

Alan Marcero Algorithm Analysis

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1. Use the LCS algorithm developed in class to compute the LCS of “exponential” and “polynomial”. Show all of your work including the table.

LCS: PONIAL

			E	X	P	O	N	E	N	T	I	A	L
		0	1	2	3	4	5	6	7	8	9	10	11
	0	0	0	0	0	0	0	0	0	0	0	0	0
P	1	0	0	0	1	1	1	1	1	1	1	1	1
O	2	0	0	0	1	2	2	2	2	2	2	2	2
L	3	0	0	0	1	2	2	2	2	2	2	2	3
Y	4	0	0	0	1	2	2	2	2	2	2	2	3
N	5	0	0	0	1	2	3	3	3	3	3	3	3
O	6	0	0	0	1	2	3	3	3	3	3	3	3
M	7	0	0	0	1	2	3	3	3	3	3	3	3
I	8	0	0	0	1	2	3	3	3	3	4	4	4
A	9	0	0	0	1	2	3	3	3	3	4	5	4
L	10	0	0	0	1	2	3	3	3	3	4	5	6

2. Assume we are computing the LCS of S and T using the algorithm we developed in class (not the bad recursive algorithm).

- a) **Find and prove a run-time Big-Oh for the LCS algorithm that computes the length of the LCS.**

```

LCS(s, t) {
    // Fill in row and column 0
    for j = 0 to s c[0][j] = 0; // O(s)
    for i = 0 to t C[i][0] = 0; // O(t)

    for i = 1 to s    // for each row i
        for j = 1 to t // for each column j of I O(s*t)

            //do constant operation
}

```

The two initial for loops run $(S + T)$ times, the for loop with the nested for loop runs $(S * T)$ times. Thus this algorithm takes $(S + T) + (S * T)$ or $2n + n^2$ which is $O(n^2)$

Proof:

$$2n + n^2 \leq C_1 n^2$$

replace $2n$ with $2n^2$:

$$2n^2 + n^2 \leq C_1 n^2 = 3n^2 \leq C_1 n^2$$

set the constant = 3:

$$3n^2 \leq 3n^2$$

b) **Find and prove a Big-Oh that describes the amount of memory that LCS uses.**

```
LCS(s, t) {  
    // Fill in row and column 0  
    for j = 0 to s c[0][j] = 0;  
    for i = 0 to t C[i][0] = 0;  
  
    for i = 1 to s    // for each row i  
        for j = 1 to t // for each column j of I O(s*t)  
  
            //do constant operation  
  
}
```

In the second pair of for loops, the algorithm fills a two dimensional array of size (S * T). Thus this algorithm takes up (S * T) space in memory or O(ST) space.

Proof:

$$ST \leq C_1 ST$$

Set the constant = 1:

$$ST \leq ST$$

3. Develop a version of LCS that uses $O(n+m)$ space.

```
// The LCS algorithm only looks at the array position to the top
// of, and directly to the left of the current iteration in order
// to calculate the LCS. This algorithm utilizes that fact in
// order to implement an LCS algorithm that uses only the space
// of two arrays in memory -  $O(n+m)$ .
```

```
LCS(m, n) {
    // Fill in the top array with 0s and position 0 of the bottom
    // array with a 0
    for (i = 0 to m.length + 1) { top[i] = 0; }
    bottom[0] = 0;

    for i = 1 to m.length    // for each row i
        for j = 1 to n.length { // for each column j of i

            // if the string chars are equal at the current iteration
            // set the current array position to 1 + the
            // left-diagonal array position
            if m[i] = n[j]
                bottom[i] = 1 + top[j-1];

            // if they are not equal, set the current array position
            // to the max of the left array position, and the
            // the adjacent top array position
            else
                bottom[j] = max(bottom[j-1], top[j])

        }

        // copy the bottom row to the top row and continue
        top = bottom;
    }

    // the algorithm is now complete, return the last position of
    // the bottom array
    return bottom[m.length];
} // end LCS
```

