

# *Multiple Choice KP (MCKP) is NP-Hard*

**MCKP**: in addition to the input data for KP01:

the set of the  $n$  items is *partitioned* into  $k$  disjoint subsets  $N_1, N_2, \dots, N_k$ .

- determine a subset of the  $n$  items, **with at most one item for each subset**  $N_h$  ( $h = 1, \dots, k$ ), so as to maximize the global profit, and such that the global weight is not larger than the knapsack capacity  $C$ .
- **Input**:  $m, C, k, (P_j), (W_j)$  ( $j = 1, \dots, n$ ),  $N_h$  ( $h = 1, \dots, k$ )
- **Size**:  $3 + 2n + k * n$  (matrix  $A_{hj}$ ), with  $k \leq n : n * n$
- **Size**:  $3 + 2n + n$  (partition of the set  $\{1, 2, \dots, n\}$ ) :  $n$ .
- *Binary Decision Tree*: similar to the decision tree of *KP-01*):  $n$  levels, 2 descendent nodes and constant time for each node:
- **MCKP**  $\in$  **Class NP** ;
- **MCKP** is a “generalization” of *KP-01* : *KP-01*  $\propto$  **MCKP**

# *BLP Model for MCKP*

\* *Binary Matrix*  $A_{hj}$  ( $h = 1, \dots, k; j = 1, \dots, n$ ), with:

- $A_{hj} = 1$  if  $j \in N_h$  ;  $A_{hj} = 0$  otherwise.

$$\max \sum_{j=1,n} P_j x_j$$

$$\sum_{j=1,n} W_j x_j \leq C$$

$$\sum_{j=1,n} A_{hj} x_j \leq 1 \quad (h = 1, \dots, k)$$

$$x_j \in \{0, 1\} \quad (j = 1, \dots, n)$$

The *BLP Model* has a number of binary variables  $x_j$  polynomial in the size of *MCKP*:

# *Multiple Knapsack Problem (MKP01)* is NP-Hard

*MKP01: given:  $n$  items,  $m$  containers (knapsacks),  
 $P_j$  profit of item  $j$ ,  $W_j$  weight of item  $j$  ( $j = 1, \dots, n$ ),  
 $C_i$  capacity of container  $i$  ( $i = 1, \dots, m$ ):*

*insert a subset of the  $n$  items in each of the  $m$  containers  
in order to maximize the global profit of the inserted  
items, and in such a way that the global weight of the  
items inserted in each container  $i$  ( $i = 1, \dots, m$ ) is not  
greater than the corresponding capacity  $C_i$*

**Input:**  $n, m, (P_j), (W_j)$  ( $j = 1, \dots, n$ ),  $(C_i)$  ( $i = 1, \dots, m$ )

# *MKP01* is NP-Hard

*MKP01*: given:  $n$  items,  $m$  containers (knapsacks),

$P_j$  profit of item  $j$ ,  $W_j$  weight of item  $j$  ( $j = 1, \dots, n$ ),

$C_i$  capacity of container  $i$  ( $i = 1, \dots, m$ ):

insert a subset of the  $n$  items in each of the  $m$  containers in order to maximize the global profit of the inserted items, and in such a way that the global weight of the items inserted in each container  $i$  ( $i = 1, \dots, m$ ) is not greater than the corresponding capacity  $C_i$

- **Size:**  $2 + 2n + m : n + m$ , ( $m \leq n$  : Size  $n$ )
- **Decision Tree:**  $n$  levels (one for each item  $j$ );  
( $m + 1$ ) descendent nodes (insert item  $j$  in knapsack 1, or 2, ..., or  $m$ , or **in no knapsack**) and constant time for each node:

***MKP01***  $\in$  **Class NP** ;

(**BLP** model with ( $m * n$ ) binary variables  $x_{ij}$ )

- ***MKP01*** is a “generalization” of ***KP-01*** :  
***KP-01***  $\propto$  ***MCKP***

# *Bin Packing Problem (BPP) is NP-Hard*

**Given:**  $n$  items;  $m$  bins (each with capacity  $C$ );

$W_j$  weight of item  $j$  ( $j = 1, \dots, n$ ):

insert all the  $n$  items *in the bins in order to minimize the number of used bins*, and in such a way that *the global weight of the items inserted in a bin is not greater than the capacity  $C$ .*

- **Input:**  $n, m, C, (W_j)$  ( $j = 1, \dots, n$ ); **Size:**  $3 + n : n$
- $m \leq n$

# *Feasibility Problem of BPP (F-BPP)*

Given:  $n$  items;  $m$  bins (each with capacity  $C$ );

$W_j$  weight of item  $j$  ( $j = 1, \dots, n$ ):

insert all the  $n$  items in the  $m$  bins in such a way that the global weight of the items inserted in a bin is not greater than the capacity  $C$ .

