

DM841  
Discrete Optimization

**Exercises**

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# Outline

1. Examples  
SAT

2. Examples

To activate the `Observer` from command line with `--main::observer 1`:

1. declare the parameter from command line:

```
Parameter<unsigned> observer("observer", "Attach the observers", main_parameters);
```

2. declare the observer:

```
RunnerObserver<BDS_Input, BDS_State, FlipOrSwap, int> ob(observer, 0);
```

3. attach the observer to the runner:

```
if (observer.IsSet())  
    bds_sa.AttachObserver(ob);
```

You can print the number of iterations made by the runner with

```
runner.Iteration()
```

this can be useful for assessing speedups.

It is possible to set the time limit from command line. In order to do this you have to move the declaration of the Solver, for example:

```
SimpleLocalSearch<XYZ_Input, XYZ_Output, XYZ_State, int> XYZ_solver(in, XYZ_sm,  
XYZ_om, "XYZ solver");
```

before the second call of `CommandLineParameters::Parse()`

# Efficiency and Effectiveness

After implementation and test of the above components, improvements in **efficiency** (ie, computation time) can be achieved by:

- A. fast incremental evaluation (ie, delta evaluation)
- B. neighborhood pruning
- C. clever use of data structures

Improvements in **effectiveness**, ie, quality, can be achieved by:

- D. application of a metaheuristic
- E. definition of a larger neighborhood

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Notation:

- ▶  $n$  0-1 variables  $x_j$ ,  $j \in N = \{1, 2, \dots, n\}$ ,
- ▶  $m$  clauses  $C_i$ ,  $i \in M$ , and weights  $w_i$  ( $\geq 0$ ),  $i \in M = \{1, 2, \dots, m\}$
- ▶  $\max_{\mathbf{a} \in \{0,1\}^n} \sum \{w_i \mid i \in M \text{ and } C_i \text{ is satisfied in } \mathbf{a}\}$
- ▶  $\bar{x}_j = 1 - x_j$
- ▶  $L = \bigcup_{j \in N} \{x_j, \bar{x}_j\}$  set of literals
- ▶  $C_i \subseteq L$  for  $i \in M$  (e.g.,  $C_i = \{x_1, \bar{x}_3, x_8\}$ ).



Let's take the case  $w_i = 1$  for all  $i \in M$

- ▶ Assignment:  $\mathbf{a} \in \{0, 1\}^n$
- ▶ Evaluation function:  $f(\mathbf{a}) = \#$  unsatisfied clauses
- ▶ Neighborhood: one-flip
- ▶ Pivoting rule: best neighbor

Naive approach: exhaustive neighborhood examination in  $O(nmk)$  ( $k$  size of largest  $C_i$ )

A better approach:

- ▶  $C(x_j) = \{i \in M \mid x_j \in C_i\}$  (i.e., clauses dependent on  $x_j$ )
- ▶  $L(x_j) = \{l \in N \mid \exists i \in M \text{ with } x_l \in C_i \text{ and } x_j \in C_i\}$
- ▶  $f(\mathbf{a}) = \#$  unsatisfied clauses
- ▶  $\Delta(x_j) = f(\mathbf{a}) - f(\mathbf{a}')$ ,  $\mathbf{a}' = \delta_{1E}^{x_j}(\mathbf{a})$  (score of  $x_j$ )

Initialize:

- ▶ compute  $f$ , score of each variable, and list unsat clauses in  $O(mk)$
- ▶ init  $C(x_j)$  for all variables

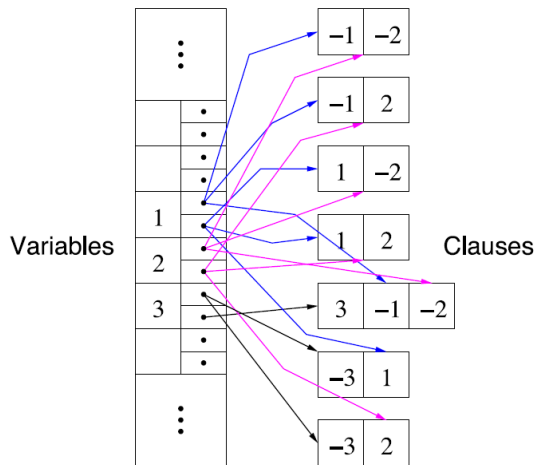
Examine Neighborhood

- ▶ choose the var with best score

Update:

- ▶ change the score of variables affected, that is, look in  $L(\cdot)$  and  $C(\cdot)$   $O(mk)$

## $C(x_j)$ Data Structure



Even better approach (though same asymptotic complexity):

↪ after the flip of  $x_j$  only the score of variables in  $L(x_j)$  that **critically depend** on  $x_j$  actually changes

- ▶ Clause  $C_i$  is **critically satisfied** by a variable  $x_j$  in  $\mathbf{a}$  iff:
  - ▶  $x_j$  is in  $C_i$
  - ▶  $C_i$  is satisfied in  $\mathbf{a}$  and flipping  $x_j$  makes  $C_i$  unsatisfied (e.g.,  $1 \vee 0 \vee 0$  but not  $1 \vee 1 \vee 0$ )

Keep a list of such clauses for each var

- ▶  $x_j$  is **critically dependent** on  $x_l$  under  $\mathbf{a}$  iff:  
there exists  $C_i \in C(x_j) \cap C(x_l)$  and such that flipping  $x_j$ :
  - ▶  $C_i$  changes satisfaction status
  - ▶  $C_i$  changes satisfied /critically satisfied status

Initialize:

- ▶ compute score of variables;
- ▶ init  $C(x_j)$  for all variables
- ▶ init status criticality for each clause

Update:

change sign to score of  $x_j$

for all  $C_i$  in  $C(x_j)$  do

┌ for all  $x_l \in C_i$  do

└ update score  $x_l$  depending on its critical status before flipping  $x_j$

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# Examples

- ▶ Permutations
  - ▶ TSP
  - ▶ SMWTP
- ▶ Assignments
  - ▶ SAT
  - ▶ Coloring
  - ▶ Parallel machines
- ▶ Sets
  - ▶ Max Weighted Independent Set
  - ▶ Steiner tree

# Single Machine Total Weighted Tardiness

**Given:** a set of  $n$  jobs  $\{J_1, \dots, J_n\}$  to be processed on a single machine and for each job  $J_i$  a processing time  $p_i$ , a weight  $w_i$  and a due date  $d_i$ .

