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## Research Report

MIP formulation of the adjustment  
problem for the MST and MHP  
problems: computational results

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# MIP formulation of the adjustment problem for the MST and MHP problems: computational results

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## Abstract

We consider a pair of optimization problems for a given weighted undirected graph: the minimum spanning tree (MST) problem and its restriction, the minimum Hamiltonian path (MHP) problem. For this pair of problems the adjustment problem consists in finding such minimum norm perturbations of weights of edges, which guarantee that the set of optimal solutions of the MST problem in the perturbed graph contains a Hamiltonian path.

In this paper we consider a mixed integer programming (MIP) formulation of the adjustment problem and describe computational results obtained for randomly generated graphs.

# 1 Introduction

The adjustment problem described in [1] is formulated for a pair of optimization problems: an initial optimization problem with linear objective function

$$v(c, X) = \max\{c^T x : x \in X\}, \quad (1)$$

where  $c \in \mathbb{R}^n$  and  $X \subseteq \mathbb{R}^n$ , and its restriction for a given subset  $F \subseteq X$  of feasible solutions:

$$\max\{c^T x : x \in F\}. \quad (2)$$

Given such a pair of problems one wants to adjust the objective function coefficients in problem (1), such that an optimal solution of the perturbed problem is also a feasible solution for the restriction (2). The adjustment problem seeks among admissible perturbations, one that is least costly according to some given norm.

Let  $\Delta \subseteq \mathbb{R}^n$  be the set of all admissible perturbations of the original vector of weights  $c$ , and let  $\|\delta\|$  denotes a norm of  $\delta \in \mathbb{R}^n$ .

The adjustment problem related to  $F$  and  $\Delta$  is formally stated as follows:

$$a(F, \Delta) = \min\{\|\delta\| : v(c + \delta, X) = v(c + \delta, F), \delta \in \Delta\}. \quad (3)$$

Given sets  $\Delta \subseteq \mathbb{R}^n$  and  $F \subseteq X$ , the optimal value  $a(F, \Delta)$  of the problem (3) is called the adjustment cost with respect to  $F$  and  $\Delta$ . In this paper we will consider only  $l_1$  norm, i.e., for  $\delta = \delta^+ - \delta^-$ ,  $\delta^+, \delta^- \in \mathbb{R}_+^n = \{x \in \mathbb{R}^n : x \geq 0\}$ ,  $\|\delta\| = \mathbf{1}^T \delta^+ - \mathbf{1}^T \delta^-$ , where  $\mathbf{1} \in \mathbb{R}^n$  is a vector of ones. Moreover, we will take  $\Delta = \mathbb{R}^n$ .

In the following we assume that the initial optimization problem (1) is a linear programming problem:

$$\max c^T x \quad (4a)$$

$$A'x \leq b' \quad (4b)$$

$$A''x = b'' \quad (4c)$$

$$x \geq 0, \quad (4d)$$

where  $A' \in \mathbb{R}^{m' \times n}$ ,  $A'' \in \mathbb{R}^{m'' \times n}$ ,  $b' \in \mathbb{R}^{m'}$  and  $b'' \in \mathbb{R}^{m''}$ .

A restriction of this problem will be defined by adding new linear constraints

$$Cx \leq d, \quad (5)$$

where  $C \in \mathbb{R}^{p \times n}$ ,  $d \in \mathbb{R}^p$ , and requiring that feasible solutions belong to the set  $\mathbb{B}^n = \{0, 1\}^n$ . Thus, the restricted problem (2) considered in the paper

has the form:

$$\begin{aligned} & \max c^T x \\ & A'x \leq b' \\ & A''x = b'' \\ & Cx \leq d \\ & x \in \mathbb{B}^n. \end{aligned} \tag{6}$$