## *F-BPP is NP-Hard*

- **Input:**  $n, m, C, (W_j)$  ( $j = 1, \dots, n$ ); **Size:**  $3 + n : n$
- **Decision Tree:**  $n$  levels (one for each item  $j$ );
- \*  $m$  descendent nodes (insert item  $j$  in bin 1, or 2, ..., or  $m$ ) and constant time for each node ( $m \leq n$ ):

**F-BPP**  $\in$  **Class NP** ;

Also **BPP**  $\in$  **Class NP** (same Size and Decision Tree as F-BPP);  
(BLP model with  $(m * n + m)$  binary variables  $x_{ij}, y_i$ )

# *F-BBP is NP-Hard*

Given:  $n$  items;  $m$  bins (each with capacity  $C$ );

$W_j$  weight of item  $j$  ( $j = 1, \dots, n$ ):

insert all the  $n$  items in the  $m$  bins in such a way that the global weight of the items inserted in a bin is not greater than the capacity  $C$ .

- **$PP \propto F-BPP$  :**
- **Given any instance of  $PP$ :  $t, (a_j), b$  (Size:  $t$ )**
  - 1) **Define (in time  $O(t)$ ) an instance  $(n, (W_j), m, C)$  of  $F-BPP$ :**
    - \*  $n := t$
    - \*  $C := b$
    - \*  $m := 2$
    - \*  $W_j := a_j$  ( $j = 1, \dots, n$ ).
  - 2) **Determine (if it exists) a feasible solution  $(x)$  of  $F-BPP$ .**
  - 3) **If a feasible solution  $(x_{1j}, x_{2j})$  of  $F-BPP$  exists, then  $PP$  has a feasible solution  $(x_{1j}, x_{2j})$** 

Otherwise:  $PP$  has no feasible solution.

Computing time  $O(n)$  (hence  $O(t)$ , polynomial in the size of  $PP$ )

# *Generalized Assignment Problem* *(GAP) is NP-Hard*

**Given:**  $m$  machines and  $n$  jobs:

$c_{ij}$  cost ( $r_{ij}$  amount of resource utilized) for assigning job  $j$  to machine  $i$  ( $i = 1, \dots, m; j = 1, \dots, n$ );

$b_i$  amount of resource available for machine  $i$  ( $i = 1, \dots, m$ ):

Assign each job to a machine so as to minimize the global cost, and in such a way that the global resource utilized by each machine  $i$  is not greater than the corresponding available resource  $b_i$ .

**Input:**  $m, n, (c_{ij}), (r_{ij})$  ( $i = 1, \dots, m; j = 1, \dots, n$ );

$(b_i)$  ( $i = 1, \dots, m$ )

**Size:**  $2 + 2m * n + m : m * n$



# *Feasibility Problem of GAP (F-GAP)*

**Given:**  $m$  machines and  $n$  jobs:

$r_{ij}$  amount of resource utilized for assigning job  $j$  to machine  $i$  ( $i = 1, \dots, m; j = 1, \dots, n$ );

$b_i$  amount of resource available for machine  $i$  ( $i = 1, \dots, m$ ):

Assign each job to a machine in such a way that the global resource utilized by each machine  $i$  is not greater than the corresponding available resource  $b_i$ .

**Input:**  $m, n, (r_{ij})$  ( $i = 1, \dots, m; j = 1, \dots, n$ );  $(b_i)$  ( $i = 1, \dots, m$ ):

**Size:**  $m * n$

- **Decision Tree:**  $n$  levels (one for each job  $j$ );
- \*  $m$  descendent nodes (insert job  $j$  in machine 1, or 2, ..., or  $m$ ) and constant time for each node:

**F-GAP**  $\in$  **Class NP** ;

**Also GAP**  $\in$  **Class NP** (same Size and Decision Tree as F-GAP);  
(BLP model with  $(m * n)$  binary variables  $x_{ij}$  )

# Feasibility Problem of GAP (*F-GAP*)

Given:  $m$  machines and  $n$  jobs:

$r_{ij}$  amount of resource utilized for assigning job  $j$  to machine  $i$  ( $i = 1, \dots, m; j = 1, \dots, n$ );

$b_i$  amount of resource available for machine  $i$  ( $i = 1, \dots, m$ ):

Assign each job to a machine in such a way that the global resource utilized by each machine  $i$  is not greater than the corresponding available resource  $b_i$ .

*PP*  $\propto$  *F-GAP* :

• Given any instance of *PP*:  $t, (a_j), b$  (Size:  $t$ )

1) Define (in time  $O(t)$ ) an instance  $(m, n, (r_{ij}), (b_i))$  of *F-GAP*:

\*  $n := t$

\*  $m := 2; b_1 := b; b_2 := \sum_{j=1,t} a_j - b$

\*  $r_{1j} := a_j; r_{2j} := a_j$  ( $j = 1, \dots, n$ ).

2) Determine (if it exists) a feasible solution  $(x_{1j}, x_{2j})$  of *F-GAP*.

