

 **BM40A1500 DATA STRUCTURES AND ALGORITHMS**

ALGORITHM DESIGN PRINCIPLES 2

2024

ALGORITHM DESIGN PRINCIPLES

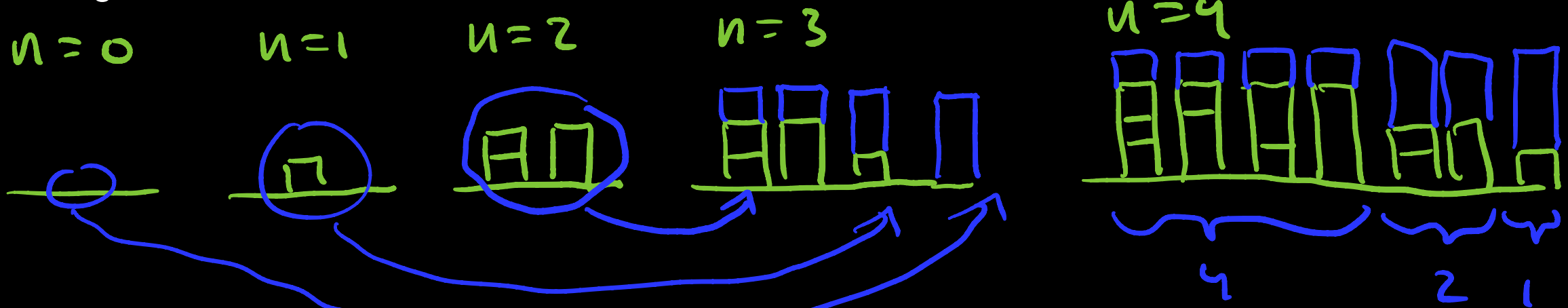
- ❖ Greedy approach
- ❖ Backtracking
 - ❖ Branch and bound
- ❖ Divide and conquer
- ❖ Dynamic programming
- ❖ Probabilistic algorithms
 - ❖ Las Vegas algorithms
 - ❖ Monte Carlo algorithms

DYNAMIC PROGRAMMING

- ❖ A way to improve the efficiency of any inherently recursive algorithm that repeatedly re-solves the same subproblems.
- ❖ Steps of utilizing dynamic programming:
 1. Find a recursive solution to your problem
 2. Identify the subproblems that are redundantly solved many times.
 3. Optimize the algorithm by eliminating re-solving of subproblems
 - **Storing subproblem results in a table**
- ❖ The final algorithm can be either recursive or iterative.
 - ❖ The iterative form is commonly referred to by the term dynamic programming.

DYNAMIC PROGRAMMING

❖ **Example:** In how many ways we can build a tower with the height n by using blocks with heights of 1, 2, and 3?



DYNAMIC PROGRAMMING

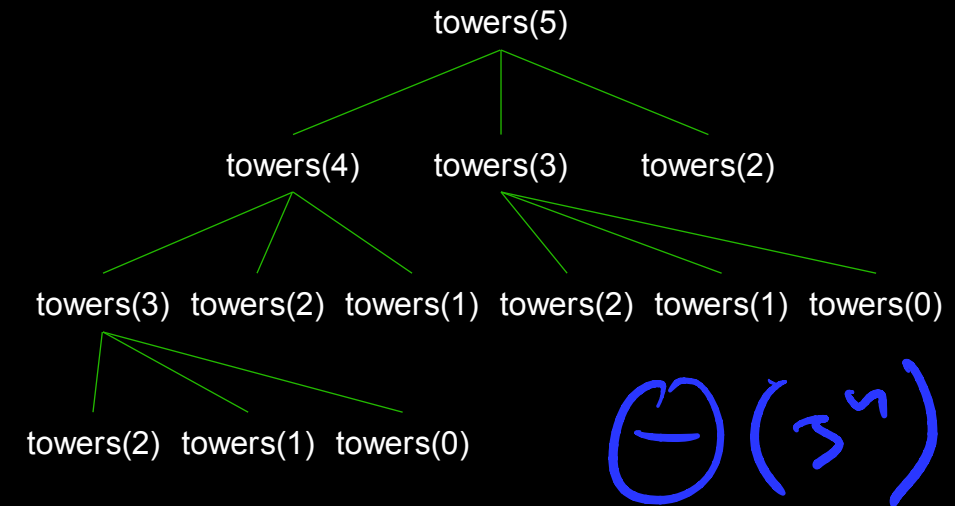
❖ Example: In how many ways we can build a tower with the height n by using blocks with heights of 1, 2, and 3?