In [1] it is shown that the adjustment problem for a pair of problems (4) and (6) and  $\|\delta\| = l_1$ ,  $\Delta = \mathbb{R}^n$ , can be stated as follows:

$$\min(1^T \delta^+ - 1^T \delta^-) \tag{7a}$$

$$\left\{ \begin{array}{l} (A')^T y' - \delta^+ + \delta^- \geq c \\ (A'')^T y'' - \delta^+ + \delta^- \geq c \\ y' \geq 0, y'' \in \mathbb{R}^{m''} \end{array} \right. \tag{7b}$$

$$(b')^T y' + (b'')^T y'' - c^T x - (\delta^+)^T x + (\delta^-)^T x = 0 \tag{7c}$$

$$\left\{ \begin{array}{l} A'x \leq b' \\ A''x = b'' \\ Cx \leq d \\ x \in \mathbb{B}^n \end{array} \right. \tag{7d}$$

$$\delta^+, \delta^- \geq 0. \tag{7e}$$

In the above problem the constraints (7b) and variables  $y', y''$ , correspond to the dual problem for linear program (4), whereas constraints (7d) and variables  $x$  correspond to the restricted problem (6).

The nonlinear constraint (7c) states that the value of the objective function in the perturbed initial linear program (4) is equal to the value of the objective function in its dual. To solve problem (7) with standard optimization packages for MIP it is necessary to linearize the nonlinear terms  $(\delta^+)^T x$  and  $(\delta^-)^T x$ . In the next section we describe this step and give all details for MIP formulation of the adjustment problem for a pair of problems: the minimum spanning tree (MST) problem, regarded as the initial optimization problem (4), and the minimum Hamiltonian path problem – as its restriction (6).

## 2 MST, MHP and the adjustment problems

Let  $G = (V, E, c)$  be a connected weighted graph, where  $V = \{1, \dots, n\}$ ,  $E \subseteq \{\{i, j\} : i, j \in V\}$ ,  $c \in \mathbb{R}^{|E|}$ . Our goal will be to find such perturbations  $\delta \in \mathbb{R}^{|E|}$  of the weight vector  $c$ , that some minimum spanning tree in the perturbed graph is a Hamiltonian path in this graph. First, we will formulate the minimum spanning tree problem in  $G$  as the initial linear program (4). Then, we describe additional constraints to get the minimum Hamiltonian path problem as the restriction (6). Finally, we give the adjustment problem (7) for this pair of optimization problems.

We use the following notation:

$$\begin{aligned} K &= V \setminus \{1\}, \\ D &= \{(i, j) \in V \times V : \{i, j\} \in E\}. \end{aligned} \tag{8}$$

Define for  $(i, j) \in D$  nonnegative variables  $x_{ij} \in \mathbb{R}$  and introduce for  $k \in K$  and  $(i, j) \in D$  auxiliary nonnegative variables  $f_{ij}^k \in \mathbb{R}$ . It is known (see e.g. [2]), that the following continuous linear programming problem (9) may be used to calculate the minimum weight spanning tree  $T^*$  in graph  $G = (V, E, c)$ . Namely, given an optimal solution  $x^*, f^*$  of problem (9), the optimal tree  $T^*$  is composed of the edges  $\{i, j\} \in E$  for which  $x_{ij}^* + x_{ji}^* > 0$ .

$$\max \sum_{\{i, j\} \in E} -c_{ij} \cdot (x_{ij} + x_{ji}) \tag{9a}$$

$$x_{ij} - f_{ij}^k \geq 0, \quad k \in K, (i, j) \in D \tag{9b}$$

$$\begin{cases} \sum_{(i, j) \in D} x_{ij} = n - 1 \\ \sum_{(j, 1) \in D} f_{j1}^k - \sum_{(1, j) \in D} f_{1j}^k = -1, \quad k \in K \\ \sum_{(j, k) \in D} f_{jk}^k - \sum_{(k, j) \in D} f_{kj}^k = 1, \quad k \in K \\ \sum_{(j, i) \in D} f_{ji}^k - \sum_{(i, j) \in D} f_{ij}^k = 0, \quad i, k \in K, i \neq k \end{cases} \tag{9c}$$

$$\begin{cases} x_{ij} \geq 0, \quad (i, j) \in D \\ f_{ij}^k \geq 0, \quad k \in K, (i, j) \in D. \end{cases} \tag{9d}$$

Parts (9a), (9b), (9c) in the above formulation of the MST problem correspond, respectively, to (4a), (4b), (4c) in the initial linear program (4).