3) If a feasible solution of *F-GAP* exists, then *PP* has a feasible solution  $(x_{1j}, x_{2j})$

Otherwise: *PP* has no feasible solution.

Computing time  $O(n)$  (hence  $O(t)$ , polynomial in the size of *PP*)

# *F-BPP is a particular case of F-GAP*

*F-GAP: given:  $m$  machines and  $n$  jobs:*

*$r_{ij}$  amount of resource utilized for assigning job  $j$  to machine  $i$  ( $i = 1, \dots, m; j = 1, \dots, n$ );*

*$b_i$  amount of resource available for machine  $i$  ( $i = 1, \dots, m$ ):*

*assign each job to a machine so that the global resource utilized by each machine  $i$  is not greater than the available resource  $b_i$ .*

*F-BPP: given:  $n$  items;  $m$  bins (each with capacity  $C$ );*

*$W_j$  weight of item  $j$  ( $j = 1, \dots, n$ ):*

*insert all the  $n$  items in the  $m$  bins so that the global weight of the items inserted in a bin is not greater than the capacity  $C$ .*

**Arising when:**

*$r_{ij} := W_j$  ( $i = 1, \dots, m; j = 1, \dots, n$ );*

*$b_i := b$  ( $i = 1, \dots, m$ )*

# Knapsack Problem with Minimization Objective Function (*KP01-Min*)

**Given:**

$n$  items,

$P_j$  “profit” of item  $j$ ,  $j = 1, \dots, n$  ( $P_j > 0$ ),

$W_j$  “weight” of item  $j$ ,  $j = 1, \dots, n$  ( $W_j > 0$ ),

*one* container (“knapsack”) with “threshold”  $B$ :

determine a subset of the  $n$  items so as to **minimize the global profit**, and such that the **global weight is not smaller** than the knapsack threshold  $B$ .

***KP01-Min* is NP-Hard**

# Mathematical Model of *KP01-Min*

$$y_j = \begin{cases} 1 & \text{if item } j \text{ is inserted in the knapsack} \\ 0 & \text{otherwise} \end{cases} \quad (j = 1, \dots, n)$$

$$\min \quad \sum_{j=1,n} P_j y_j$$

$$\sum_{j=1,n} W_j y_j \geq B$$

$$y_j \in \{0, 1\} \quad (j = 1, \dots, n)$$

**BLP Model (Binary Linear Programming Model)**

***KP01-Min* is “equivalent” to *KP01*.**

***KP01-Min* is “equivalent” to *KP01*.**

**Set  $y_j = 1 - x_j$  ( $j = 1, \dots, n$ ) and replace  $y_j$  with  $1 - x_j$**

$$1) \quad \min T = \sum_{j=1, n} P_j y_j = \sum_{j=1, n} P_j (1 - x_j) =$$

$$P - \max \sum_{j=1, n} P_j x_j$$

$$\text{where } P = \sum_{j=1, n} P_j$$

***KP01-Min* is “equivalent” to *KP01* (2).**

$$2) \quad \sum_{j=1, n} W_j y_j = \sum_{j=1, n} W_j (1 - x_j) =$$

$$\sum_{j=1, n} W_j - \sum_{j=1, n} W_j x_j \geq B$$

$$\sum_{j=1, n} W_j x_j \leq C'$$

where  $C' = \sum_{j=1, n} W_j - B$

***KP01-Min*** is “equivalent” to ***KP01*** (3).

$$\text{Min } T = P - \max \sum_{j=1, n} P_j x_j$$

$$\sum_{j=1, n} W_j x_j \leq C'$$

$$x_j \in \{0, 1\} \quad (j = 1, \dots, n)$$

where:  $P = \sum_{j=1, n} P_j$ ;  $C' = \sum_{j=1, n} W_j - B$

- *Problem KP01*
- ***KP01-Min is NP-Hard***



# Variant of *KP01*: *Equality-KP01 (E-KP01)*

- Same input data as for the *KP01*:  $n$ ,  $C$ ,  $(P_j)$ ,  $(W_j)$
- \* Determine a subset of the  $n$  items so that the global weight is equal to the knapsack capacity  $C$ .

$$\begin{aligned} \max \quad & \sum_{j=1,n} W_j x_j \\ & \sum_{j=1,n} W_j x_j = C \\ & x_j \in \{0, 1\} \quad (j = 1, \dots, n) \quad (\text{BLP Model}) \end{aligned}$$

**The Feasibility Problem of *E-KP01* is NP-Hard**

***E-KP01* is NP-Hard**

# *Feasibility Problem of E-KP01*

## *(F-E-KP01)*

- Same input data as for the *KP01*:  $n, C, (W_j)$
- Determine a subset of the  $n$  items so that the global weight is equal to the knapsack capacity  $C$ .
- *F-E-KP01* is NP-Hard
- Input:  $n, C, (W_j)$  :  
Size:  $2 + n : n$
- Binary Decision Tree of *KP-01*:  
*F-E-KP01*  $\in$  Class NP