4. Develop a modified version of LCS that computes the *minimal* number of insert and delete edits needed to convert a string S into a string T (rather than the length of the LCS of S and T).

```
//
// This modified version of LCS computes the minimal number of
// edits (inserts and deletes) required to change the input
// string s into the input string t
//
// First it fills the top row and left-most column with
// 0..s.length or 0..t.length
// Second it calculates the edit distance by going through each
// cell in the 2-dimensional array.
//
// It does this in two ways:
// If the chars of the two strings s&t are equal at the current
// iteration, set the current cell block to the key in the
// left-diagonal cell.
//
// If they are not equal, calculate the min of the position to
// the left of the current iteration + 1 and to the top of the
// current iteration + 1 and set the current cell block to the
// min.
//
// Then proceed to fill the rest of the table cells until it is full
//
editDistance(s, t) {

    // Fill in row and column 0 with 0..n

    // fill row 0 with 0..s.length
    for j = 0 to s.length C[0][j] = j;

    // fill column 0 with 0..t.length
    for i = 0 to t.length C[i][0] = i;

    for i = 1 to s.length { // for each row i
        for j = 1 to t.length { // for each column j of i

            // if the two strings are equal at this iteration
            if (s[i] == t[j])
                C[i][j] = C[i-1][j-1];
            // if they are not equal, calculate the min of the
            // the top position + 1 or the left position + 1
            // then place the min in the current table cell
            else
                C[i][j] = min(
                    C[i][j-1]+1,
                    C[i-1][j]+1
                );
        }
    }
    // the algorithm is complete, return the final position in
    // the table
    return C[s.length][t.length];
}
```

5. Modify the algorithm you developed in 4 to compute the *minimal* number of edits including the change operation needed to convert S into T. Trace your algorithm on “exponential” and “polynomial”.

```
//
// This modified version of the LCS based editDistance algorithm
// also accounts for the “change operation” required to change a
// string s into the string t.
//
// It is nearly exactly the same as the original algorithm with
// one key difference:
//
// If they are not equal, calculate the min of the position to
// the left of the current iteration + 1, to the top of the
// current iteration + 1, and to the diagonal-left of the current
// position + 1. Then set the current cell block to the min.
//
editDistance(s, t) {

    // Fill in row and column 0 with 0..n

    // fill row 0 with 0..s.length
    for j = 0 to s.length C[0][j] = j;

    // fill column 0 with 0..t.length
    for i = 0 to t.length C[i][0] = i;

    for i = 1 to s.length { // for each row i
        for j = 1 to t.length { // for each column j of i

            // if the two strings are equal at this iteration
            if (s[i] == t[j])
                C[i][j] = C[i-1][j-1];

            // Here is the key difference:
            // if they are not equal, calculate the min of the
            // the top position + 1, the left position + 1,
            // and the diagonal-left position + 1. Then place the
            // min in the current table cell.
            else
                C[i][j] = min(
                    C[i][j-1]+1,
                    C[i-1][j]+1,
                    C[i-1][j-1]+1
                );

        }
    }
    // the algorithm is complete, return the final position in
    // the table
    return C[s.length][t.length];
}
```

After the first two for loops run:

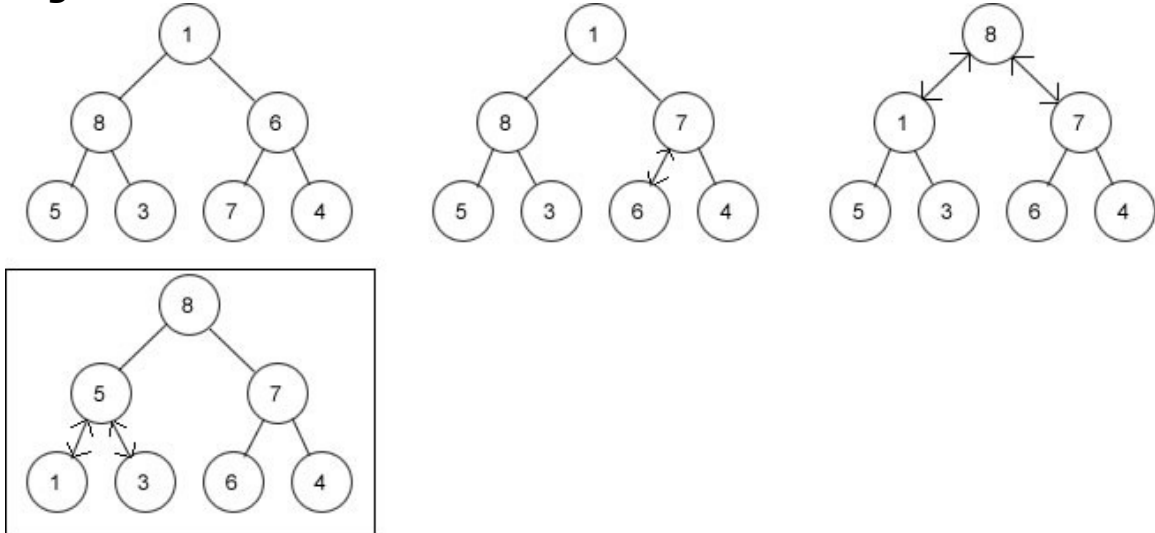
		e	x	p	o	n	e	n	t	i	a	l
	0	1	2	3	4	5	6	7	8	9	10	11
p	1											
o	2											
l	3											
y	4											
n	5											
o	6											
m	7											
i	8											
a	9											
l	10											

When the algorithm is complete:

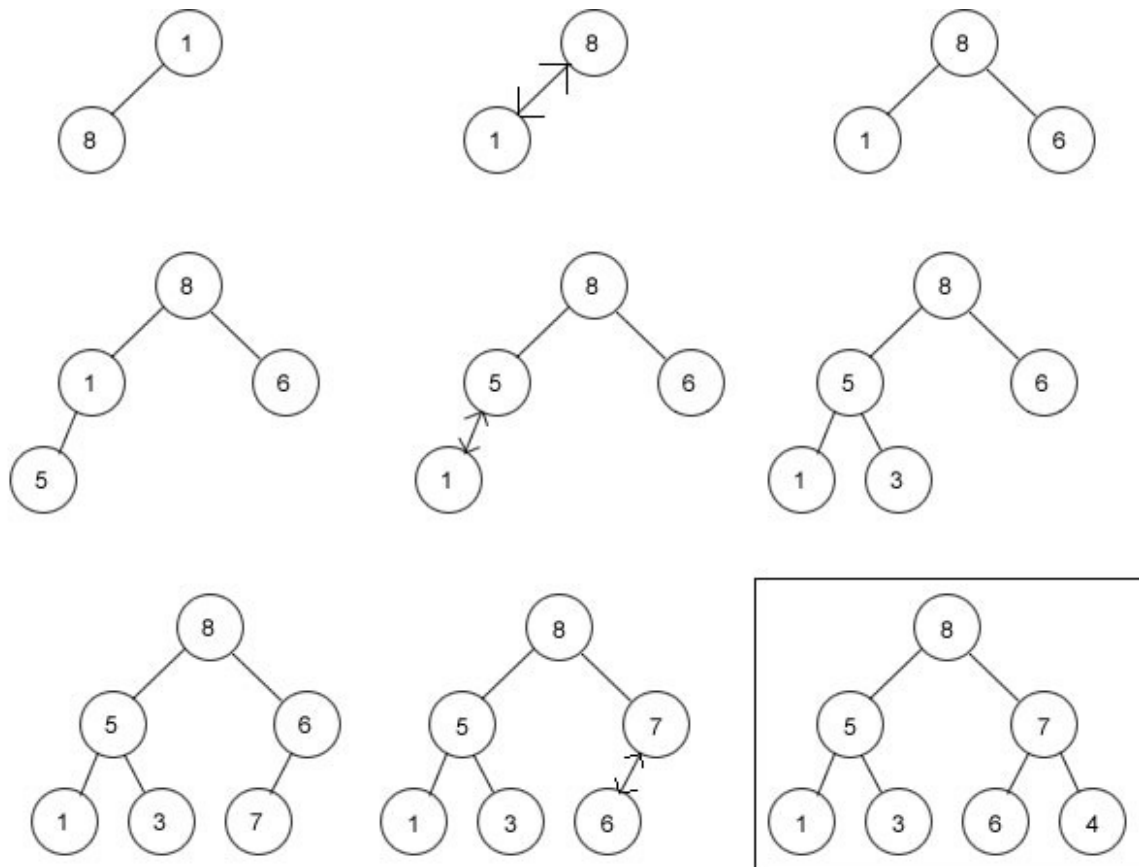
		e	x	p	o	n	e	n	t	i	a	l
	0	1	2	3	4	5	6	7	8	9	10	11
p	1	1	2	2	3	4	5	6	7	8	9	10
o	2	2	2	3	2	3	4	5	6	7	8	9
l	3	3	3	3	3	3	4	5	6	7	8	8
y	4	4	4	4	4	4	4	5	6	7	8	9
n	5	5	5	5	5	4	5	4	5	6	7	8
o	6	6	6	6	5	5	5	5	5	6	7	8
m	7	7	7	7	6	6	6	6	6	6	7	8
i	8	8	8	8	7	7	7	7	7	6	7	8
a	9	9	9	9	8	8	8	8	8	7	6	7
l	10	10	10	10	9	9	9	9	9	8	7	6