To get the MHP problem, being a restriction of (9), it is necessary to add the following constraints:

$$\sum_{j \in V} (x_{ij} + x_{ji}) \leq 2, \quad i \in V, \quad (10)$$

$$x_{ij} \in \mathbb{B}, \quad (i, j) \in D. \quad (11)$$

Observe that constraints (10) correspond to linear constraints (5). The obtained optimization problem corresponds to (6).

Let  $w_{ij}^k \geq 0$  for  $k \in K$ ,  $(i, j) \in D$ , be dual variables for inequalities (9b), and  $l \in \mathbb{R}$ ,  $v_1^k \in \mathbb{R}$  for  $k \in K$ ,  $v_k^k \in \mathbb{R}$  for  $k \in K$ ,  $v_i^k \in \mathbb{R}$  for  $i, k \in K$ ,  $i \neq k$  denote dual variables for consecutive groups of constraints in problem (9c). The dual problem of linear program (9) can be stated as follows:

$$\min \sum_{k \in K} -(v_k^k - v_1^k) - (n - 1) \cdot l \quad (12a)$$

$$\left\{ \begin{array}{l} v_j^k - v_i^k - w_{ij}^k \leq 0, \quad k \in K, (i, j) \in D \\ \sum_{k \in K} w_{ij}^k + l \leq c_{ij}, \quad \{i, j\} \in E \\ \sum_{k \in K} w_{ji}^k + l \leq c_{ij}, \quad \{i, j\} \in E \\ w_{ij}^k \geq 0, \quad k \in K, (i, j) \in D \\ v_i^k \in \mathbb{R}, \quad i \in V, k \in K \\ l \in \mathbb{R}. \end{array} \right. \quad (12b)$$

We may now formulate the adjustment problem (7) for the considered pair of problems. It is enough to use constraints (12b) as constraints (7b) and constraints (9b), (9c), (9d), (10), (11) as constraints (7d). The constraint (7c) corresponds now to the objective functions (9a), (12a), and has the following form:

$$\sum_{k \in K} (v_k^k - v_1^k) + (n - 1) \cdot l - \sum_{(i, j) \in E} c_{ij} (x_{ij} + x_{ji}) - \sum_{(i, j) \in D} (\delta_{ij}^+ - \delta_{ij}^-) x_{ij} = 0, \quad (13)$$

where  $\delta_{ij}^+, \delta_{ij}^- \geq 0$  for  $(i, j) \in D$ .

Finally, the adjustment problem for a pair: the MST and the MHP problems is stated as follows:

$$\min \sum_{(i,j) \in D} (\delta_{ij}^+ - \delta_{ij}^-) \quad (14a)$$

$$\begin{cases} v_j^k - v_i^k - w_{ij}^k \leq 0, k \in K, (i,j) \in D \\ \sum_{k \in K} w_{ij}^k + l - \delta_{ij}^+ + \delta_{ij}^- \leq c_{ij}, \{i,j\} \in E \\ \sum_{k \in K} w_{ji}^k + l - \delta_{ji}^+ + \delta_{ji}^- \leq c_{ij}, \{i,j\} \in E \\ w_{ij}^k \geq 0, k \in K, (i,j) \in D \\ v_i^k \in \mathbb{R}, i \in V, k \in K \\ l \in \mathbb{R} \end{cases} \quad (14b)$$

$$\sum_{k \in K} (v_k^k - v_1^k) + (n-1) \cdot l - \sum_{\{i,j\} \in E} c_{ij}(x_{ij} + x_{ji}) - \sum_{(i,j) \in D} (\delta_{ij}^+ - \delta_{ij}^-) x_{ij} = 0 \quad (14c)$$