# *F-E-KP01* is NP-Hard

- Same input data as for the *KP01*:  $n, C, (W_j)$
  - Determine a subset of the  $n$  items so as to that the global weight is equal to the knapsack capacity  $C$ .
  - $PP \propto F-E-KP01$  :
  - Given any instance of *PP*:  $t, (a_j), b$  (Size:  $t$ )
- 1) Define (in time  $O(t)$ ) an instance  $(n, C, (W_j))$  of *F-E-KP01*:
    - \*  $n := t$
    - \*  $C := b$
    - \*  $W_j := a_j$  ( $j = 1, \dots, n$ ).
  - 2) Determine (if it exists) a feasible solution  $(x_j)$  of *F-E-KP-01*.
  - 3) If a feasible solution  $(x_j)$  of *F-E-KP-01* exists, then *PP* has a feasible solution  $(x_j)$   
Otherwise: *PP* has no feasible solution.
- Computing time  $O(n)$  (hence  $O(t)$ , polynomial in the size of *PP*)

# Variant of *KP01*: *Subset Sum Problem (SSP)*

- Item  $j$  has weight  $W_j$  and profit  $P_j = W_j$  ( $j = 1, \dots, n$ ):  
Determine a subset of the  $n$  items so that the global weight is maximum and not greater than  $C$ .
- Cut of metal planks with minimization of the waste.

$$\begin{aligned} \max \quad & \sum_{j=1,n} W_j x_j \\ & \sum_{j=1,n} W_j x_j \leq C \\ & x_j \in \{0, 1\} \quad (j = 1, \dots, n) \quad \text{(BLP Model)} \end{aligned}$$

***SSP* is NP-Hard**

# SSP is NP-Hard

- **SSP**: input data:  $n, C, (W_j)$ .
- Determine a **subset of the  $n$  items** so that the **global weight** is maximum and not greater than  $C$ .
- **Size**:  $2 + n : n$
- **Binary Decision Tree of KP-01**:  $SSP \in \text{Class NP}$

\*  $PP \propto SSP$

**Given any instance of  $PP$ :  $t, (a_j), b$  (Size:  $t$ )**

1) **Define (in time  $O(t)$ ) an instance  $(n, (W_j), C)$  of  $SSP$ :**

\*  $n := t ; C := b ; W_j := a_j ( j = 1, \dots, n)$ .

2) **Determine the optimal solution  $(x_1, x_2, \dots, x_n, z)$  of  $SSP$ .**

3) **If  $z = C$  :  $PP$  has a feasible solution  $(x_1, x_2, \dots, x_n)$**

**If  $z < C$  :  $PP$  has no feasible solution**

**Computing time  $O(n)$  (hence  $O(t)$ , polynomial in the size of  $PP$ )**

# Subset Sum Problem (SSP)

- Item  $j$  has weight  $W_j$  and profit  $P_j = W_j$  ( $j = 1, \dots, n$ ):
- Determine a subset of the  $n$  items so that the global weight is maximum and not greater than  $C$ .

$$\begin{aligned} \max \quad & \sum_{j=1,n} W_j x_j \\ & \sum_{j=1,n} W_j x_j \leq C \\ & x_j \in \{0, 1\} \quad (j = 1, \dots, n) \quad \text{(BLP Model)} \end{aligned}$$

**SSP is NP-Hard.**

**SSP is a special case of KP01**

**The feasibility problem of SSP is polynomial**

# Variant of *KP01*:

## *Change Making Problem (CMP)*

- Given  $n$  banknotes and a cheque (check),
- \*  $W_j$  is the value of banknote  $j$  ( $j = 1, \dots, n$ ), with  $W_j > 0$ ,
- $C$  is the value of the cheque:
- select a minimum cardinality subset of banknotes so that the global value is equal to  $C$ .

$$\begin{aligned} \min \quad & \sum_{j=1,n} x_j \\ & \sum_{j=1,n} W_j x_j = C \\ & x_j \in \{0, 1\} \quad (j = 1, \dots, n) \quad \text{(BLP Model)} \end{aligned}$$

*CMP* is NP-Hard (its Feasibility Problem is NP-Hard)

# *Feasibility Problem of CMP (F-CMP)*

- Input:  $n$ ,  $C$ ,  $(W_j)$
- select a subset of **banknotes** so that the global value **is equal** to  $C$ .
- *F-CMP is NP-Hard:*
- “Banknotes” correspond to “items”;
- “Cheque value” corresponds to “capacity  $C$ ”;
- *F-CMP is identical to F-E-KP01.*
- *CMP is NP-Hard*



# Variant of *KP01*: *Two-Constrained KP (2C-KP)*

**Given:**

$n$  items,

$P_j$  “profit” of item  $j$ ,  $j = 1, \dots, n$  ( $P_j > 0$ ),

$W_j$  “weight” of item  $j$ ,  $j = 1, \dots, n$  ( $W_j > 0$ ),

$V_j$  “volume” of item  $j$ ,  $j = 1, \dots, n$  ( $V_j > 0$ ),

*one* container (“knapsack”) with:

\* “weight capacity”  $C$ , and “volume capacity”  $D$ :

determine a subset of the  $n$  items so as to **maximize** the **global profit**, and such that the **global weight** is not greater than the weight capacity  $C$  *and* the **global volume** is not greater than the volume capacity  $D$ .

