Theta(n^2)
Merge Sort Worst: \Theta(n \lg n), Average: \Theta(n \lg n)
Quick Sort Worst: \Theta(n^2), Average: \Theta(n \lg n)
Heap Sort Worst: \Theta(n \lg n), Average: \Theta(n \lg n)
```

2 Growth Classes

2.1 O-notation

 $O(g(n)) = \{f(n) : \text{ there exists positive constant } c \text{ and } n_0 \text{ such that } 0 \le f(n) \le c g(n) \text{ for all } n \ge n_0\}$

2.2 Ω -notation

 $\Omega(g(n)) = \{f(n) : \text{ there exists positive constants } c \text{ and } n_0 \text{ such that } 0 \le c \, g(n) \le f(n) \text{ for all } n \ge n_0\}$

2.3 Θ -notation

```
\Theta(g(n)) = \{f(n) : \text{ there exists positive constants } c_1, c_2 \text{ and } n_0 \text{ such that } 0 \le c_1 \ g(n) \le f(n) \le c_2 \ g(n) \text{ for all } n \ge n_0\}
```

2.4 o-notation (Little-o)

 $f(n) \in o(g(n))$ iff for every c > 0 there is an $n_0 > 0$ such that

$$0 \le f(n) < c \ g(n)$$

for all $n \geq n_0$

2.5 ω -notation (Little-omega)

$$f(n) \in \omega(g(n))$$
 iff $g(n) \in o(f(n))$.

or $f(n) \in \omega(g(n))$ iff for every c > 0 there is an $n_0 > 0$ such that

$$0 \le c \ g(n) < f(n)$$

for all $n \ge n_0$

3 Master Theorem

The master theorem can only be applied to recurrence equations of the form:

$$T(n) = aT(n/b) + f(n)$$

3.1 Constants

- ${f n}$ The size of the problem
- a The number of subproblems
- n/b The size of each subproblem
- f(n) cost outside of recursive calls (divide, combine)

3.2 Cases

$f(n) \in O(n^{\log_b a - \epsilon})$	$T(n) \in \Theta(n^{\log_b a})$
$f(n) \in \Theta(n^{\log_b a})$	$T(n) \in \Theta(n^{\log_b a} \lg n)$
$f(n) \in \Omega(n^{\log_b a + \epsilon})$	$T(n) \in \Theta(f(n))$

4 Definitions

Algorithm Any well-defined computational procedure that takes a set of values as input and produces a set of values as output in a finite number of steps

Correct Algorithm One returns the correct solution for every valid instance of a problem

Loop Invariance Define a key property about the relationship among variables of the algorithms.

• Holds in the *initial* case.

Initialization The loop invariance must be true prior to the first iteration.

• Is maintained each iteration.

Maintenance If the property holds prior to an iteration, it must still hold after the iteration is complete.

 $\bullet\,$ Yields correctness when the loop terminates.

Termination The invariant provides a useful property that helps demonstrate the algorithm is correct.

Heapify Go all the way down to the heap and fix the violations of the max-heap property by sifting-up