

Homework 3 - CS317 Spring 2019

Matt Fletcher

April 18, 2019

1. The goal of dynamic programming is to create polynomial time solutions to problems by holding solutions to previous subproblems in a table to be re-used in future uses of that subproblem. A problem is said to have overlapping subproblems if it can be broken down into subproblems which are used multiple times. Optimal substructure is when the globally optimal solution can be constructed from locally optimal solutions to subproblems.

The four steps to dynamic programming approach are:

- (a) Characterize the structure of an optimal solution.
- (b) Recursively define the value of an optimal solution by expressing the solution of the original problem in terms of optimal solutions for smaller problems.
- (c) Bottom up computation, compute the value of an optimal solution in a bottom-up fashion by using a table structure.
- (d) Construct the optimal solution by expressing the solution of the original problem in terms of optimal solutions for smaller subproblems.

-
2. (a) Step 1 is finding the optimal substructure and using it to construct an optimal solution to the subproblems. By finding an optimal parenthesization of a subset of the matrices, we can use that subset to find an optimal parenthesization of the full set of matrices.
(b) Now, define the cost of an optimal solution recursively in terms of the optimal solution to subproblems.
(c) Compute the optimal costs. Without dynamic programming, this algorithm would take exponential time. We can compute the optimal costs using a tabular method. The Matrix Chain order defined in the text results in an $O(n^3)$ algorithm instead of an exponential algorithm that would result from a brute force approach.
(d) The Matrix Chain order defined in the previous problem determines the number of multiplications required but not directly how to multiply them. The table produced gives us the required information.

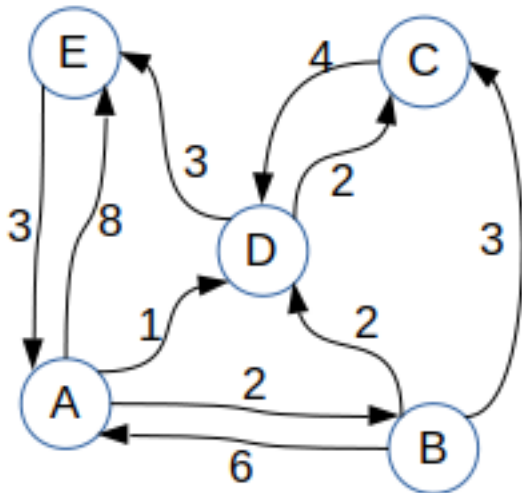
-
3. (a) A brute force solution to this problem would result in an exponential algorithm. The LCS problem has an optimal substructure property, however. In other words, a solution to a subproblem can be re-used in further steps.
(b) The recursive solution to the LCS problem involves establishing a recurrence for the value of an optimal solution. This is the LCS of X_{i-1} and Y_{i-1} if the two characters at matching locations are equal. Otherwise, find the LCS of X and Y_{i-1} .
(c) The algorithm takes 2 sequences X and Y and stored them in a table. This table is filled out in row-major order. Each cell in the table points to the table entry containing the optimal subproblem solution. This allows a solution to be found after filling out the table.
(d) The table filled with the arrows allows for a construction of the LCS. Trace through the table following the arrows, and whenever a diagonal arrow is encountered, it implied that that particular row/column is a solution of the LCS. Hence, a solution is easily found.

	y	0	1	0	1	1	0	1	1	0
x	0	0	0	0	0	0	0	0	0	0
1	0	0 [^]	1\	1<	1\	1\	1<	1\	1\	1<
0	0	1\	1 [^]	2\	2<	2<	2\	2<	2<	2\
0	0	1\	1 [^]	2\	2 [^]	2 [^]	3\	3<	3<	3\
1	0	1 [^]	2\	2 [^]	3\	4\	4<	4\	4\	4<
0	0	1\	2 [^]	2\	3 [^]	4 [^]	5\	5<	5<	5\
1	0	1 [^]	2\	2 [^]	3\	4\	5 [^]	6\	6<	6<
0	0	1\	2 [^]	2\	3 [^]	4 [^]	5\	6 [^]	6 [^]	6\
1	0	1 [^]	2\	3 [^]	4\	5\	5 [^]	6\	7\	7 [^]

The longest common subsequence is 7, as traced in the following image.