***2C-KP* is NP-Hard**

# *Mathematical Model for 2C-KP*

$$\mathbf{max} \quad \sum_{j=1,n} P_j x_j$$

$$\sum_{j=1,n} W_j x_j \leq C$$

$$\sum_{j=1,n} V_j x_j \leq D$$

$$x_j \in \{0, 1\} \quad (j = 1, \dots, n) \quad \text{(BLP Model)}$$

# 2C-KP is NP-Hard

Input:  $n$ ,  $C$ ,  $D$ ,  $(P_j)$ ,  $(W_j)$ ,  $(V_j)$   $j = 1, \dots, n$  ;

determine a subset of the  $n$  items so as to **maximize the global profit**, and such that the **global weight is not greater** than the weight capacity  $C$  *and* the **global volume is not greater** than the volume capacity  $D$ .

- **Size:**  $3 + 3n : n$
- *Binary Decision Tree of KP-01:*

$2C-KP \in \text{Class NP}$

\* *2C-KP is a generalization of KP-01:*

$(KP-01 \propto 2C-KP)$

**The feasibility problem of 2C-KP is polynomial.**

# Variant of *KP01*: *Bounded-KP (BKP)*

In addition to the input data for *KP01*:

- \*  $d_j$  = number of **available** items of **item-type  $j$**  ( $j = 1, \dots, n$ )
- $x_j$  = number of items **selected** for **item-type  $j$**  ( $j = 1, \dots, n$ )

$$\max \sum_{j=1,n} P_j x_j$$

$$\sum_{j=1,n} W_j x_j \leq C$$

$$0 \leq x_j \leq d_j \quad \text{INTEGER} \quad (j = 1, \dots,$$

$n)$

**ILP Model;**

***BKP* is NP-Hard**

# *BKP* is NP-Hard

Input:  $n$ ,  $C$ ,  $(P_j)$ ,  $(W_j)$ ,  $(d_j)$   $j = 1, \dots, n$  ;

determine for each item-type  $j$  ( $j = 1, \dots, n$ ) the number of items to be inserted in the knapsack so as to maximize the global profit, and such that the global weight is not greater than the capacity  $C$ .

- **Size:**  $2 + 3n$  :  $n$  (number of “symbols” required to represent the input data)
- **Decision Tree:**  $n$  levels, one for each item-type  $j$  ( $d_j + 1$ ) descendent nodes (one node for each possible number of inserted items of item-type  $j$ ) and constant time for each node:

$BKP \in \text{Class } NP$  (???)

# *BKP* is NP-Hard

Input:  $n$ ,  $C$ ,  $(P_j)$ ,  $(W_j)$ ,  $(d_j)$   $j = 1, \dots, n$  ;

determine for each item-type  $j$  ( $j = 1, \dots, n$ ) the number of items to be inserted in the knapsack so as to **maximize the global profit**, and such that the **global weight is not greater** than the weight capacity  $C$ .

- **Size:**  $2 + 3n$  :  $n$  (number of “symbols” required to represent the input data)
- *Decision Tree:*  $n$  levels, one for each item-type  $j$   
 $(d_j + 1)$  *descendent nodes* (one node for each possible number of inserted items of item-type  $j$ ) *and constant time for each node:*

$BKP \in \text{Class NP}$  (???)

\* is  $d_j$  a polynomial function of the size  $n$ ?

# BKP is NP-Hard

Input:  $n$ ,  $C$ ,  $(P_j)$ ,  $(W_j)$ ,  $(d_j)$   $j = 1, \dots, n$  ;

determine for each item-type  $j$  ( $j = 1, \dots, n$ ) the number of items to be inserted in the knapsack so as to maximize the global profit, and such that the global weight is not greater than the weight capacity  $C$ .

- **Size:**  $2 + 3n$  :  $n$  (number of “symbols” required to represent the input data)
- **Decision Tree:**  $n$  levels, one for each item-type  $j$   
( $d_j + 1$ ) descendent nodes (one node for each possible number of inserted items of item-type  $j$ ) and constant time for each node:

$BKP \in \text{Class NP}$  (???)

\* is  $d_j$  a polynomial function of the size  $n$ ?