$$\begin{cases} \sum_{(i,j) \in D} x_{ij} = n - 1 \\ \sum_{(j,1) \in D} f_{j1}^k - \sum_{(1,j) \in D} f_{1j}^k = -1, k \in K \\ \sum_{(j,k) \in D} f_{jk}^k - \sum_{(k,j) \in D} f_{kj}^k = 1, k \in K \\ \sum_{(j,i) \in D} f_{ji}^k - \sum_{(i,j) \in D} f_{ij}^k = 0, i, k \in K, i \neq k \\ x_{ij} - f_{ij}^k \geq 0, k \in K, (i,j) \in D \\ x_{ij} \geq 0, (i,j) \in D \\ f_{ij}^k \geq 0, k \in K, (i,j) \in D \end{cases} \quad (14d)$$

$$\begin{cases} \sum_{j \in V} (x_{ij} + x_{ji}) \leq 2, i \in V \\ x_{ij} \in \{0, 1\}, (i,j) \in D \\ \delta_{ij}^+ = \delta_{ji}^+ \geq 0, (i,j) \in D \\ \delta_{ij}^- = \delta_{ji}^- \geq 0, (i,j) \in D. \end{cases} \quad (14e)$$

To linearize terms  $\delta_{ij}^+ x_{ij}$ ,  $\delta_{ij}^- x_{ij}$ ,  $(i,j) \in D$ , we use a standard technique (see e.g. [3]). Namely, we introduce new nonnegative variables  $z_{ij}^+$ ,  $z_{ij}^-$ ,  $(i,j) \in D$ , satisfying the following conditions:

$$z_{ij}^+ = \delta_{ij}^+ x_{ij}, (i,j) \in D, \quad (15a)$$

$$z_{ij}^- = \delta_{ij}^- x_{ij}, (i,j) \in D. \quad (15b)$$

Now we can replace the constraint (14c) in the problem (14), with the following linear constraint:

$$\sum_{k \in K} (v_k^k - v_1^k) + (n-1) \cdot l - \sum_{(i,j) \in E} c_{ij}(x_{ij} + x_{ji}) - \sum_{(i,j) \in D} z_{ij}^+ + \sum_{(i,j) \in D} z_{ij}^- = 0. \quad (16)$$

For any new variable  $z_{ij}^+, z_{ij}^-, (i, j) \in D$ , we have to add also constraints which would guarantee that equations (15a) and (15b) hold. Let us take for example the equation  $z_{ij}^+ = \delta_{ij}^+ x_{ij}$  for some ordered pair  $(i, j) \in D$ . This equation is equivalent to two implications:

$$\begin{aligned} x_{ij} = 0 &\Rightarrow z_{ij}^+ = 0, \\ x_{ij} = 1 &\Rightarrow z_{ij}^+ = \delta_{ij}^+, \end{aligned} \quad (17)$$

which can be modeled by adding the following new constraints:

$$\begin{aligned} z_{ij}^+ - Mx_{ij} &\leq 0, \\ -\delta_{ij}^+ + z_{ij}^+ &\leq 0, \\ \delta_{ij}^+ - z_{ij}^+ + Mx_{ij} &\leq M, \end{aligned} \quad (18)$$

where  $M$  is a sufficiently large constant satisfying the inequality  $\delta_{ij}^+ \leq M$  for any  $(i, j) \in D$ . It can be shown (see [1]) that for  $\Delta = \mathbb{R}^n$  we can take  $M = \sum_{(i,j) \in E} |c_{ij}|$ . So, to linearize all nonlinear terms in (14c) it is necessary to add for any  $(i, j) \in D$  the following inequalities:

$$\begin{aligned} z_{ij}^+ - Mx_{ij} &\leq 0 \\ -\delta_{ij}^+ + z_{ij}^+ &\leq 0 \\ \delta_{ij}^+ - z_{ij}^+ + Mx_{ij} &\leq M \\ z_{ij}^- - Mx_{ij} &\leq 0 \\ -\delta_{ij}^- + z_{ij}^- &\leq 0 \\ -\delta_{ij}^- - z_{ij}^- + Mx_{ij} &\leq M \\ z_{ij}^+, z_{ij}^- &\geq 0. \end{aligned} \quad (19)$$

### 3 Computational results

In this section we describe computational experiment provided to test our MIP approach for the adjustment problem.