	y	0	1	0	1	1	0	1	1	0
x	0	0	0	0	0	0	0	0	0	0
1	0	0 [^]	1\	1<	1\	1\	1<	1\	1\	1<
0	0	1\	1 [^]	2\	2<	2<	2\	2<	2<	2\
0	0	1\	1 [^]	2\	2 [^]	2 [^]	3\	3<	3<	3\
1	0	1 [^]	2\	2 [^]	3\	4\	4<	4\	4\	4<
0	0	1\	2 [^]	2\	3 [^]	4 [^]	5\	5<	5<	5\
1	0	1 [^]	2\	2 [^]	3\	4\	5 [^]	6\	6<	6<
0	0	1\	2 [^]	2\	3 [^]	4 [^]	5\	6 [^]	6 [^]	6\
1	0	1 [^]	2\	3 [^]	4\	5\	5 [^]	6\	7\	7 [^]

4. Begin by visualizing the graph.



Now, write the beginning table.

	A	B	C	D	E
A	0	2	∞	1	8
B	6	0	3	2	∞
C	∞	∞	0	4	∞
D	∞	∞	2	0	3
E	3	∞	∞	∞	0

Update the table by seeing if any paths are shortened by going through A.

	A	B	C	D	E
A	0	2	∞	1	8
B	6	0	3	2	14
C	∞	∞	0	4	∞
D	∞	∞	2	0	3
E	3	5	∞	4	0

Now do the same through B.

	A	B	C	D	E
A	0	2	5	1	8
B	6	0	3	2	14
C	∞	∞	0	4	∞
D	∞	∞	2	0	3
E	3	5	∞	4	0

Node C: Nothing is optimized, table is unchanged.

	A	B	C	D	E
A	0	2	5	1	8
B	6	0	3	2	14
C	∞	∞	0	4	∞
D	∞	∞	2	0	3
E	3	5	∞	4	0

	A	B	C	D	E
A	0	2	3	1	4
B	6	0	3	2	5
C	∞	∞	0	4	7
D	∞	∞	2	0	3
E	3	5	6	4	0

	A	B	C	D	E
A	0	2	3	1	4
B	6	0	3	2	5
C	10	12	0	4	7
D	6	8	2	0	3
E	3	5	6	4	0

-
5. The knapsack problem is as follows: "Given n items of weights w_i and values v_j , and a knapsack of capacity W , find the most valuable subset of the items that fit into the knapsack. The dynamic approach to solving the knapsack problem consists of breaking the set of all the subsets of the first i items that fit the knapsack of capacity J into 2 categories: those that do not include the i^{th} item and those that do.

Use a table to reduce the number of recalculations needed.

	0	1	2	3	4	5	6
0	0	0	0	0	0	0	0
1	0	0	0	25	25	25	25
2	0	0	20	25	25	45	45
3	0	15	20	35	40	45	45
4	0	15	20	35	40	55	60
5	0	15	20	35	40	55	65

Items 3 and 5 will create the optimal solution of \$65 in the knapsack.

6. Class P problems are those that have a polynomial time solution. This includes problems such as sorting a list or finding the largest element. Problems in class NP have non-polynomial time solutions but are able to be checked in polynomial time. For example, the traveling salesman problem can only be solved in non-polynomial time. The P-NP problem asks whether every problem which has a solution that can be checked in polynomial time also has a solution that can be solved in polynomial time. If P is shown to be equal to NP, then one major effect will be that modern RSA encryption will be easily able to be broken, given that RSA encryption relies on a solution only being able to be found in NP time.

Problems are classified into different complexity classes in order to understand how to approach them. These complexity classes are important because it lets us know if we can use that particular problem in a certain way. For example, prime factorization is an NP problem. Because of this, we can use it as the basis for RSA encryption. If it was not an NP problem (aka it had a solution solveable in P time) then it would not be a good choice.

7. The backtracking technique is stopping a particular branch when a solution using that approach is no longer possible. In the N-Queens problem, n queens must be placed on an $n \times n$ board such that no 2 queens are in the same row, column, or diagonal. To use backtracking, stop attempting to solve a board when it is impossible for the next queen to be placed.

Q	-	-	-
—	\		
—		\	
—			\

Q	-	-	-
—	\	Q	-
—	/	\	\
—		—	\

Not enough squares remain. Go back and try the next square for queen 2.

Q	-	-	-
—	\	-	Q
—		\	—
—	/		\

Q	-	-	-
—	\	-	Q
—	Q	\	—
—	/	\	\

Not enough squares remain. Go back to the first queen and try the next square.