The edit distance between exponential and polynomial is 6.

6a. Construct a heap for the list 1, 8, 6, 5, 3, 7, 4 by the bottom-up algorithm



6b. Construct a heap for the list 1, 8, 6, 5, 3, 7, 4 by successive key insertions (top-down algorithm)



7. Outline an algorithm for checking whether an array $H[1...n]$ is a heap and determine its time efficiency.

```
// initially call isHeap with the array h and 1
// precondition: the array has no value set, or an
// abstract value set, in position 0

isHeap(array h, index i) {

    // base case, if we've reached the end of the parents
    // without returning false, it is a heap, return true

    if(i > h.length/2) return true;

    // verify that i is >= its children

    // we must first verify that the right child of i exists
    if(h[i*2+1] != NULL) {

        // right child of i exists
        if(h[i] >= h[i*2] && h[i] >= h[i*2+1])
            isHeap(h, i+1);
        else
            return false; // not a heap
    }

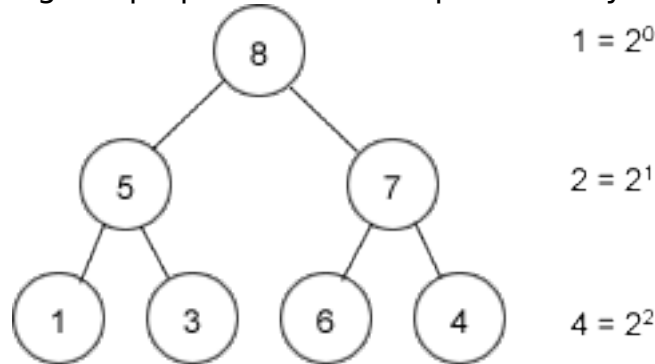
    else {

        // right child of i does not exist
        if(h[i] >= h[i*2])
            isHeap(h, i+1);
        else
            return false; // not a heap
    }
} // end isHeap
```

The isHeap algorithm will recur, at most, $n/2$ times where n is the length of the array. Each recurrence is C time, $n/2$ is $O(n)$. Thus this is a $O(n)$ algorithm.

