

1.)

a)

$$T(n) = T(n-2) + n$$

$$\text{Guess } T(n) = \Theta(n^2)$$

$$\text{Hypothesis: } T(n) \geq cn^2 \text{ or } T(n) = \Omega(n^2)$$

$$T(n) = T(n-2) + n$$

$$c(n-2)^2 + n$$

$$c(n-2)(n-2) \rightarrow \text{foil} \rightarrow c(n^2 - n - n + 4) + n \rightarrow cn^2 - 2cn + 4c + n$$

$$\geq cn^2$$

If  $n(-2c+1) + c \geq 0$  which requires  $n \geq 0$  and  $c$  is  $0 < c \leq 1/2$ .

$$\text{Upper bound } T(n) = O(n^2)$$

$$\text{Hypothesis: } T(n) \leq cn^2$$

$$T(n) = T(n-2) + n$$

$$c(n-2)^2 + n$$

$$c(n-2)(n-2) \rightarrow \text{foil} \rightarrow c(n^2 - n - n + 4) + n \rightarrow cn^2 - 2cn + 4c + n$$

$$\geq cn^2$$

If  $n(-2c+1) + c \geq 0$  which requires  $n \geq 1$  and  $c$  is 1.

$$T(n) = \Omega(n^2) \text{ and } T(n) = O(n^2) \text{ thus } T(n) = \Theta(n^2)$$

$$\text{b) } T(n) = T(n-1) + 3$$

$$T(n) = T(n-1) + 3$$

$$= T(n-2) + 6$$

$$= T(n-3) + 9$$

$$= T(n-4) + 12$$

$$= T(n-k) + 3k$$

Replace  $k$  with  $n-1$

$$= T(n-(n-1)) + 3(n-1)$$

$$= T(n) = 3n-3$$

$$T(n) = \Theta(n)$$

c) Case 3 as pointed out by Tim Thomas.

$$n^{\log_4 2} = n^{1/2} = \sqrt{n}$$

$$\text{so } f(n) = \Omega(n^{.5+\epsilon}) \text{ for } \epsilon < .5$$

Regularity check:

$$af(n/b) = 2(n/4) = n/2 \leq cn$$

satisfied for  $c$  of  $\rightarrow 1/2 \leq c < 1$

$$\text{d) } T(n) = 4T\left(\frac{n}{2}\right) + n^2\sqrt{n}$$

$$\text{Master theorem: } T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$$f(n) = n^2\sqrt{n} = n^{5/2} \text{ and } n^{\log_b a} = n^{\log_2 4} = n^2$$

$$\text{since } n^{5/2} = \Omega(n^{\log_2 4 + 3/2}) - n^{5/2} = n^{\log_2 4 + 3/2} \checkmark$$

Case 3:

$$af(n/b) = 4(n/2)^2\sqrt{n}/2 = n^{5/2}\sqrt{2} \leq cn^{5/2} \text{ for } 1/\sqrt{2} \leq c < 1$$

$$T(n) = \Theta(n^2\sqrt{n})$$

2.)

A.)

$$T(n) = 5 * T(n/2) + O(n)$$

$$a=5; b=2; d=1$$

$$d < \log(a) = 1 < \log_2(5)$$

$$\text{so } O(n^{\log_2 5}) = \underline{O(n^{2.3219})}$$

B.)

$$T(n) = 2 * T(n-1) + O(1)$$

$$T(2) = 2 * T(1)$$

$$T(3) = 2 * T(2) = 2 * 2 * T(1)$$

$$T(5) = 2 * T(4) = 2 * 2 * 2 * 2 * T(1)$$

$$\text{so } \underline{O(2^n)}$$

C.)

$$T(n) = 9 * T(n/3) + O(n^2)$$

$$a=9; b=3; d=2$$

$$d = \log_2 a \rightarrow 2 = \log_3 9$$

$$\text{so } \underline{O(n^2 \log n)} \checkmark$$

3.)