\*  $B = \max \{d_j : j = 1, \dots, n\} \leq (2)^k$

where  $k$  is the number of bits needed to represent  $B$

\* **Size:**  $2 + 2n + k*n$  :  $k*n$  (number of “bits”)

\*  $d_j$  is defined by an exponential function of the size  $k*n$

# Transformation of an ILP model with $n$ variables into a BLP model with $n*k$ variables

- \*  $x_j$  integer variable with  $x_j \geq 0$ ,  $x_j \leq d_j$  (with  $d_j \leq B$ ).
- \* for each variable  $x_j$  ( $j = 1, \dots, n$ ) introduce  $k$  binary variables  $t_{jh}$ , with  $h = 1, \dots, k$

$$x_j = \sum_{h=1,k} 2^h t_{jh}$$

$$t_{jh} \in \{0, 1\} \quad h = 1, \dots, k \quad (j = 1, \dots, n)$$

$$k = \lceil \log_2(B + 1) \rceil \quad \text{with } z = \log_2(B + 1)$$

**BLP model with  $n*k$  variables:**

**Binary Decision Tree with  $n*k$  levels (polynomial function of the size  $k*n$ ).**



# BKP is NP-Hard

Input:  $n$ ,  $C$ ,  $(P_j)$ ,  $(W_j)$ ,  $(d_j)$   $j = 1, \dots, n$  ;

determine for each item-type  $j$  ( $j = 1, \dots, n$ ) the number of items to be inserted in the knapsack so as to maximize the global profit, and such that the global weight is not greater than the weight capacity  $C$ .

- **Size:**  $2 + 2n + k*n : k*n$
- for each variable  $x_j$  ( $j = 1, \dots, n$ ) introduce  $k$  binary variables  $t_{jh}$ , with  $h = 1, \dots, k$
- *Binary Decision Tree:  $k*n$  levels (one for each binary variable  $t_{jh}$ );*  
*2 descendent nodes and constant time for each node:*  
*BKP  $\in$  Class NP*
- \* *BKP is a generalization of KP-01 (KP-01  $\propto$  BKP)*

**The feasibility problem of BKP is polynomial.**

# Variant of *KP01*: *Unbounded-KP (UKP)*

No limit on the number of items available for each item-type  $j$  ( $j = 1, \dots, n$ ).

- $x_j$  = number of items selected for item-type  $j$  ( $j = 1, \dots, n$ )

$$\max \sum_{j=1,n} P_j x_j$$

$$\sum_{j=1,n} W_j x_j \leq C$$

$$x_j \geq 0 \quad \text{INTEGER} \quad (j = 1, \dots, n)$$

**ILP Model**

**It is known that *UKP* is NP-Hard**

# Variant of *KP01*: *Unbounded-KP (UKP)*

No limit on the number of items available for each item-type ( $d_j = \infty, j = 1, \dots, n$ )

- $x_j$  = number of items **selected** for **item-type  $j$**  ( $j = 1, \dots, n$ )

$$\max \sum_{j=1,n} P_j x_j$$

$$\sum_{j=1,n} W_j x_j \leq C$$

$$x_j \geq 0 \quad \text{INTEGER} \quad (j = 1, \dots, n)$$

**UKP** is a special case of **BKP**:  $d_j = \text{int}(C / W_j), j = 1, \dots, n$

# Set Covering Problem (SCP)

- Given: a “Binary Matrix”  $A$  with  $m$  rows e  $n$  columns;

$C_j$  “cost” of column  $j$  ( $j = 1, \dots, n$ ) ( $C_j > 0$ )

If  $A_{ij} = 1$  ( $i = 1, \dots, m; j = 1, \dots, n$ ):

column  $j$  “covers” row  $i$

row  $i$  “is covered” by column  $j$

Select a subset of the  $n$  columns of  $A_{ij}$  so that:

- the **sum of the costs** of the selected columns is **minimum**,
- all the  $m$  rows are **covered** at least once by the selected columns

# Example of SCP

$n=8, m=5;$

binary matrix:

		j						
		1	2	3	4	5	6	7
i	1	0	1	0	1	1	0	0
	2	1	0	0	0	0	0	0
	3	0	0	1	1	0	1	0
	4	0	0	0	0	0	0	0
costs		5	5	3	6	2	4	5
feasible solution:		0	1	1	0	1	0	1
Cost =		5	5	3				
		0	0	1	0	1	0	1
		0						

$Cost = 5 + 5 + 3 = 13$

# Example of *SCP*

		j						
		1	2	3	4	5	6	7
i	1	0	1	0	1	1	0	0
	2	1	0	0	0	0	0	0
	3	0	0	1	1	0	1	0
	4	1	1	0	0	1	0	0
costs		5	5	30	6	20	4	51

optimal solution:  $\text{Cost} = 5 + 3 + 2 = 10$

# Mathematical Model of *SCP*

$$x_j = \begin{cases} 1 & \text{if column } j \text{ is selected} \\ 0 & \text{otherwise} \end{cases} \quad (j = 1, \dots, n)$$

$$\min \sum_{j=1,n} C_j x_j$$

$$\sum_{j=1,n} A_{ij} x_j \geq 1 \quad (i = 1, \dots, m)$$

$$x_j \in \{0, 1\} \quad (j = 1, \dots, n)$$

## BLP Model

The Feasibility Problem of *SCP* is polynomial.