We used randomly generated graphs with the number of vertices  $|V| = 11, \dots, 20$ . Following families of benchmarks were considered:

1. *complete1* – complete graphs with the number of vertices  $|V| = 11, \dots, 20$ , in which the weight of each edge is an integer random variable uniformly distributed in the interval  $[0; |V|^2]$ . For each number of vertices 10 instances were generated.
2. *complete2* – complete graphs with the number of vertices  $|V| = 11, \dots, 20$ , in which the weight of each edge is taken with the same probability from the set  $\{0, 1\}$ . For each number of vertices 10 instances were generated.
3. *density* – graphs with 20 vertices in which each edge appears with the probability  $d = 0.2, 0.3, \dots, 1.0$ . The weight of an edge is an integer random variable uniformly distributed in the interval  $[0; |V|^2]$ . For each density considered 10 instances were generated.
4. *data* – a family of three groups of benchmarks (*integer*, *boolean*, *real*), constructed to investigate the influence of data (discrete, continuous). In the *integer* and *real* group the weight of each edge is an integer or a continuous random variable uniformly distributed in the interval  $[0; |V|^2]$ , respectively. In the *boolean* group the weight of each edge is taken with the same probability from the set  $\{0, 1\}$ . For each group considered 10 instances of graphs with  $|V| = 20$  vertices were generated.

A computer program in Java was used to generate all benchmarks and related MIP problems. The appendix contains a listing of this program. To solve these MIP problems we use in our program ILOG CPLEX 10 solver libraries. For all instances optimal solutions for the adjustment problem were obtained. Resulting MIP problems for  $|V| = 20$  had roughly 17000 variables (including 380 binaries), 17500 constraints, and 66500 nonzeros.

Computational times (in seconds) for AMD Opteron Processor 248 with 4 GB RAM for families *complete1*, *complete2*, *density* and *data* are presented in the following tables and figures. For any 10 instances from these families, we computed the following statistic parameters:

1. *min* – minimum computational time,
2. *median* – median of computational times,
3. *avg* – average of computational times,
4. *stddev* – standard deviation of computational times,
5. *max* – maximum computational time.

| complete1<br>benchmarks | Number of vertices |      |      |       |       |        |        |        |        |         |
|-------------------------|--------------------|------|------|-------|-------|--------|--------|--------|--------|---------|
|                         | 11                 | 12   | 13   | 14    | 15    | 16     | 17     | 18     | 19     | 20      |
| min                     | 0.15               | 0.19 | 0.27 | 0.42  | 1.1   | 1.59   | 2.09   | 2.99   | 7.29   | 10.37   |
| median                  | 0.19               | 1.24 | 2.47 | 1.62  | 3.92  | 2.63   | 31.23  | 9.49   | 13.8   | 27.2    |
| avg                     | 0.38               | 2.35 | 3.78 | 6.17  | 10.78 | 25.53  | 87.64  | 42.77  | 101.41 | 252.49  |
| stddev                  | 0.59               | 2.67 | 3.6  | 7.69  | 17.66 | 35.44  | 148.78 | 54.35  | 198.91 | 516.4   |
| max                     | 2.07               | 7.42 | 9.95 | 22.39 | 58.52 | 103.08 | 492.37 | 158.64 | 644.21 | 1552.35 |

Table 1: Computational times (in seconds) for the *complete1* family.