-	Q	-	-
/	—	\	
	—		\
	—		

-	Q	-	-
/	—	\	Q
	—	/	\
	—		—

-	Q	-	-
/	—	\	Q
Q	—	/	\
—	—		—

-	Q	-	-
/	—	\	Q
Q	—	/	\
—	—	Q	—

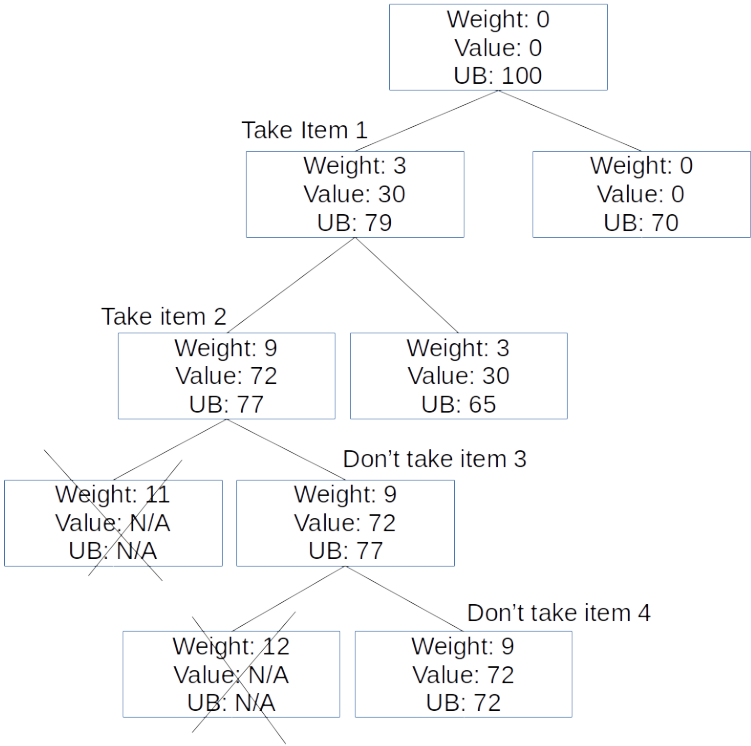
As all 4 queens have been placed, and no 2 are in the same row, column, or diagonal, this is the final solution to the problem.

8. Begin by sorting the table on the weight/value ratio.

Item	Weight	Value	Val/Weight
1	3	30	10
2	6	42	7
3	2	10	5
4	3	15	5

Now, for each decision, calculate the upper bound with the following equation:

$$ub = v + (W - w)(v_{i+1}/w_{i+1})$$



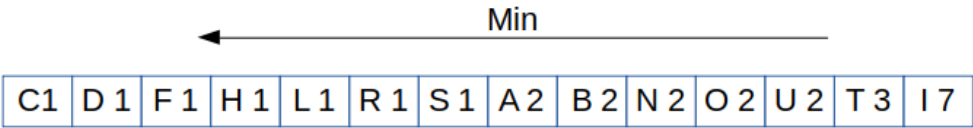
Therefore, the optimal solution takes items 1 and 2, with a value of 72.

9. The given word is Honorificabilitudinitatibus.

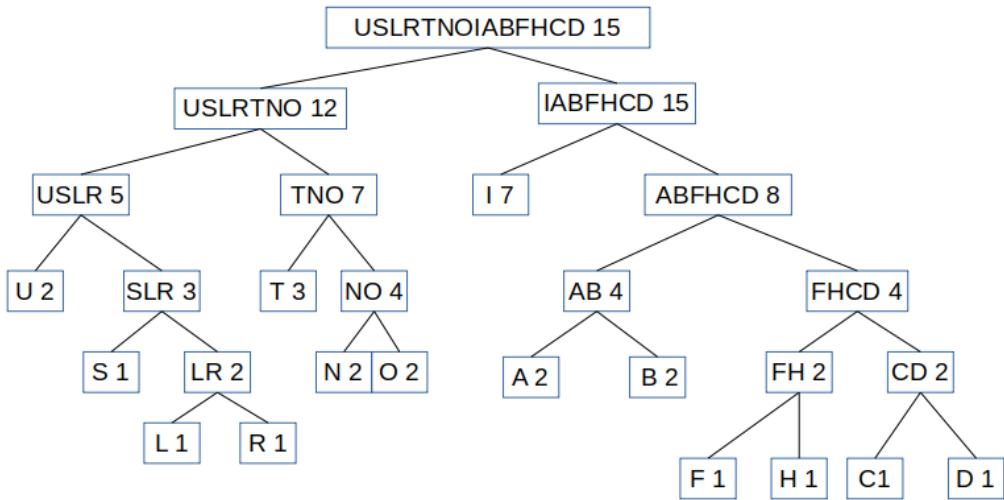
This table contains the frequencies of each letter.