*SCP*  $\in$  Class *NP* (Binary Decision Tree with  $n$  levels)

*SCP* is known to be NP-Hard

# Variant: *Set Partitioning (SPP)*

Select a subset of the  $n$  columns of matrix  $A_{ij}$  so that:

- the **sum of the costs** of the selected columns is **minimum**,
- all the  $m$  rows are **covered exactly once** by the selected columns.

$$\min \sum_{j=1,n} C_j x_j$$

$$\sum_{j=1,n} A_{ij} x_j = 1 \quad (i = 1, \dots, m)$$

$$x_j \in \{0, 1\} \quad (j = 1, \dots, n) \quad \text{BLP model}$$

**The Feasibility Problem of *SPP* is known to be NP-Hard**  
***SPP* is NP-Hard**





# Example of *SPP*

optimal solution of *SCP*, cost = 10

$\text{cost}(SPP) \square \text{cost}(SCP)$

Number of covering columns

		j							
		1	2	3	4	5	6	7	
i	1	0	1	0	1	1	0	0	1
	2	1	0	0	0	0	0	0	1
	3	0	0	1	1	0	1	0	1
	costs	5	5	3	6	2	4	5	0

optimal solution of *SPP*: Cost = 5 + 9 = 14

5 0 0 1 0 1 0 1 0

# Strong Formulation of the *BPP*

Let  $S$  = subset of the  $n$  items corresponding to a *feasible loading* of a bin:

$S$  contained in  $\{1, 2, \dots, n\}$  and such that  $\sum_{j \in S} W_j \leq C$

$P$  = family of all the feasible subsets  $S = \{S_1, S_2, \dots, S_k\}$   
(  $k$  can grow exponentially with  $n$  ).

A subset  $S_h$  is *maximal* if the addition of an item generates an infeasible subset.

For  $j = 1, 2, \dots, n$ ;  $h = 1, 2, \dots, k$ :

$A_{jh} = 1$  if item  $j$  belongs to feasible subset  $S_h$

$A_{jh} = 0$  otherwise

# Strong Formulation of the *BPP* (2)

$$x_h = \begin{cases} 1 & \text{if subset } S_h \text{ is selected} \\ 0 & \text{otherwise} \end{cases} \quad (h = 1, \dots, k)$$

$$\min \sum_{h=1,k} x_j$$

$$\sum_{h=1,k} A_{jh} x_j = 1 \quad (j = 1, \dots, n)$$

$$x_h \in \{0, 1\} \quad (h = 1, \dots, k)$$

**Set Partitioning Formulation (BLP Model)**

Consider only “maximal” feasible subsets  $S_h$

$$\sum_{h=1,k} A_{jh} x_j \geq 1 \quad (j = 1, \dots, n)$$

**Set Covering Formulation (BLP Model)**

# *Transportation Problem (TP)*

*Given: m origins and n destinations:*

$a_i$  amount of product to be transported from origin  $i$   
( $i = 1, \dots, m$ ),  $a_i \geq 0$ ;

$b_j$  amount of product to be transported to destination  $j$   
( $j = 1, \dots, n$ ),  $b_j \geq 0$ ;

$c_{ij}$  cost for transporting one unit of product from origin  $i$   
to destination  $j$  ( $i = 1, \dots, m; j = 1, \dots, n$ ):

Determine the *amount of product* ( $x_{ij}$ ) to be transported from *each origin*  $i$  ( $i = 1, \dots, m$ ) to *each destination*  $j$  ( $j = 1, \dots, n$ ) so as to *minimize the global cost*.

