

# *Analysis and Synthesis of Algorithms*

## *Design of Algorithms*

### Brute-Force Algorithmic Approach

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# Overview

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- What is the Brute-Force Approach
- Why Study it?
- Examples
- Limitations

# What is Brute Force?

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- A straightforward Approach to Problem Solving
- Based on Problem Statement and definitions of the concepts involved
- Examples:
  - Numeric Calculations
  - Search by Enumeration of all possible domain points
- Why the name “Brute Force” ?
  - No cleverness on the implementation
  - Straightforward Implementation (very simple algorithm)
  - Uses computing power not cleverness...

# Why Brute Force?

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- Measure (or yard stick) for Algorithm Performance
  - Space and Time
- Correctness and Optimality
  - Algorithm is simple – it's correctness is trivially established
  - Optimality is usually ensured – all domain is explored...

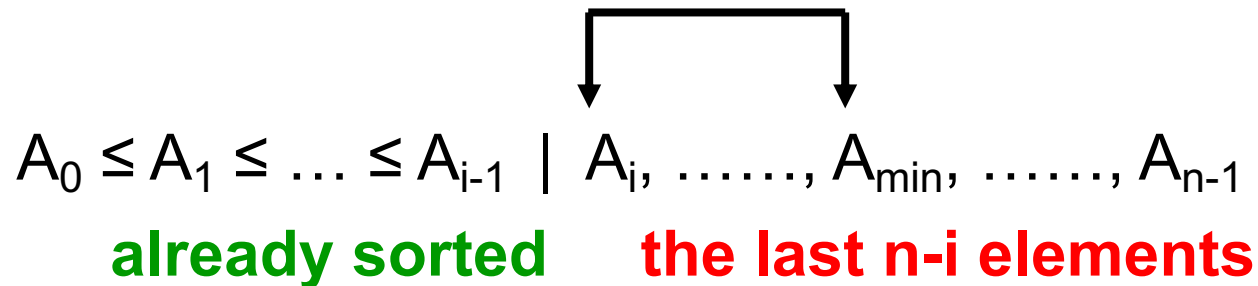
# Brute Force Example

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- We want to compute  $a^n = \underbrace{a \times a \times \dots \times a}_{n \text{ times}}$
- In RSA (Rivest, Shamir, and Adleman) encryption algorithm we need to compute  $a^n \bmod m$  for  $a > 1$  and large  $n$ .
- Basic Approach: Multiply **1** by **a** **n** times
  - So, **n-1** multiplications by **a**...
- Can we do Better?

# Brute Force Example

- Given  $n$  orderable items (*e.g.*, numbers) how can you rearrange them in non-decreasing order?
- **SelectionSort:**
  - Several passes ( $i$  goes from 0 to  $n-2$ )
  - Searches for the smallest item among the last  $(n-i)$  elements and swaps it with  $A_i$



# Brute Force: Selection Sort

**SelectionSort**(A[0,..n-1])

**for** i  $\leftarrow$  0 **to** n-2 **do**

min  $\leftarrow$  i

**for** j  $\leftarrow$  i+1 **to** n-1 **do**

**if** A[j] < A[min]

min  $\leftarrow$  j

swap A[i] and A[min]

**Input size:** n (number of integer values)

**Key op:** “<”, does not depend on type

$$\begin{aligned} C(n) &= \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 \\ &= \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1] \\ &= \frac{(n-1)n}{2} \end{aligned}$$

$$\begin{aligned} C(n) &= \Theta(n^2) \\ \# \text{ of key swaps} &= \Theta(n) \end{aligned}$$

| 89 45 68 90 29 34 **17**

17 | 45 68 90 **29** 34 89

17 29 | 68 90 45 **34** 89

17 29 34 45 | 90 **68** 89

17 29 34 45 68 | 90 **89**

17 29 34 45 68 89 | 90

# Searching

---

- Search for a key,  $K$  in an array  $A[0 \dots n-1]$ ?