i	t	u	o	n	b	a	s	r	l	h	f	d	c
7	3	2	2	2	2	2	1	1	1	1	1	1	1

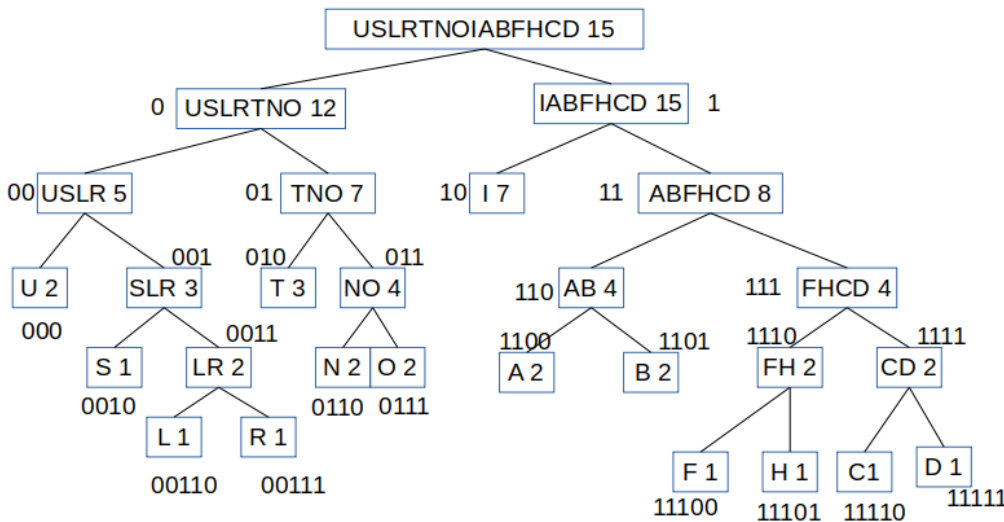
Now, begin by placing these in a minimum priority queue.



Take the first 2 elements out of the queue, place them into a binary tree, and place the root node of that tree back into the minimum priority queue, putting it as far forward in the queue as it will go.
Repeat this step until all nodes have been placed in the tree.



Now, go down the tree. Every time you travel left, put a 0. Every time you go right, put a 1.



For each of the leaf nodes, this represents the encoding used for that character.

h	o	n	r	i	f	c	a	b	l	t	u	d	s
11101	0111	0110	00111	10	11100	11110	1100	1101	00110	010	000	11111	0010