- ❖ $\text{towers}(n) = \text{towers}(n-1) + \text{towers}(n-2) + \text{towers}(n-3)$
- ❖ Can be solved easily using recursion.
- ❖ Very slow due to the repetitive solving of the subproblems with small value of n .

❖ Solution using dynamic programming (values from the subproblems stored in a table):

```
towers[0] = 1
towers[1] = 1
towers[2] = 2
for i = 3 to n
```

```
    towers[i] = towers[i-1] + towers[i-2] + towers[i-3]
```



$\Theta(n)$

DYNAMIC PROGRAMMING

❖ Knapsack problem (subset sum problem):

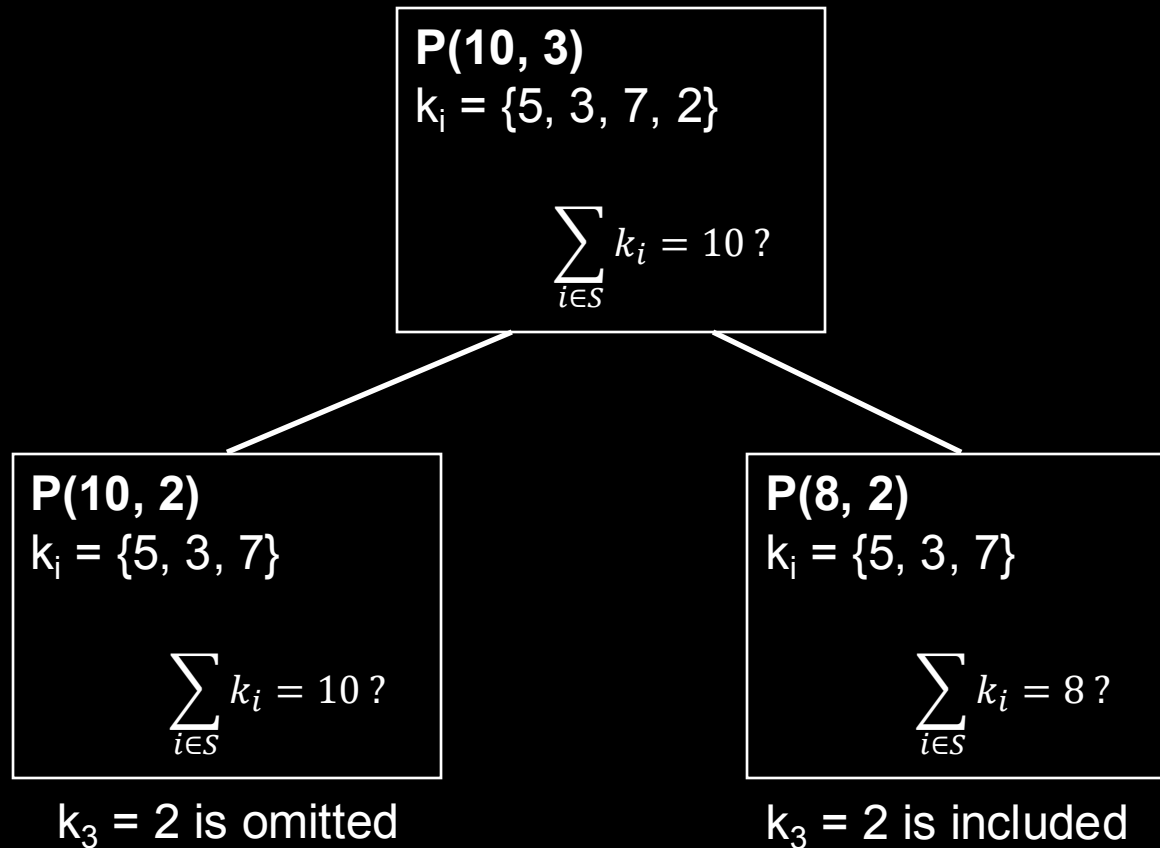
- ❖ find a subset of the n items whose sizes exactly sum to the size of the knapsack, if one exist.