8b) Prove that the height of a heap with n nodes is equal to $\lfloor \log_2(n) \rfloor$

For i = the current depth in the heap
The max nodes per level = 2^i
(using the properties of a complete binary tree)



By summation formula #5, Page 470 (Levitin), the max amount of nodes n in a tree of height h is equal to:

$$\sum_{i=0}^h 2^i = 2^{h+1} - 1 = n$$

Add one to the right side to make an inequality:

$$n < 2^{h+1}$$

2^h is less than or equal to n since 2^{h+1} is strictly greater than n

$$2^h \leq n < 2^{h+1}$$

Take the log of all three:

$$\log(2^h) \leq \log(n) < \log(2^{h+1})$$

By log law #3, Page 469 (Levitin):

$$\log(2^h) = h$$

$$\log(2^{h+1}) = h + 1$$

On a number line, h is less than or equal to $\log(n)$ and $h+1$ is strictly greater than $\log(n)$:

$$h \leq \log(n) < h + 1$$

Taking the floor of $\log(n)$ makes $\log(n) = h$

$$\text{Thus: } h = \lfloor \log_2(n) \rfloor$$

9. Prove the following equality:

$$\sum_{i=0}^{h-1} 2(h-i)2^i = 2(n - \log_2(n+1)) \quad \text{where} \quad n = 2^{h+1} - 1$$

Express the right-hand-side as a function of h:

$$\sum_{i=0}^{h-1} 2(h-i)2^i = 2(2^{h+1} - 1 - \log_2(2^{h+1}))$$

Simplify the summation through distribution:

$$\sum_{i=0}^{h-1} 2(h-i)2^i = \sum_{i=0}^{h-1} (2h - 2i)2^i = \sum_{i=0}^{h-1} 2h2^i - 2i2^i$$

By Sum Manipulation Rule #2, Page 470 (Levitin)

$$\sum_{i=0}^{h-1} 2h2^i - 2i2^i = \sum_{i=0}^{h-1} 2h2^i - \sum_{i=0}^{h-1} 2i2^i$$

By Sum Manipulation Rule #1, Page 470 (Levitin)

$$\sum_{i=0}^{h-1} 2h2^i = 2h \sum_{i=0}^{h-1} 2^i$$

By Summation Formula #6, Page 470 (Levitin)

$$\sum_{i=0}^{h-1} 2i2^i = 2[(h-2)2^h + 2]$$

By Summation Formula #5, Page 470 (Levitin)

$$2h \sum_{i=0}^{h-1} 2^i = 2h(2^h - 1)$$

Thus:

$$\begin{aligned} \sum_{i=0}^{h-1} 2h2^i - \sum_{i=0}^{h-1} 2i2^i &= \\ 2h(2^h - 1) - 2[(h-2)2^h + 2] \end{aligned}$$

Simplify:

$$2h(2^h - 1) - 2[(h-2)2^h + 2]$$

Factor out 2:

$$2[h(2^h - 1) - ((h-2)2^h + 2)]$$

Distribute:

$$2(h2^h - h - h2^h + 2^{h+1} - 2)$$

Cancel like terms:

$$2(-h + 2^{h+1} - 2)$$

We are done simplifying, put the simplified left term = the un-simplified right term

$$2(-h + 2^{h+1} - 2) = 2(2^{h+1} - 1 - \log_2(2^{h+1}))$$

By log law #3, Page 469 (Levitin):

$$\log_2(2^{h+1}) = (h+1)\log(2)$$

By log law #2, Page 469 (Levitin):

$$\log_2(2) = 1$$

Thus:

$$\log_2(2^{h+1}) = h+1$$

Put h+1 back into the equation for $\log_2(2^{h+1})$:

$$2(-h + 2^{h+1} - 2) = 2(2^{h+1} - 1 - (h+1))$$

Distribute the negative on the right side:

$$2(-h + 2^{h+1} - 2) = 2(2^{h+1} - 1 - h - 1)$$

Combine like terms and reorder:

$$2(2^{h+1} - h - 2) = 2(2^{h+1} - h - 2)$$

The two sides are equal, thus $\sum_{i=0}^{h-1} 2(h-i)2^i = 2(n - \log_2(n+1))$ is true.