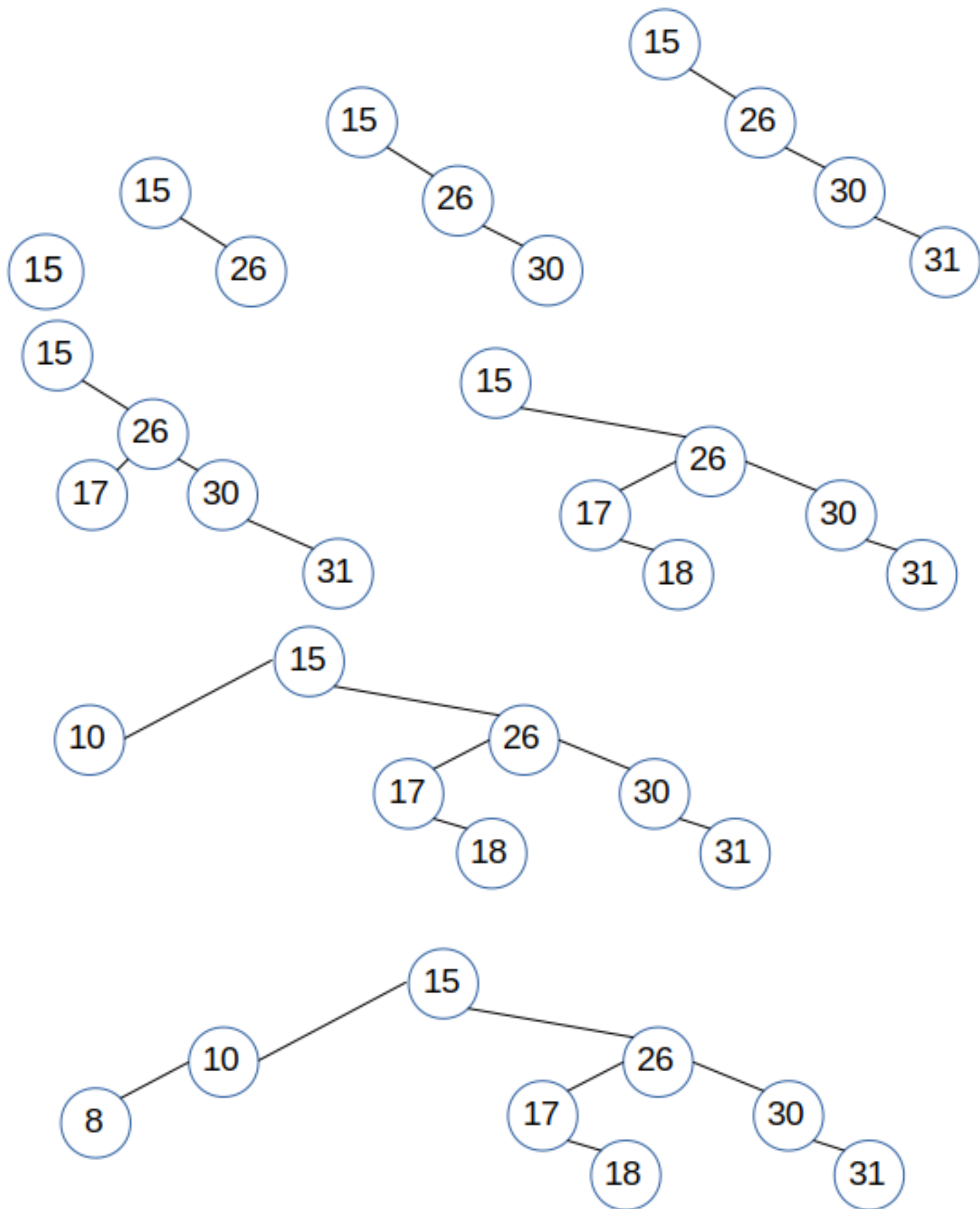
Now, replace each letter in the word with its corresponding value.