$$\sum_{i \in S} k_i = K$$

S : subset of items

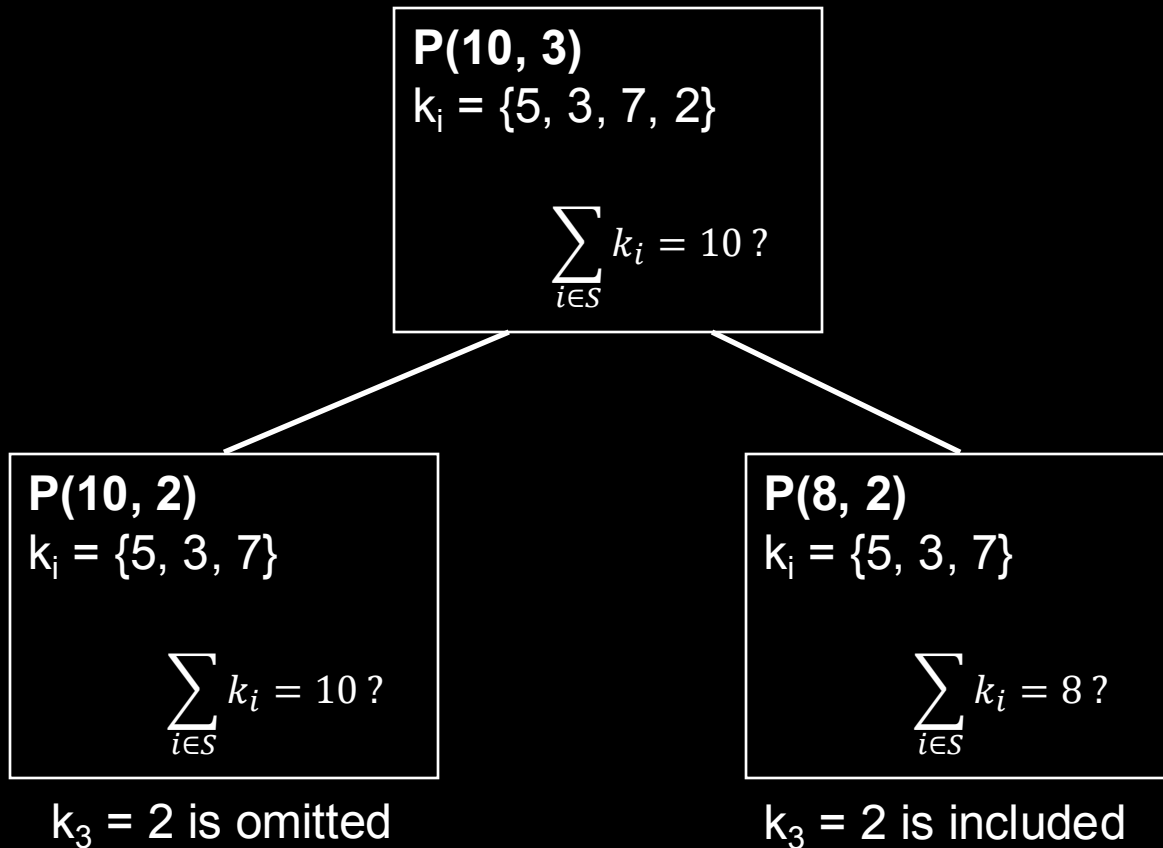
- ❖ E.g If we have 4 items of sizes 3, 8, 7, and 5, and $K = 10$ there exists a solution ($3 + 7 = 10$), but if $K = 14$, there is no solution.
- ❖ Let's denote an instance of the problem as $P(n, K)$
 - ❖ n is the index of the last item (the number of items - 1).
 - ❖ K is the size of the knapsack

DYNAMIC PROGRAMMING



- ❖ The problem $P(n, K)$ can be divided into simpler subproblems:
 - ❖ $P(n-1, K)$ – n th item is omitted
 - ❖ $P(n-1, K-k_n)$ -- n th item is included
- ❖ Can be solved recursively
 - ❖ Base cases are those, where there are only one item, or the knapsack has the size of 0.
- ❖ To avoid solving the same subproblems multiple times, the solutions can be stored in a table.