# Sequential Search

**SequentialSearch**(A[0..n-1], K)

//Output: index of the first element in A, whose

//value is equal to K or -1 if no such element is found

i = 0

**while** i < n **and** A[i] ≠ K **do**

i = i+1

**if** i < n

**return** i

**else**

**return** -1

Input size: n

Basic op: <, ≠

$$C_{\text{worst}}(n) = 2n+2$$

**Can you improve on it?**

# Searching

---

- Search for a key,  $K$  in an array  $A[0 \dots n-1]$ ?
- Basic Brute-Force is  $\Theta(n)$
- What if the Elements are Sorted (non-decreasing order)?

# Exhaustive Search

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- Traveling Salesman Problem (TSP)
  - Find the shortest tour through a given set of  $n$  cities that visits each city exactly once before returning to the city where it started
  - Can be conveniently modeled by a weighted graph; vertices are cities and edge weights are distances
  - Same as finding “Hamiltonian Circuit”: find a cycle that passes through all vertices exactly once

# Exhaustive Search: TSP (cont.)

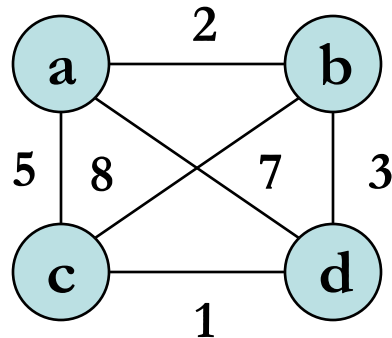
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- Hamiltonian circuit: A sequence of  $n+1$  adjacent vertices  $v_{i_0}, v_{i_1}, \dots, v_{i_{n-1}}, v_{i_0}$

*How can we solve TSP?*

- **Solution:** Derive all possible tours by generating all permutations of  $(n-1)$  intermediate cities, and compute the tour length. Find the shortest among them

# Traveling Salesman (TSP)



Consider only when b precedes c

$\frac{1}{2} (n-1)!$  permutations

$a \rightarrow b \rightarrow c \rightarrow d \rightarrow a$

$$2+8+1+7 = 18$$

$a \rightarrow b \rightarrow d \rightarrow c \rightarrow a$

$$2+3+1+5 = 11 \leftarrow \text{optimal}$$

$a \rightarrow c \rightarrow b \rightarrow d \rightarrow a$

$$5+8+3+7 = 23$$

$a \rightarrow c \rightarrow d \rightarrow b \rightarrow a$

$$5+1+3+2 = 11 \leftarrow \text{optimal}$$

$a \rightarrow d \rightarrow b \rightarrow c \rightarrow a$

$$7+3+8+5 = 23$$

$a \rightarrow d \rightarrow c \rightarrow b \rightarrow a$

$$7+1+8+2 = 18$$

# Exhaustive Search: Knapsack Problem

---

- Given  $n$  items of weights  $w_1, w_2, \dots, w_n$  and values  $v_1, v_2, \dots, v_n$  and a knapsack of capacity  $W$ , find the most valuable subset of the items that fit into the knapsack
  - A transport plane has to deliver the most valuable set of items to a remote location without exceeding its capacity

*How can we solve it ?*

# Knapsack Problem (contd.)

---

- **Solution:** Generate all possible subsets of the  $n$  items, compute total weight of each subset to identify feasible subsets, and find the subset of the largest value

*What is the time efficiency ?*

$$\Omega(2^n)$$

# Knapsack Problem (contd.)

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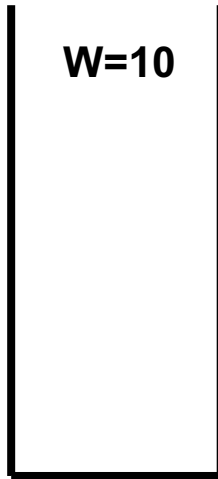
- Alternative Algorithmic Strategy:
  - How about taking items in decreasing order of value/weight?
  - Subject to fitting in the remaining knapsack availability...
  - Sort all items, then pick them linearly while they fit in knapsack.