**Task:** Find a schedule that minimizes

the total weighted tardiness  $\sum_{i=1}^n w_i \cdot T_i$

where  $T_i = \max\{C_i - d_i, 0\}$  ( $C_i$  completion time of job  $J_i$ )

Example:

Job	$J_1$	$J_2$	$J_3$	$J_4$	$J_5$	$J_6$
Processing Time	3	2	2	3	4	3
Due date	6	13	4	9	7	17
Weight	2	3	1	5	1	2

Sequence  $\phi = J_3, J_1, J_5, J_4, J_2, J_6$

Job	$J_3$	$J_1$	$J_5$	$J_4$	$J_2$	$J_6$
$C_i$	2	5	9	12	14	17
$T_i$	0	0	2	3	1	0
$w_i \cdot T_i$	0	0	2	15	3	0

# The Max Independent Set Problem

Also called “stable set problem” or “vertex packing problem”.

**Given:** an undirected graph  $G(V, E)$  and a non-negative weight function  $\omega$  on  $V$  ( $\omega : V \rightarrow \mathbf{R}$ )

**Task:** A largest weight **independent set** of vertices, i.e., a subset  $V' \subseteq V$  such that no two vertices in  $V'$  are joined by an edge in  $E$ .

Related Problems:

## Vertex Cover

**Given:** an undirected graph  $G(V, E)$  and a non-negative weight function  $\omega$  on  $V$  ( $\omega : V \rightarrow \mathbf{R}$ )

**Task:** A smallest weight **vertex cover**, i.e., a subset  $V' \subseteq V$  such that each edge of  $G$  has at least one endpoint in  $V'$ .

## Maximum Clique

**Given:** an undirected graph  $G(V, E)$

**Task:** A maximum cardinality **clique**, i.e., a subset  $V' \subseteq V$  such that every two vertices in  $V'$  are joined by an edge in  $E$

# Graph Partitioning

**Input:** A graph  $G = (V, E)$ , weights  $w(v) \in \mathbb{Z}^+$  for each  $v \in V$  and  $l(e) \in \mathbb{Z}^+$  for each  $e \in E$ .

**Task:** Find a partition of  $V$  into disjoint sets  $V_1, V_2, \dots, V_m$  such that  $\sum_{v \in V_i} w(v) \leq K$  for  $1 \leq i \leq m$  and such that if  $E' \subseteq E$  is the set of edges that have their two endpoints in two different sets  $V_i$ , then  $\sum_{e \in E'} l(e)$  is minimal.

Consider the specific case of graph bipartitioning, that is, the case  $|V| = 2n$  and  $K = n$  and  $w(v) = 1, \forall v \in V$ .



# Example: Scheduling in Parallel Machines

## Total Weighted Completion Time on Unrelated Parallel Machines Problem

**Input:** A set of jobs  $J$  to be processed on a set of parallel machines  $M$ . Each job  $j \in J$  has a weight  $w_j$  and processing time  $p_{ij}$  that depends on the machine  $i \in M$  on which it is processed.

**Task:** Find a schedule of the jobs on the machines such that the sum of weighted completion time of the jobs is minimal.

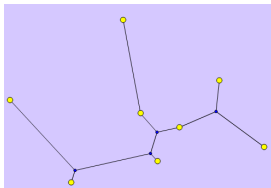
# Example: Steiner Tree

## Steiner Tree Problem

**Input:** A graph  $G = (V, E)$ , a weight function  $\omega : E \mapsto \mathbf{N}$ , and a subset  $U \subseteq V$ .

**Task:** Find a Steiner tree, that is, a subtree  $T = (V_T, E_T)$  of  $G$  that includes all the vertices of  $U$  and such that the sum of the weights of the edges in the subtree is minimal.

Vertices in  $U$  are the special vertices and vertices in  $S = V \setminus U$  are Steiner vertices.



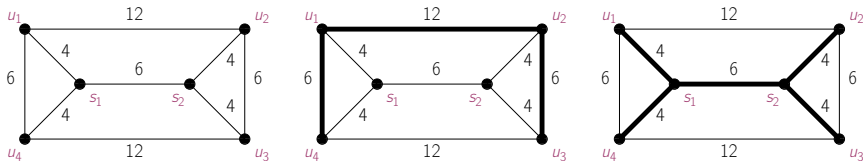


Figure : Vertices  $u_1, u_2, u_3, u_4$  belong to the set  $U$  of special vertices to be covered and vertices  $s_1, s_2$  belong to the set  $S$  of Steiner vertices. The Steiner tree in the second graph has cost 24 while the one in the third graph has cost 22.

1. Design one or more local search algorithms for the Steiner tree problem. In particular, define the solution representation and the neighborhood function.
2. Provide an analysis of the computational cost of the basic operations in the local search algorithms designed at the previous point. In particular, consider the size of the neighborhood, and the cost of evaluating a neighbor.