$$T(n) = 4T\left(\frac{n}{2}\right) + n$$

4T for the 4 recursive calls, n/2 for the work done, and n for the for loop

Master Method: a=4, b=2, f(n)=n

$$n^{\log_2 4} = n^2$$

Case 1 satisfied.

$$\underline{\Theta(n^2)}$$

4.)

ternaryFunction(array[], searchValue, int startOfArray, int endOfArray)

```
{
    int firstThird = end/3;
    int secThird = 2*end/3;
    if(firstThird == searchValue)
        return true;
    else if(secThird == searchValue)
        return true;
    else if(searchValue < array[firstThird])
        then return first third to function
        return ternaryFunction(array, searchValue, startOfArray, firstThird - 1;
        //- 1 because we already checked the firstThird value.
    else if(searchValue > array[secThird])
        then return the third third to function to search
        return ternaryFunction(array, searchValue, secThird+1, endOfArray)
        //+1 because of similar reasons above.
    else
        return ternaryFunction(array, searchValue, firstThird, secThird)
}
```

$$\text{Binary} = T_2(n/2) + 1$$

$$\text{Ternary} = T_3(n/3) + 2$$

Just like binary search, the second case of Masters Theorem is relevant with  $a = 2$ ,  $b = 3$

And  $f(n) = \Theta(2)$

$$\text{Thus: } f(n) = \Theta(2) = \Theta(n^{\log_b a})$$

$$\text{Using the second form, } T_3(n) = \Theta(\log n)$$

5.)

A.)

Merging two sorted arrays of  $n_1$  and  $n_2$  elements will take  $O(n_1 + n_2)$  time,

Then we will add another sorted array of size  $n$ , and so on.

$$(n_1 + n_2) + (2n + n_3) + (3n + n_4) + \dots + ((k-1)n + n_k)$$

$$= 2n + 3n + 4n + \dots + kn$$

$$= \underline{\underline{O(k^2 n)}}$$

B.)

$O(kn)$  work required to move  $k$  arrays of size  $n$  into  $k/2$  arrays of size  $2n$ . We continue doing  $O(kn)$  work  $O(\log k)$  times until we have an array of size  $kn$ .

$$\text{Runtime} = \underline{\underline{O(kn \log k)}}$$

6.)

A.)

Insertion sort takes  $\Theta(k^2)$  time to sort a list size  $k$ . Since we have a list size of  $\frac{n}{k}$  we have  $\frac{k^2 * n}{k}$  or this breaks down into  $\Theta(nk)$  worst case.

B.)

$$\Theta(n \lg \left(\frac{n}{k}\right))$$

C.)

$$T(nk + n \lg \left(\frac{n}{k}\right)) = \Theta(n \lg n) \rightarrow$$

$$\text{Let } k = \Theta(\lg n)$$

$$\Theta(n \lg n + n \lg(n / \lg n))$$

$$\Theta(n \lg n + n(\lg(n) - \lg(\lg n)))$$

$$\Theta(n \lg n + n \lg n - n \lg(\lg n))$$

$$\Theta(2n \lg n - n \lg(\lg n))$$

$$\Theta(n \lg n) \text{ so } k = \underline{\underline{\Theta(\lg n)}}$$

D.)

For certain values of  $k$ , insertion sort runs much faster than merge sort.

We could pick  $k$  values by setting the insertion sort equation equal to the

Merge sort equation, simplify, and plug and chug. After  $k$  is larger than

43, merge sort is quicker than insert sort. We determined this in last week

Homework.

7.)

compare(array[(a-3),(a-2),(a-1), a] of numbers) -> (minLR, maxLR)

if ( $n == 1$ )

return (array[1], array[1])

else if ( $n == 2$ )

if (array[1] < array[2])

return (array[1], array[2])

else

return (array[2], array[1])

else

```
(maxL, minL) = compare(array[1...(n/2)])
(maxR, minR) = compare(array[(n/2 + 1)...n])
if (maxL < maxR)
    max = maxR
else
    max = maxL
if (minL < minR)
    min = minL
else
    min = minR
```

Recurrence =  $T(n) = 2T(n/2) + 3$

$\Theta(n)$