*What is the time efficiency ?*

**$\Omega(n \log n)$**

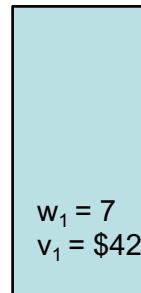


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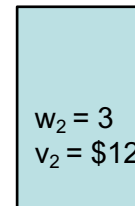


knapsack

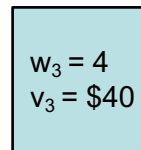
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Item 2: \$4/unit  
Item 3: \$10/unit  
Item 4: \$5/unit



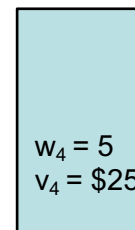
Item 1



Item 2



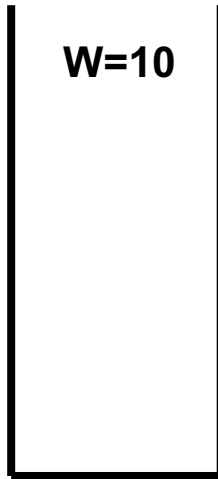
Item 3



Item 4

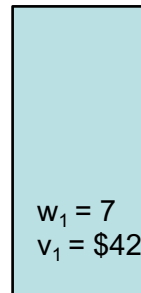
subset	weight	value
$\emptyset$	0	\$0
{1}	7	\$42
{2}	3	\$12
{3}	4	\$40
{4}	5	\$25
{1,2}	10	\$54
{1,3}	11	!feasible
{1,4}	12	!feasible
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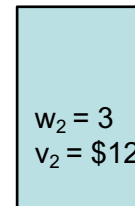


knapsack

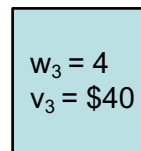
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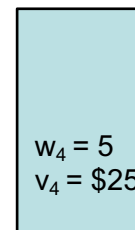
Item 1



Item 2



Item 3

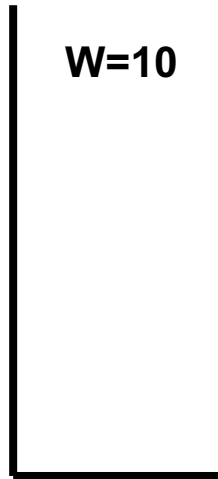


Item 4

Picking Order: {3, 1, 4, 2}      Weight:

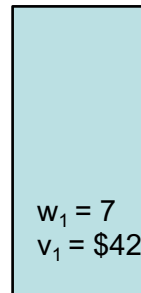
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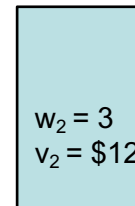


knapsack

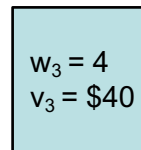
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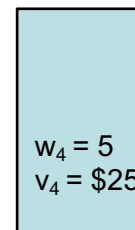
Item 1



Item 2



Item 3

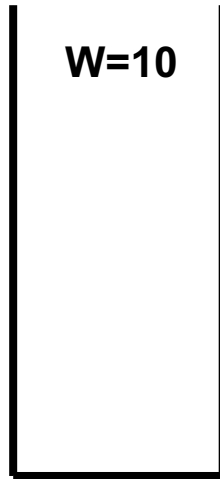


Item 4

Picking Order: {**3**, 1, 4, 2}    Weight: **4**

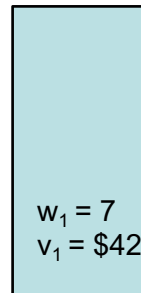
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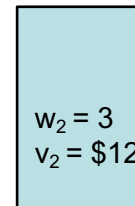


knapsack

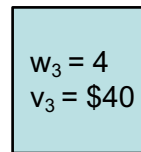
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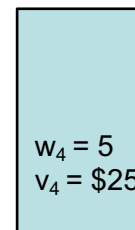
Item 1



Item 2



Item 3

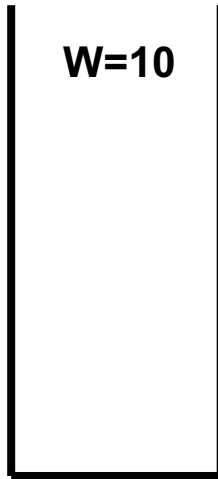


Item 4

Picking Order: {3, 1, 4, 2}    Weight: ~~4+7=11~~

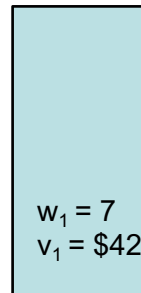
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# Knapsack Problem (contd.)

