

## Making Change

### Making Change

**Goal.** Given currency coin denominations, e.g., {100, 25, 10, 5, 1} devise a method to pay **an integer** amount using the **fewest coins**.

**Example:** 34¢.



25 5 1 1 1 1

**Cashier's algorithm.** At each iteration, add coin of the largest value that does not take us past the amount to be paid.

**Example:**  
\$2.89 = 289¢.



100, 100, 25, 25, 25, 10, 1, 1, 1, 1

## Greedy Algorithm

**Cashier's algorithm.** Use the maximal number of the largest denomination coin

```

x - amount to be changed
Sort coins denominations by value:  $c_1 < c_2 < \dots < c_n$ .
S ← empty ← coins selected
while (x > 0) {
  let k be largest integer such that  $c_k \leq x$ 
  if (k == 0) # all  $c_k > x$ 
    return "no solution found"
  x ← x -  $c_k$ 
  append(S, k)
}
return S

```

Does this Greedy algorithm always work?

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## Greedy doesn't always work

1. Greedy fails changing **30 optimally** with coin set **{25, 10, 1}** as it produces [25,1,1,1,1,1] instead of [10,10,10]
2. Greedy fails changing **30 at all** with coin set **{25, 10}** even though there is a solution: [10,10,10]
3. But the Greedy algorithm works for US coin set  
Proof: number theory (canonical coin systems)

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### Different problem: number of ways to pay

Given a sorted coin set  $\text{coins} = \{c_0, c_1, \dots, c_{d-1}\}$   $c_0$  the smallest coin value, and  $c_{d-1}$  the largest coin value, and an amount  $M$   
**how many different ways can  $M$  be paid?**

**One possible recursive either / or solution: go backwards through coins and choose to use the largest remaining coin or not**

**mkCh(n, c):**

**# n: amount still to be paid**

**# c: index of coins value currently considered**

**Base:**

if  $c == 0$ , how many ways? (is there always a way?)

**Step:**

if  $c > 0$

if largest coin cannot be used: consider  $\text{coin}_{c-1}$

else: # it can be used

**either** use one  $\text{coin}_c$  and keep considering  $\text{coin}_c$

**or** don't use  $\text{coin}_c$  and thus consider  $\text{coin}_{c-1}$

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### Make change vs. knapsack

Recurrence:

$\text{ways}(\text{amount}, i) =$

1. Base case?

2. If  $\text{amount} < \text{coin}[i]$ :  $\text{ways}(i-1, \text{amount})$

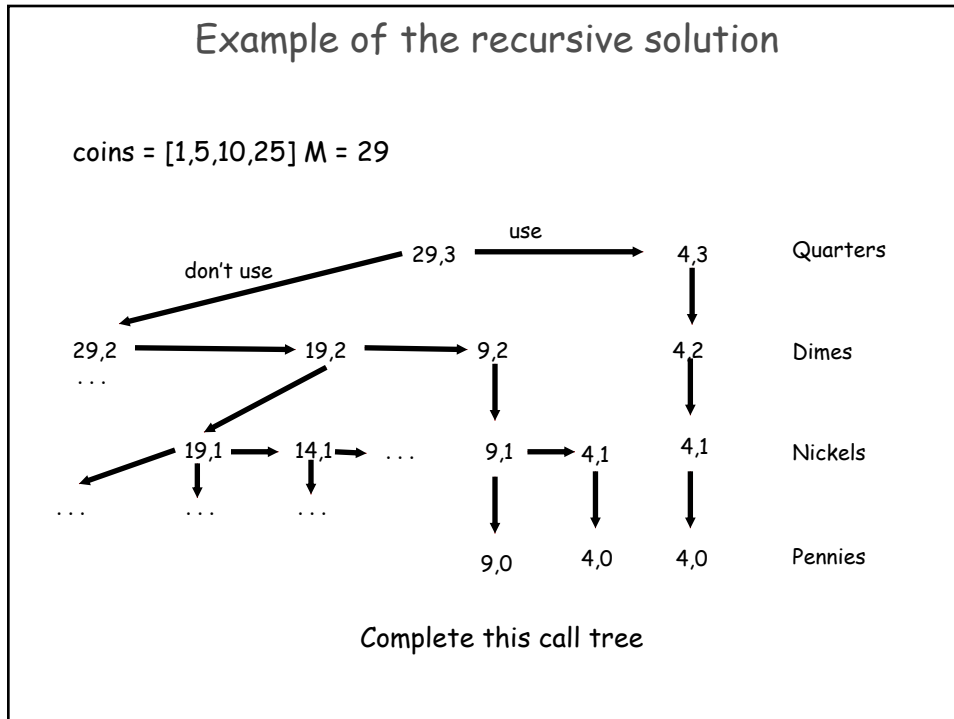
3. Else:  $\text{ways}(\text{amount} - \text{coin}[i], i) + \text{ways}(\text{amount}, i-1)$

Making change is very similar to knapsack, but:

1. We take the sum, not the maximum, of the two options.

2. We must use the same coin value a number of times. How this is reflected in the recurrence?

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### Making Change Dynamic Programming

Go through the state space bottom-up:  $i=0$  to  $n-1$

- select coin type
  - first 1 coin type, then 1&2, ....., finally all coin types
  - what does the first column look like?
- use solutions of smaller sub-problems to compute solutions of larger ones by storing previous values. Which values do you need to preserve?

In the recursive solution (DC) there are 2 (recursive) sub-problems. In the dynamic programming solution there are 2 reads:

don't use current coin ←

use current coin ↓

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## Programming Assignment

1. Write a recursive `mkChange` function based on the either or choices from slide 6, then turn it into a Dynamic Programming function.
  - Do you need a 2 D table here?
2. Determine the performance of the two algorithms. Later, in a written assignment, you will plot your data, and infer  $O$  complexity:
  - Recursive: count number of calls
  - Dynamic programming: count number of table reads

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