# Review

## The DSelect Algorithm

Elements that are > or < m

$$3\left(\lceil\frac{1}{2}\lceil\frac{n}{5}\rceil\rceil-2\right) \ge \frac{3n}{10}-6$$

Thus, the  $upper\ bound$  of the number of elements we put into the recursive call:

$$n - \left(\frac{3n}{10} - 6\right) = \frac{7n}{10} + 6$$

## Recurrence

With this in mind, our recursion complexity looks along the lines of

$$T(N) = \begin{cases} 1, N = 1 \\ T(n) \le cn + T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right) \end{cases}$$

## Hope and Check

Hope: There is some constant a [independent of a] such that  $T(n) \leq an, \forall n \geq 1$  If true, T(n) = O(n)

# Analysis

Claim: Let a = 10c, then  $T(n) \le an, \forall n \ge 1$ 

**Proof:** By induction on n.

Base Case:  $T(1) = 1 \le a \times 1 (n = 1, \text{, since } a \ge 1)$ 

Induction Step: n > 1

Induction Hypotheseis:  $T(k) \le a \times k$ , for k < n

We have 
$$T(n) \le cn + T(n/5) + T(\frac{7n}{10})$$
  
 $\le cn + a(n/5) + a\left(\frac{7n}{10}\right)$   
 $= n(c + \frac{9a}{10}) = an$ 

# **Dynamic Programming**

Dynamic programming is a method for designing efficient algorithms for recursively solvable problems with the following properties:

- 1. Optimal Substructure: An optimal solution to an instance contains an optimal solution to its sub-instances
- 2. Overlapping Subproblems: The number of subproblems is small so during the recursion same instances are referred to over and over again.

Basically fancy recursion

Four steps in solving a problem using dynamic programming:

- 1. Characterize the structure
- 2. Recursively define the value of a solution

3.

A more refined / intuitive list: <sup>1</sup>

- 1. Define subproblems
- 2. Write down the recurrence that relates the subproblems
- 3. Recognize and solve the best cases

#### **Fibonacci**

Sequence:  $1, 1, 2, 3, 5, 8, 13, \dots$ 

#### Naive Recursion Approach

```
def fib(n):
    if n <= 2:
        return 1
    return fib(n - 1) + fib(n - 2)</pre>
```

## Runtime

 $O(2^n)$  We calculate the same value multiple times.

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

 $<sup>^{1}</sup>$ Recursion + memorization

## Memoized DP Approach

```
memo = {}
def 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</pre>
```

## Bottom-Up DP Approach

```
fib = {}
for k in range(n):
    if k <= 2:
        f = 1
    else:
        f = fib[n - 1] + fib[n-2]
    fib[k] = f
return fib[n]</pre>
```

# Coin Change

Given n, find the number of different ways to write n as the sum of 1, 3, 4 Example: n = 5

$$5 = 1 + 1 + 1 + 1 + 1$$

$$= 1 + 1 + 3$$

$$= 1 + 3 + 1$$

$$= 3 + 1 + 1$$

$$= 1 + 4$$

$$= 4 + 1$$

$$5 = 1 + 1 + 1 + 1 + 1$$

$$= 3 + 1 + 1$$

$$= 1 + 3 + 1$$

$$= 4 + 1$$

$$D_4$$

$$= 1 + 1 + 3$$

## Define Subproblems

Let  $D_n$  be the number of ways to write n as the sum of 1, 3, 4

## Find the recurrence

- 1. Consider one possible solution  $n = x_1 + x_2 + \cdots + x_m$
- 2. If  $x_m = 1$ , the rest of the terms must sum to n 1
- 3. Thus, the number of sums that end with  $x_m=1$  is equal to  $\mathcal{D}_{n-1}$
- 4. Take other cases into account  $(x_m = 3, x_m = 4)$

$$D_n = D_{n-1} + D_{n-3} + D_{n-4}$$

Base Cases

$$D_0 = 1 D_1 = 1$$