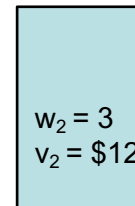


knapsack

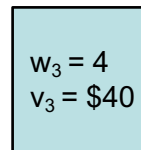
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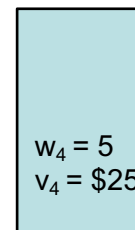
Item 1



Item 2



Item 3

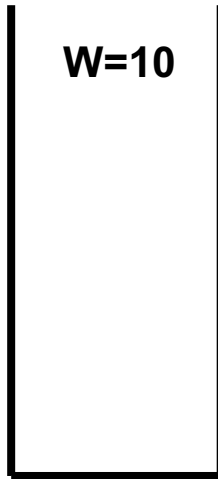


Item 4

Picking Order: {**3**, 1, **4**, 2}    Weight: **4+5=9**

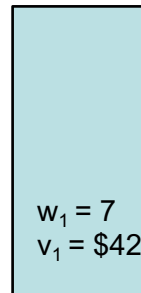
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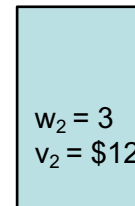


knapsack

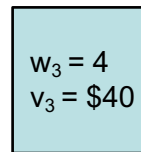
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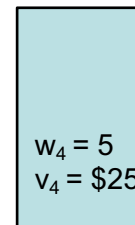
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Item 2



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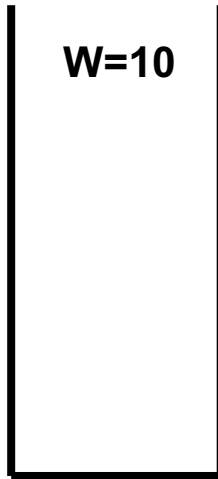


Item 4

Picking Order: {3, 1, 4, 2}    Weight:  $9+3=12$

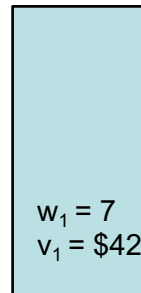
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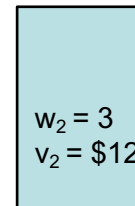


knapsack

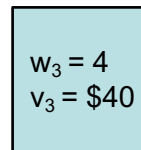
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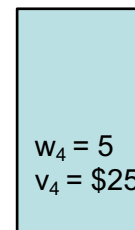
Item 1



Item 2



Item 3

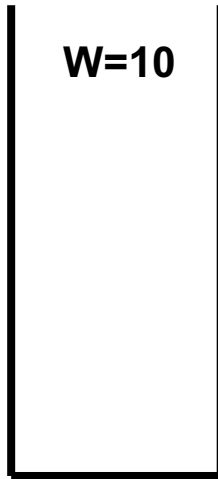


Item 4

Picking Order: {**3**, 1, **4**, 2}      Weight: **9**  
Value: **65**

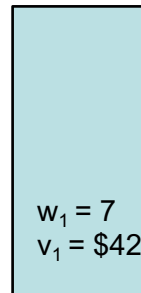
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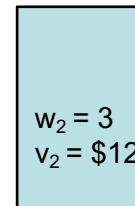


knapsack

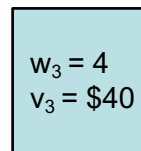
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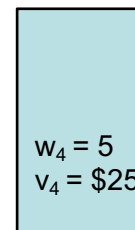
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Item 2



Item 3



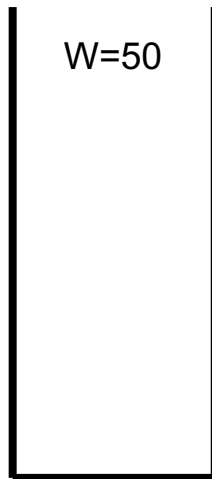
Item 4

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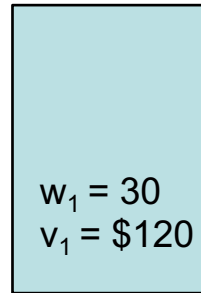
*Works for this example!*



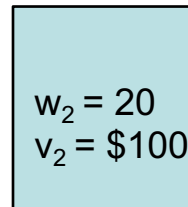
# Knapsack Problem (contd.)