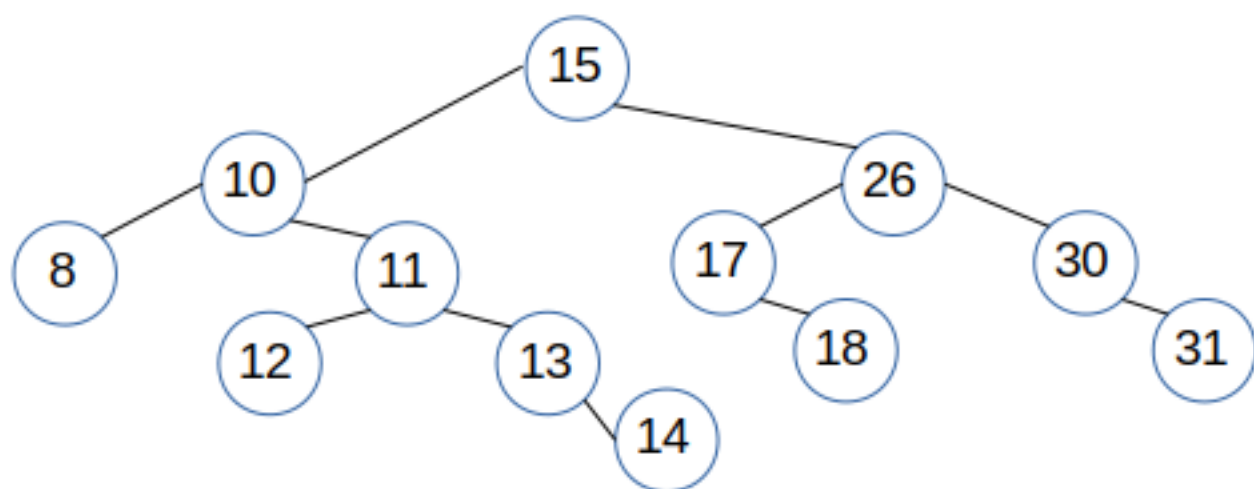
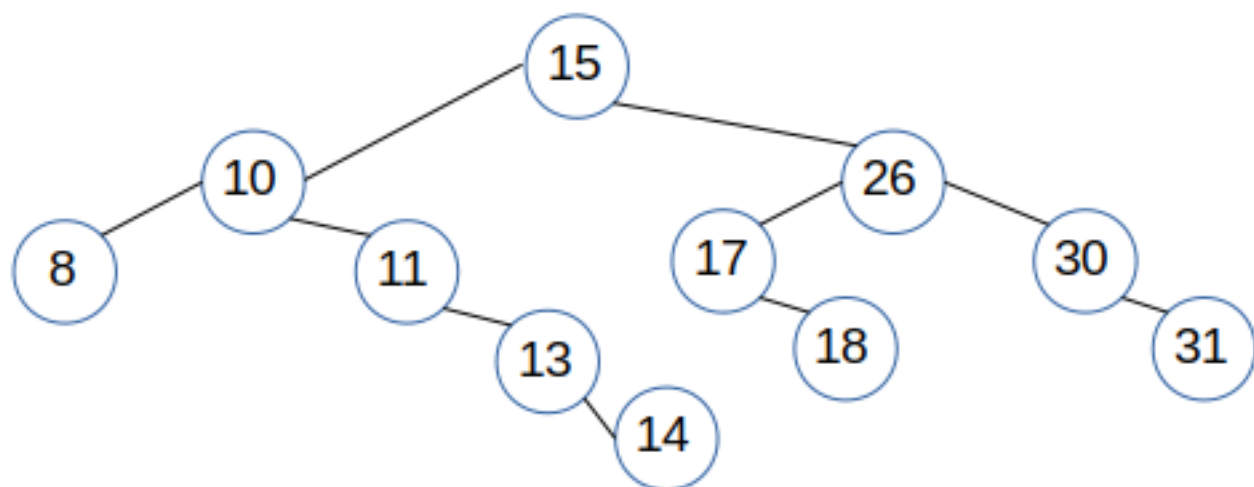
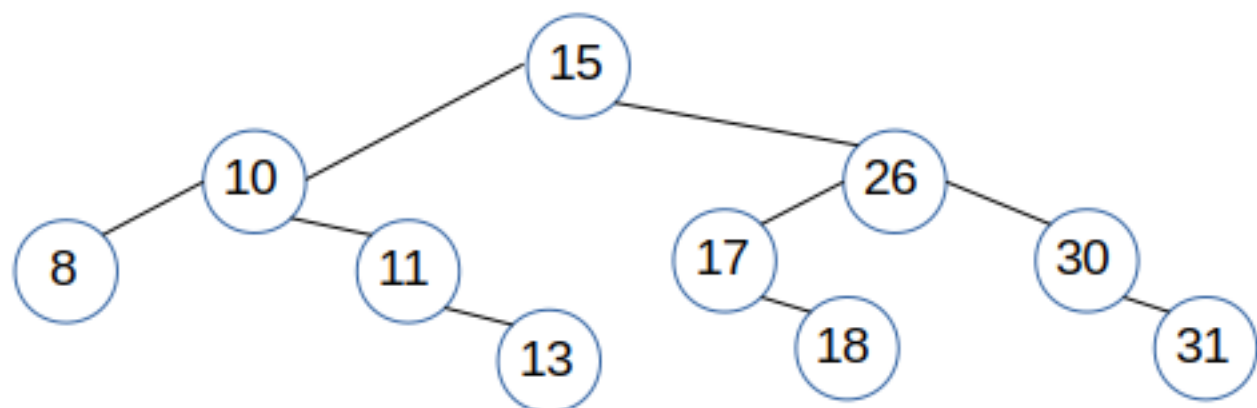
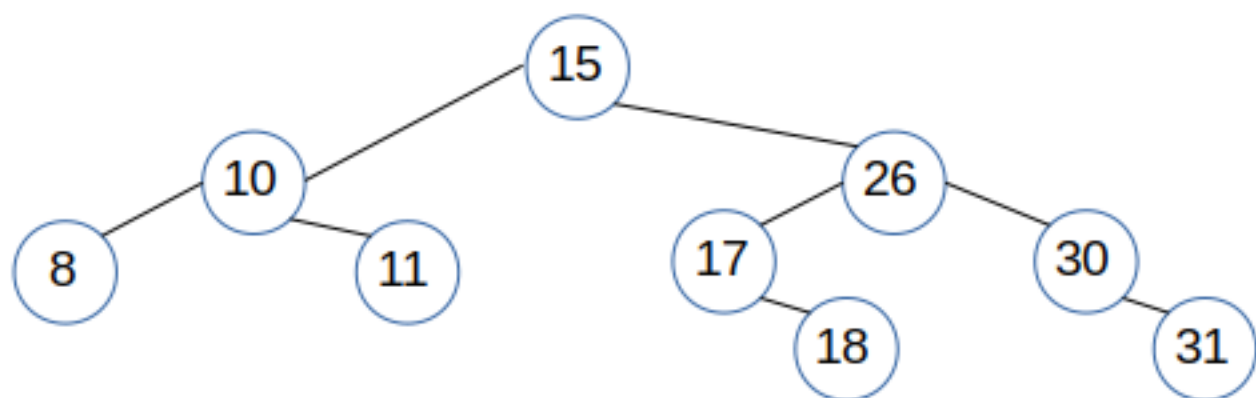
11101011101100111001111011100101111011001101100011010010000111111001101001011000101011010000010