DYNAMIC PROGRAMMING



	0	1	2	3	4	5	6	7	8	9	10
$k_0=5$	0	-	-	-	-	1	-	-	1	-	-
$k_1=3$	0	-	-	1	-	0	-	-	1	-	-
$k_2=7$	0	-	-	0	-	0	-	1	0	-	1
$k_3=2$	0	-	1	0	-	1	0	1	0	1	0

$$S = \{2, 3, 5\}$$

$$S = \{7, 3\}$$

$P(8, 1)$
 $P(8, 0)$ $P(5, 0)$

PROBABILISTIC ALGORITHMS

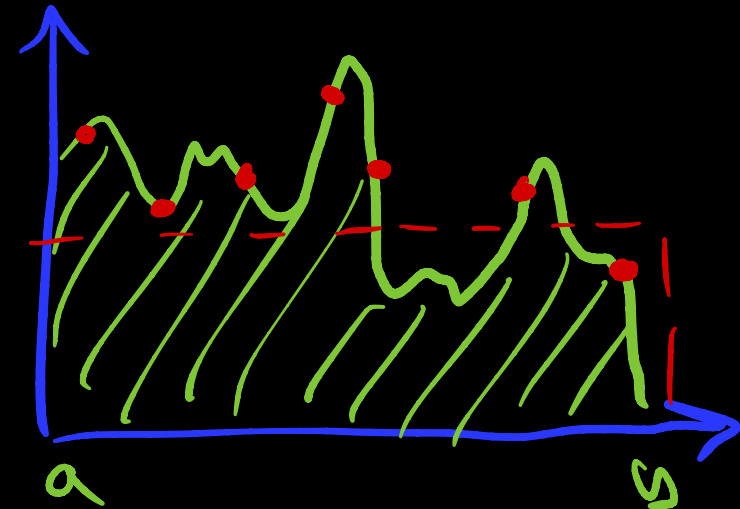
- ❖ Typically, algorithms are defined as set of instructions that are executed *deterministically*.
- ❖ If we relax the definition a bit, we can introduce randomness to our algorithms to:
 - ❖ reduce the execution time,
 - ❖ increase the probability of finding a good solution/result within time limits, and
 - ❖ reduce the probability of a bad case with long running time.
- ❖ Especially useful for very difficult problems for which efficient algorithm is not known.

MONTE CARLO ALGORITHMS

- ❖ Probabilistic algorithms that do not necessary produce exact (or optimal) result
- ❖ But produce some result fast.
- ❖ Accuracy or goodness of the result can be typically improved by increasing the computation time.
- ❖ Example: numerical integration:

```
s=0  
for i = 1 to N  
    x = random([a,b])  
    s = s + f(x)  
return (b - a)(s / N)
```

$$\int_a^b f(x) dx$$



LAS VEGAS ALGORITHMS

- ❖ Produce only correct/optimal results (or informs that result was not found),
- ❖ But the running time is not guaranteed.
- ❖ Typically, the running time is restricted (e.g., maximum number of iterations)
 - ❖ The result may not be found at all.
- ❖ Probability of finding the (correct) result can be increased by increasing the maximum running time

LAS VEGAS ALGORITHMS

❖ Example: prime factorization of large numbers

```
prime_factors = []  
for i = 1 to N  
    x = random_prime_number()  
    if mod(number, x) == 0  
        prime_factors = [prime_factors, x]  
        number = number / x  
    if isprime(number)  
        prime_factors = [prime_factors, number]  
    return prime_factors
```