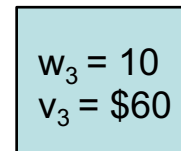
**knapsack**



Item 1



Item 2



Item 3

Item 1: \$4/unit

Item 2: \$5/unit

Item 3: \$6/unit

$$\{\text{Item3, Item2}\} = \$60 + \$100 = \$160$$

$$\{\text{Item3, Item1}\} = \$60 + \$120 = \$180$$

$$\{\text{Item2, Item1}\} = \$100 + \$120 = \$220$$

**Doesn't work for this example**

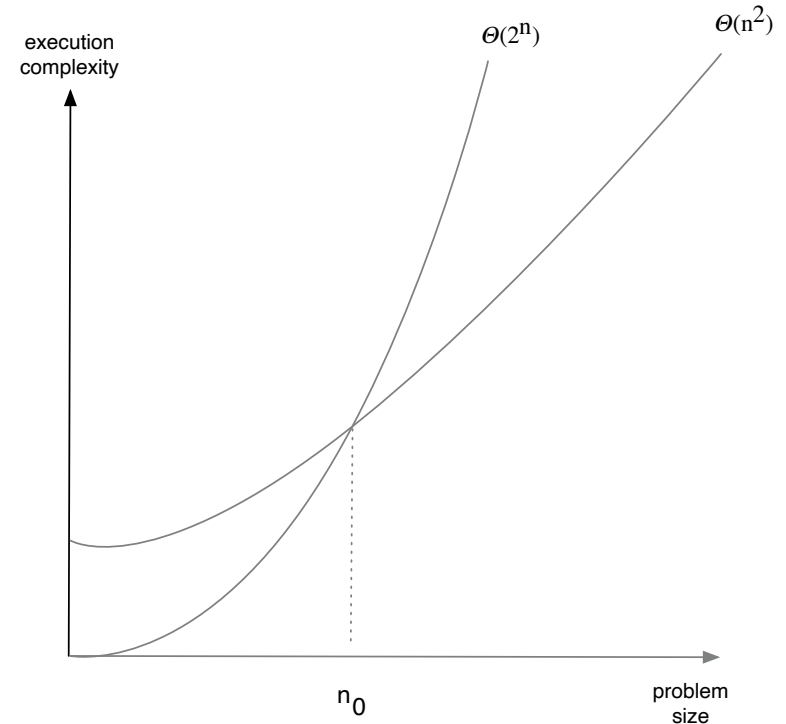
# Exhaustive Search at Large

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- Traveling Salesman and Knapsack problems:
  - Exhaustive search leads to Exponential Time Complexity.
- These are NP-hard problems
  - No known Polynomial-time Algorithms
- Most famous unsolved problem in Computer Science:
  - P vs. NP problem
  - If you solve it, no need to Graduate...
  - More details in later chapters

# Brute-Force: A Note on Complexity

- Likely Exponential Search/Optimization
- Only Feasible for Small Problem Instances
- Still...
  - Simple Approach
  - Provides Correctness



- Think "Hybrid" Algorithms
  - For small problem sizes: Brute-Force
  - For large problem sizes: Polynomial (or better)

# Summary: Exhaustive Search

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- A Brute Force Approach to Combinatorial Problems (which require generation of permutations, or subsets)
- Generate every element of Problem Domain
- Select Feasible ones (ones that satisfy constraints)
- Find the desired one (the one that optimizes some objective function)
- Works, but Only Feasible for Small Problem Sizes.