This is 95 characters, compared to the ASCII encoding of $27 \times 8 = 216$

Therefore, the encoding ratio is $\frac{95}{216} = 43.98\%$

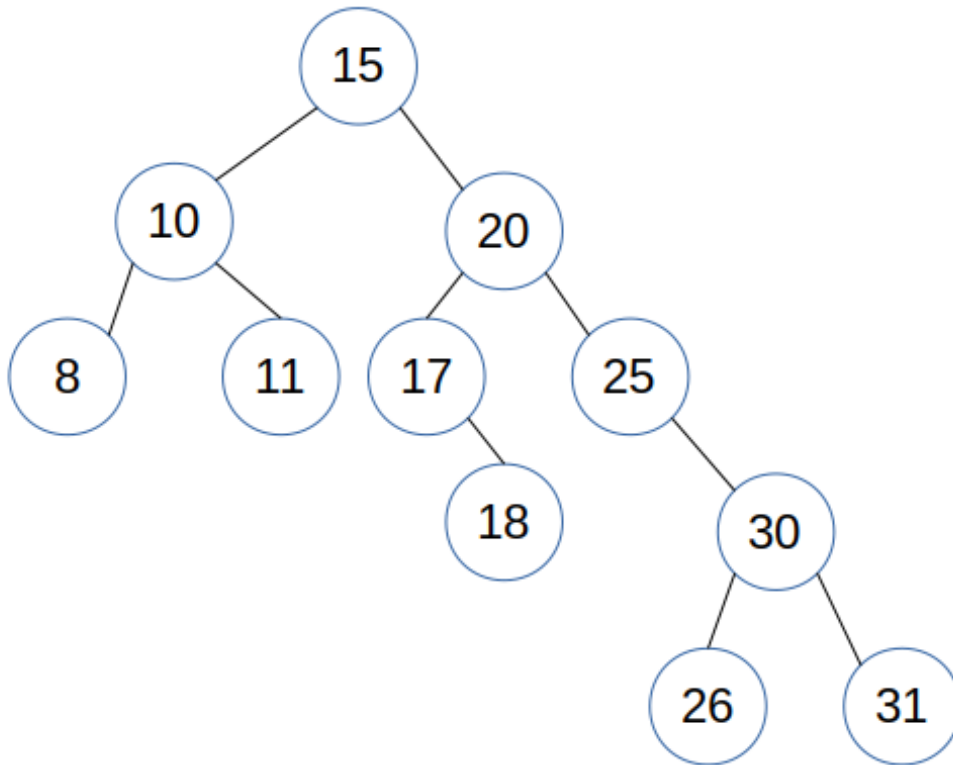
10. The following images show the construction of a binary search tree given the input list.



