

Practice problem : Dynamic Programming

Task 1: Rod cutting problem

A company buys long steel rods (of length n), and cuts them into shorter one to sell. Here,

- lengths are integers only
- cutting is free
- rods of diff lengths sold for diff. price, e.g.,

length i	1	2	3	4	5	6	7	8	9	10
price p_i	1	5	8	9	10	17	17	20	24	30

What is the best way to cut the rods?

- $n=4$: no cutting: \$9, 1 and 3: $1+8=\$9$, 2 and 2: $5+5=\$10$. So, 2 and 2 is be the best.
- Now how about $n=5$: ?

Explanation and solution have been discussed in theory class. You can refer to the slide included.

Brute force recursion algorithm:

```
Cut_Rod(p, n) {  
    // base case  
    if n == 0  
        return 0  
    // recursive case  
    q =  $-\infty$   
    for i = 1 to n {  
        q = max(q,  $p[i] + \text{Cut-Rod}(p, n - i)$ )  
    }  
    return q  
}
```

But this is a $O(2^n)$ solution because of repetitive subproblems. To avoid solving repetitive subproblems we use memoization. Add a global array named r to keep the solution of the subproblems.

Recursion with memoization algorithm:

```
init_table(n) {  
    // initialize memo (an array r[] to keep max revenue)  
    r[0] = 0  
    for i = 1 to n  
        r[i] =  $-\infty$  // r[i] = max revenue for rod with length=i  
}
```

```
Memoized_Cut_Rod(p, n) {  
    if n == 0  
        return r[n] // return the saved solution  
    q =  $-\infty$ 
```

```

    for i = 1 to n {
        if r[n-i] == -∞ {
            r[n-i] = Memoized-Cut-Rod (p, n-i)
        }
        q = max(q, p[i] + r[n-i])
    }
    r[n] = q // update memo
    return r[n]
}

```

Dynamic programming algorithm:

```

DP_Cut-Rod(p, n) {
    r[0] = 0
    for j = 1 to n { // compute r[1], r[2], ... in order
        q = -∞
        for i = 1 to j {
            q = max(q, p[i] + r[j - i])
        }
        r[j] = q
    }
    return r[n]
}

```

Now, you may want to know how the pieces were cut to produce the maximum revenue. For this we have to extend the program. Add another global array named *cut* to keep the cutting positions.

```

Extended_DP_Cut_Rod(p, n) {
    r[0] = 0
    for j = 1 to n { //compute r[1], r[2], ... in order
        q = -∞
        for i = 1 to j {
            if q < p[i] + r[j - i] {
                q = p[i] + r[j - i]
                cut[j] = i // the best first cut for len j rod
            }
        }
        r[j] = q
    }
    return r[n]
}

```

Finally print the solution using the following function:

```

Print_Cut_Rod_Solution(p, n) {
    i = n
    while i > 0 {
        print cut[i]
        i = i - cut[i] // remove the first piece
    }
}

```

Task 2: 0/1 knapsack problem

Given weights and values of n items, put these items in a knapsack of capacity W to get the maximum total value in the knapsack. In other words, given two integer arrays $val[0..n-1]$ and $wt[0..n-1]$ which represent values and weights associated with n items respectively. Also given an integer W which represents knapsack capacity, find out the maximum value subset of $val[]$ such that sum of the weights of this subset is smaller than or equal to W . You cannot break an item, either pick the complete item or don't pick it (0-1 property).

Explanation and solution have been discussed in theory class.

Brute force recursion algorithm:

```
knapsack(W, v, wt, n) {
    // base case
    if n == 0 || W == 0
        return 0
    // recursive case
    if W - wt[n] < 0 {
        q = knapsack(W, v, wt, n-1)
    }
    Else {
        a = v[n] + knapsack(W - wt[n], v, wt, n-1)
        b = knapsack(W, v, wt, n-1)
        q = max(a, b)
    }
    return q
}
```

But this is a $O(2^n)$ solution because of repetitive subproblems. To avoid solving repetitive subproblems we use memoization. Add a global 2-D array named *table* to keep the solution of the subproblems.

Recursion with memoization algorithm:

```
init_table(n, W) {
    // initialize memo (a 2 -D array to keep max profit)
    for i = 1 to n {
        table[i][0] = 0
    }
    for i = 0 to W {
        table[0][i] = 0
    }

    for i = 1 to n {
        for j = 1 to W {
            table[i][j] = -∞
        }
    }
}
```

```

Memoized_knapsack(W, v, wt, n) {
    if (n == 0 || W == 0)
        return table[n][W] // return the saved solution

    if W - wt[n] < 0 {
        if table[n-1][W] == -∞ {
            table[n-1][W] = knapsack(W, v, wt, n-1)
        }
        q = table[n-1][W]
    }
    else {
        if table[n-1][W] == -∞ {
            table[n-1][W] = knapsack(W, v, wt, n-1)
        }
        if table[n-1][W-wt[n]] == -∞ {
            table[n-1][W-wt[n]] = knapsack(W-wt[n], v, wt, n-1)
        }
        a = v[n] + table[n-1][W-wt[n]]
        b = table[n-1][W]
        q = max(a, b)
    }
    table[n][W] = q // update memo
    return table[n][W]
}

```

Dynamic programming algorithm:

```

DP_knapsack(W, v, wt, n){
    for i = 1 to n {
        table[i][0] = 0
    }
    for i = 0 to W {
        table[0][i] = 0
    }
    for i = 1 to n {
        for j = 1 to W {
            if j - wt[i] < 0 {
                table[i][j] = table[i-1][j]
            }
            else {
                a = v[i] + table[i-1][j - wt[i]]
                b = table[i-1][j]
                table[i][j] = max(a, b)
            }
        }
    }
    return table[n][W]
}

```

Now, you may want to know how the pieces were cut to produce the maximum revenue. For this we have to extend the program. Add another global array named *choice* to keep the cutting positions.

```

Extended_DP_knapsack(W, v, wt, n){
    for i = 1 to n {
        table[i][0] = 0
    }
    for i = 0 to W {
        table[0][i] = 0
    }
    for i = 1 to n {
        for j = 1 to W {
            if j - wt[i] < 0 {
                table[i][j] = table[i-1][j]
                choice[i][j] = 0
            }
            else {
                a = v[i] + table[i-1][j - wt[i]]
                b = table[i-1][j]
                if a > b {
                    table[i][j] = a
                    choice[i][j] = 1
                }
                else {
                    table[i][j] = b
                    choice[i][j] = 0
                }
            }
        }
    }
    return table[n][W]
}

```

Finally print the solution using the following function:

```

Print_knapsack_solution(W, v[], wt[], n) {
    i = n
    j = W
    while i > 0 {
        if choice[i][j] == 1 {
            print "item i taken"
            j = j - wt[i]
            i = i - 1
        }
        else {
            print "item i not taken"
            i = i - 1
        }
    }
}

```