# Mathematical Model of *TP*

$$\min \quad \sum_{i=1,m} \sum_{j=1,n} c_{ij} x_{ij}$$

$$\sum_{j=1,n} x_{ij} = a_i \quad (i = 1, \dots, m)$$

$$\sum_{i=1,m} x_{ij} = b_j \quad (j = 1, \dots, n)$$

$$x_{ij} \geq 0 \quad (i = 1, \dots, m, j = 1, \dots, n)$$

LP model

*TP* is a polynomial problem

If  $a_i$  and  $b_j$  are integer:  $x_{ij}$  integer

*AP* is a special case of *TP*

# Sequencing of Jobs

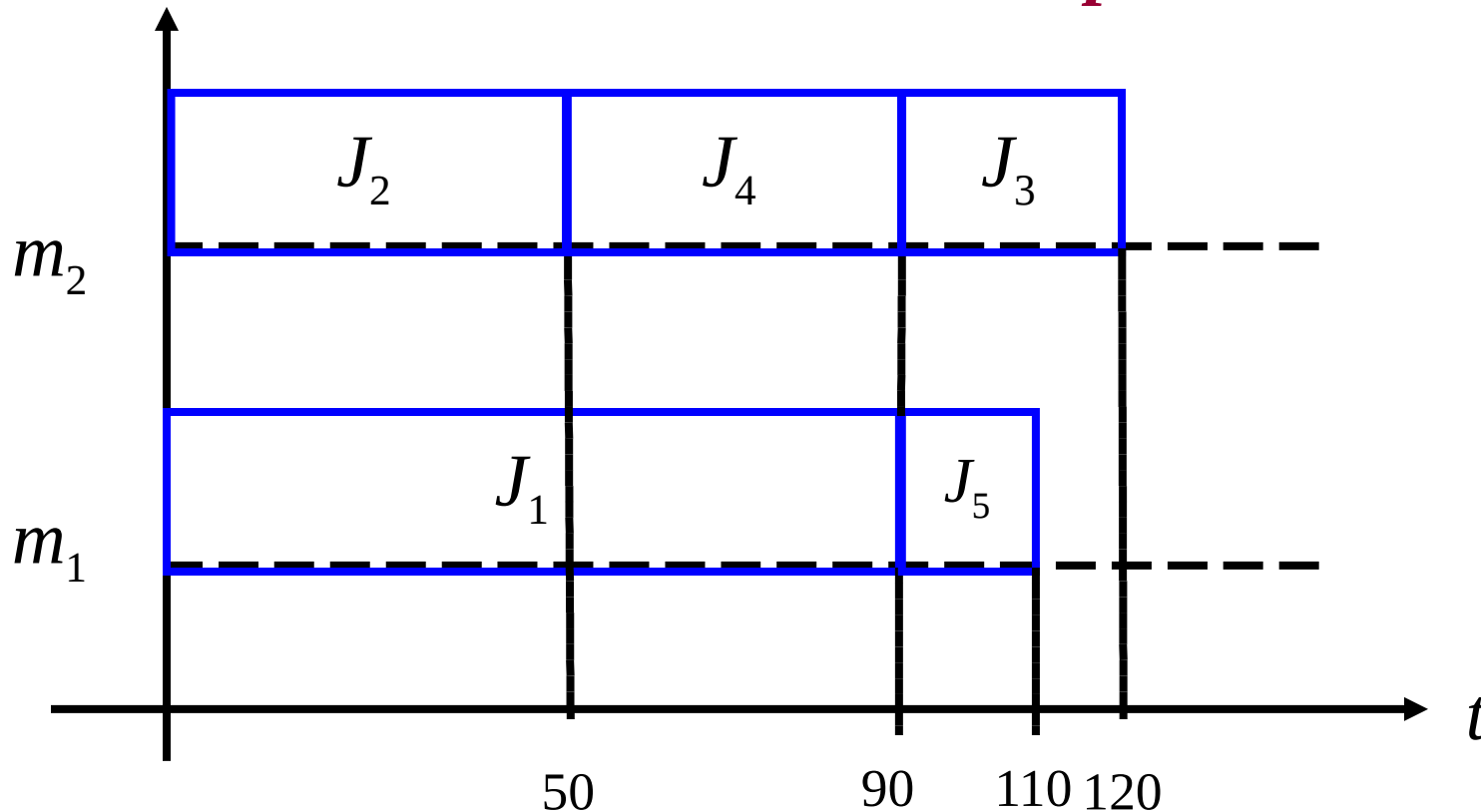
## Given:

- $n$  jobs
- $m$  identical machines
- $p_j$  processing time of job  $j$  ( $j = 1, 2, \dots, n$ ),  $p_j > 0$
- “no preemption” = the processing of a job cannot be interrupted;
- a machine cannot process more than one job at the same time;
- assign the jobs to the machines so as to minimize the time at which all the machines have finished to process the assigned jobs (“makespan”).

# Sequencing of Jobs (2)

- *Example:  $n = 5$ ,  $m = 2$ ,  $p_j = (90, 50, 30, 40, 20)$*

*makespan = 120*





# Sequencing of Jobs (3)

- $n = 5$ ,  $m = 2$ ,  $p_j = \{90, 50, 30, 40, 20\}$
- $Z = \text{makespan} = 120$
- $\text{LB} = \text{Lower Bound} = \sum_{j=1,n} P_j / m = 230 / 2 = 115$
- $Z = \text{value of the optimal solution}$
- *optimal solution : machine 1: jobs 1 and 5  
machine 2: jobs 2, 3 and 4*

# Mathematical Model

$$x_{ij} = \begin{cases} 1 & \text{if machine } i \text{ processes job } j \\ 0 & \text{otherwise} \end{cases} \quad (i = 1, \dots, m; j = 1, \dots, n)$$

**min**            **z**

$$\sum_{j=1, n} p_j x_{ij} \leq z \quad (i = 1, \dots, m)$$

$$\sum_{i=1, m} x_{ij} = 1 \quad (j = 1, \dots, n)$$

$$x_{ij} \in \{0, 1\} \quad (i = 1, \dots, m; j = 1, \dots, n)$$

$$z \geq 0$$

MLP model

**The Job Sequencing Problem is NP-Hard**