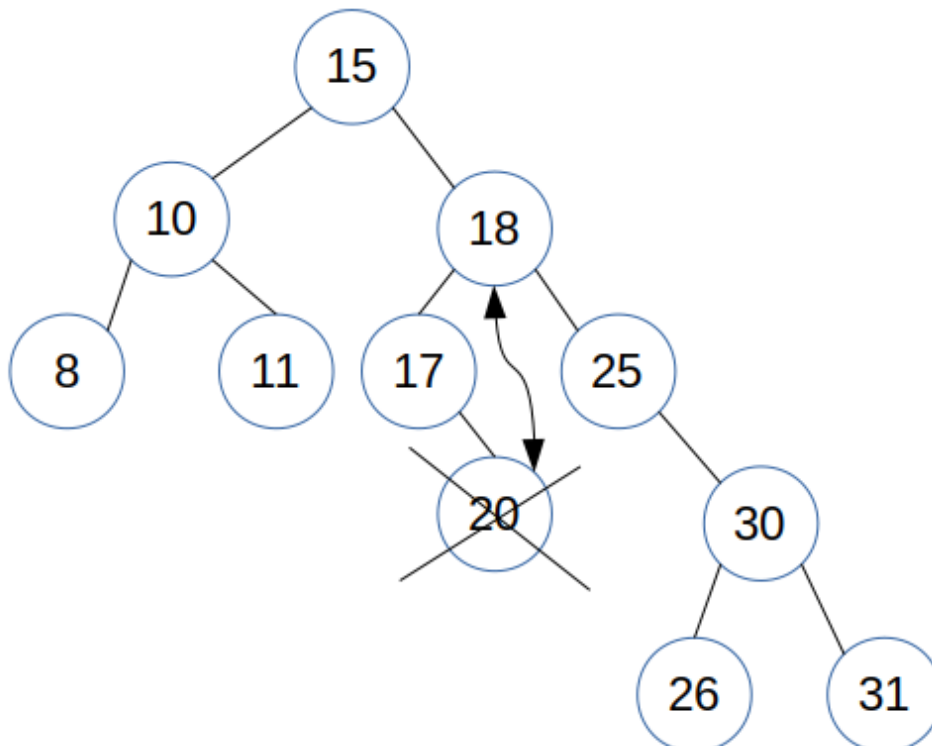


A BST insertion and deletion is least optimal when the input list is ordered. This turns the tree from a binary tree into a linked list. This changes the efficiency from $O(\log n)$ to $O(n)$. A worst case input would be 1, 2, 3, 4, 5.

11. Here is the original tree.

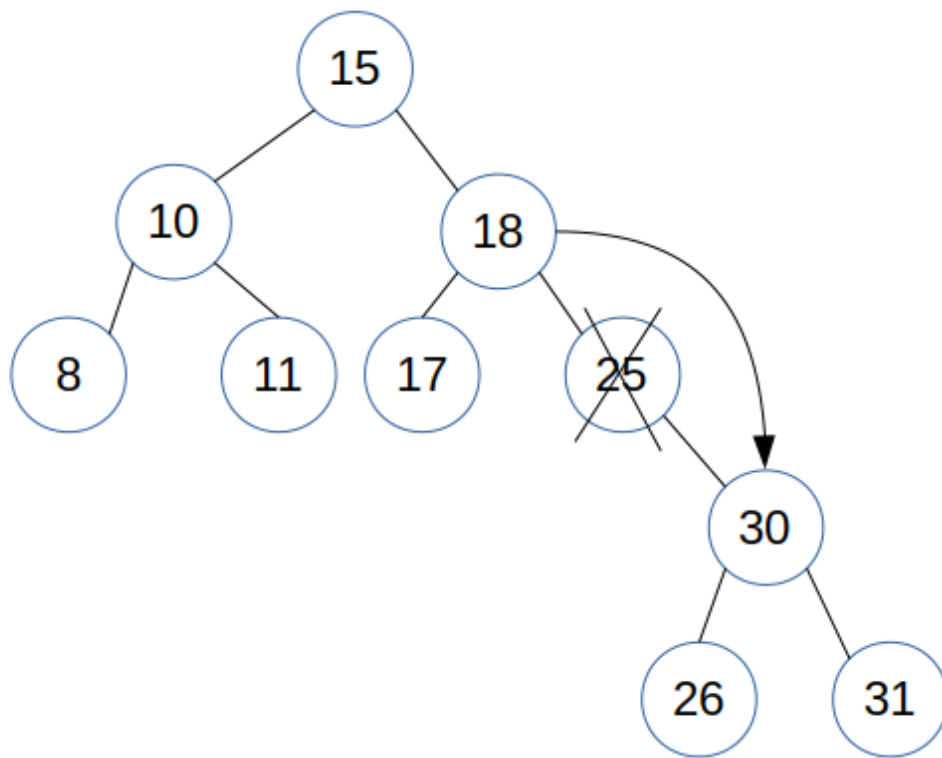


We wish to delete node #20. Because it has 2 children, we find the largest element in the left subtree, swap it with 20, and then delete that node.



Now, we wish to delete node #25. Because it only has 1 child, we simply point the parent node to point towards the

child of #25, as shown in the figure below:



In the end, the tree looks as follows:

