

Dynamic Programming (DP)

DP = subproblems + reuse

DP = recursion + memorization

time = no. of subproblems * time per subproblem (treating recursive calls as $O(1)$)

Fabonacci numbers: $F_1 = F_2 = 1, F_n = F_{n-1} + F_{n-2} — O(n)$

Naive algorithm:

fib(n):

```
    if n ≤ 2:
        f = 1
    else:
        f = fib(n-1) + fib(n-2)
    return f
```

Memorized DP algorithm:

memo = {}

fib(n):

```
    if n in memo:
        return memo(n)
    if n ≤ 2:
        f = 1
    else:
        f = fib(n-1) + fib(n-2)
    memo[n] = f
    return f
```

Runtime: $O(n)$

Bottom-up DP algorithm:

fib = []

for k = 1 to n:

```
    if k ≤ 2:
        f = 1
    else:
        f = fib[k-1] + fib[k-2]
```

```
return fib[n]
```

Here we use $O(n)$ memory, we can reduce it to $O(1)$.

Rod Cutting — $O(n^2)$

<i>length</i> — i	1	2	3	4	5	6	7	8	9
<i>price</i> — p_i	1	5	8	9	10	17	17	20	24
<i>revenue</i> — r_i	1	5	8	10	13	17	18	22	25

There are 2^{n-1} ways of cutting the rod, since we can each cut or not cut at each of the $n - 1$ positions.
(Considering permutations 1+1+2 and 1+2+1 are the same, $2^{\Theta(\sqrt{n})}$)

optimal revenue $r_n = p_i + r_{n-i}$

$r_n = \max(p_1 + r_{n-1}, p_2 + r_{n-2}, \dots, p_n + r_0)$ where we define $r_0 = 0$

CUTROD(n)

if $n = 0$:

 return 0

$q = -\infty$

for $i = 1$ to n :

$q = \max(q, p_i + \text{CUTROD}(n-i))$

return q

Runtime: $T(n) = 1 + \sum_{j=0}^{n-1} T(j) = 2^n$

CUTROD(n)

if n in memo:

 return memo[n]

if $n = 0$:

 return 0

$q = -\infty$

for $i = 1$ to n :

$q = \max(q, p_i + \text{CUTROD}(n-i))$

memo[n] = q

return q

Runtime: no. of subproblems * time per subproblem = $n \cdot \Theta(n) = \Theta(n^2)$

Bottom-up version

```

r[0] = 0
for j = 1 to n:
    q = - ∞
    for i = 1 to j:
        if q < p[i] + r[j-1]:
            q = p[i] + r[j-1]
            s[j] = i
    r[j] = q
m = n
while m > 0:
    print(s[m])
    m = m - s[m]
return r[n]

```

Remark: It is easy to also output the actual partition.

Another possible solution $r_n = \max_{i \leq j} (r_{i-1} + P_{j-i+1} + r_{n-j}), O(n^3)$

Longest Common Subsequence — $O(nm)$

Given strings $X = x_{1...m}$, $Y = y_{1...n}$, find the (length of) the longest common subsequence LCS

HYPERLINKING

DOLPHINSPEAK

Subproblems: for each prefix ($i=0, 1, \dots, m$) of X and each prefix ($j=0, 1, \dots, n$) of Y , what is LCS?

Optimal substructure: $X = x_{1...m}$, $Y = y_{1...n}$, $z = z_{1...k}$

1. If $x_m = y_n$, then $z_k = x_m = y_n$, then z_{j-1} is the LCS of x_{m-1} , y_{j-1}
2. If $x_m \neq y_n$, then
 - a. If $z_k \neq x_m$, then z_{j-1} is the LCS of x_{m-1} , y_j
 - b. If $z_k \neq y_n$, then z_{j-1} is the LCS of x_m , y_{n-1}

Proof 1. If $Z_k \neq X_m$, then we could append $x_m = y_n$ to end of Z , in contradiction.

Therefore $Z_k = X_m = y_n$. ~~Then~~ Z_{k-1} is a common subsequence of X_{m-1} and Y_{n-1} .

If it is not the longest, then there is another common subseq of length $\geq k$, and by appending $Z_k = X_m = y_n$, we get a common subseq of X and Y of len $\geq k+1$, in contradiction.

2a. Z is a common subseq of X_{m-1} and Y . If there is another common subseq of X_{m-1} & Y of length $> k$, it's also of X & Y , in contradiction.

2b. Similarly.

Recurrence:

$$C[i, j] = \begin{cases} 0 & i = j = 0 \\ C[i-1, j-1] + 1 & i, j > 0, x_i = y_j \\ \max(C[i-1, j], C[j-1, i]) & i, j > 0, x_i \neq y_j \end{cases}$$

Runtime: $O(nm)$ subproblems $\rightarrow O(nm)$

Knapsack — $O(nS)$

- list of n items each of size s_i and value v_i
- knapsack of size S
- find maximum value

1. Subproblems: items $1, \dots, i$ and remaining capacity $X \leq S$

$\theta(n \cdot S)$ subproblems

2. Guess: is item i in or not?

3. Recurrence: $DP(X, i) = \max(DP(X, i-1), DP(X - S_i, i-1) + v_i)$

4. Runtime: $\theta(n \cdot S)$