| complete2<br>benchmarks | Number of vertices |      |      |      |       |       |       |        |        |         |
|-------------------------|--------------------|------|------|------|-------|-------|-------|--------|--------|---------|
|                         | 11                 | 12   | 13   | 14   | 15    | 16    | 17    | 18     | 19     | 20      |
| min                     | 0.11               | 0.3  | 0.56 | 0.8  | 1.82  | 2.69  | 5.96  | 9.08   | 11.89  | 16.94   |
| median                  | 0.19               | 0.75 | 1.55 | 2.02 | 3.45  | 14.83 | 33.74 | 83.57  | 115.31 | 300.77  |
| avg                     | 0.19               | 0.67 | 1.57 | 2.61 | 6.56  | 23.73 | 37.27 | 88.67  | 192.18 | 332.26  |
| stddev                  | 0.06               | 0.28 | 0.86 | 2.16 | 5.68  | 28.43 | 19.01 | 54.84  | 265.18 | 309.52  |
| max                     | 0.33               | 1.1  | 2.92 | 6.77 | 18.15 | 95.04 | 75.12 | 205.92 | 794.86 | 1042.16 |

Table 2: Computational times (in seconds) for the *complete2* family.

| density<br>benchmarks | Graphs density |       |       |        |        |        |        |         |         |  |
|-----------------------|----------------|-------|-------|--------|--------|--------|--------|---------|---------|--|
|                       | 0.2            | 0.3   | 0.4   | 0.5    | 0.6    | 0.7    | 0.8    | 0.9     | 1.0     |  |
| min                   | 0.6            | 0.51  | 0.91  | 1.5    | 1.96   | 2.94   | 3.63   | 13.68   | 10.37   |  |
| median                | 4.29           | 4.81  | 13.58 | 15.34  | 106.31 | 26.65  | 82.61  | 327.24  | 27.2    |  |
| avg                   | 4.34           | 8.66  | 29.06 | 38.03  | 164.58 | 46.43  | 215.98 | 683.23  | 252.49  |  |
| stddev                | 2.96           | 9.31  | 37.33 | 47.62  | 173.37 | 54.04  | 275.54 | 734.15  | 516.4   |  |
| max                   | 9.47           | 27.95 | 98.11 | 139.64 | 493.52 | 182.67 | 830.16 | 1627.78 | 1552.35 |  |

Table 3: Computational times (in seconds) for the *density* family.

| data<br>benchmarks | Data type |         |         |
|--------------------|-----------|---------|---------|
|                    | boolean   | integer | real    |
| min                | 16.94     | 10.37   | 9.19    |
| median             | 300.77    | 27.2    | 124.14  |
| avg                | 332.26    | 252.49  | 534.27  |
| stddev             | 309.52    | 516.4   | 764.78  |
| max                | 1042.16   | 1552.35 | 1980.44 |

Table 4: Computational times (in seconds) for the *data* family.

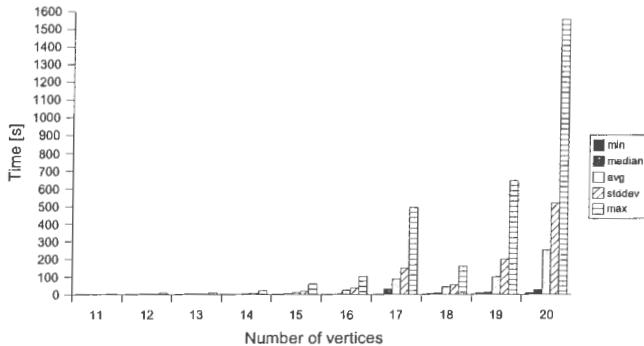


Figure 1: Computational times (in seconds) for the *complete1* family.

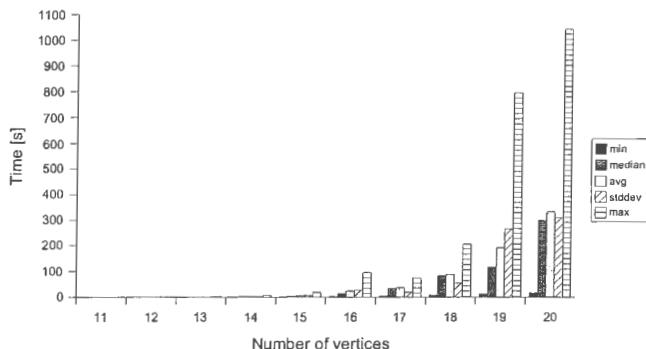


Figure 2: Computational times (in seconds) for the *complete2* family.

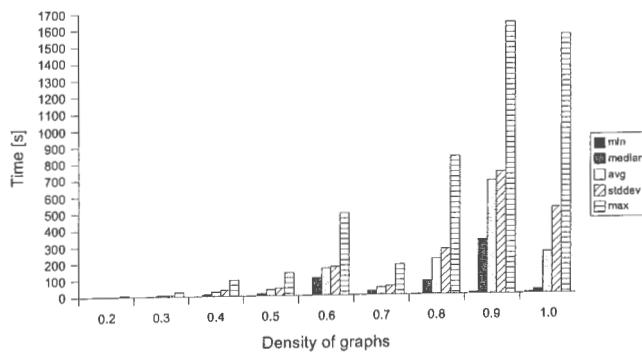


Figure 3: Computational times (in seconds) for the *density* family.

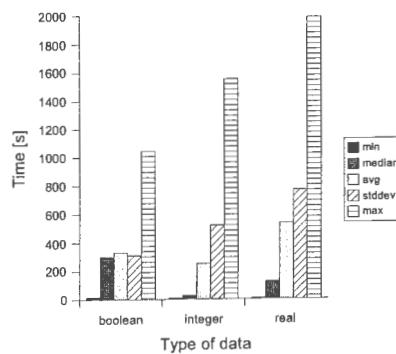


Figure 4: Computational times (in seconds) for the *data* family.

In Table 1 and Figure 1, Table 2 and Figure 2, Table 3 and Figure 3 computational results for families *complete1*, *complete2*, *density* are summarized, respectively. One can observe very fast grow of the computational time as a function of the number of vertices (*complete1* and *complete2* families) and the density (*density* family) of graphs. Only for benchmarks with 18 vertices for the *complete1* family, 17 vertices for the *complete2* family, 0.7 and 1.0 densities for the *density* family, a decrease of the computational time appeared; probably it is related to changes in ILOG CPLEX optimization strategies, but this problem requires further investigations.

In Table 4 and Figure 4 computational results for *data* family of three groups of benchmarks (*integer*, *boolean*, *real*) are summarized. The relations of the computational times for these groups are rather complicated. The *max* and *stddev* values are the smallest for problems from *boolean* group, and the largest for *real* group, but the *median* value appeared the smallest for *integer* group, and the largest for *boolean* group.

Our computational results were rather limited, for example only standard ILOG CPLEX optimization strategies were used, but all the results indicate that the MIP formulation of the adjustment problem for the considered pair of problems (MST and MHP) leads to difficult optimization problems and is limited to graphs with small number of vertices.

## References

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## 4 The appendix

This section contains a listing of Java program used to generate all benchmarks and related MIP problems presented in this paper.

```
1 import java.util.*;
2 import java.io.*;
3 import ilog.cplex.*;
4 import ilog.concert.*;
5
6
7 public class Global {
8     final static int MaxGraphSize = 1000;
9     public final static double inf = Double.POSITIVE_INFINITY;
10    static Random randomizer = new Random(
11        java.lang.System.currentTimeMillis());
12
13    public Global() {
14    }
15
16    public static void initRandomizer(long seed) {
17        randomizer = new Random(seed);
18    }
19}
20
21
22 public class Edge {
23    protected int x = -1;
24    protected int y = -1;
25    protected double value = 0.0;
26
27    public Edge(int newX, int newY) {
28        setEdge(newX, newY);
29    }
30
31    protected void setEdge(int newX, int newY) {
32        if(newX < newY) {
33            x = newX;
34            y = newY;
35        }
36        else {
37            x = newY;
38            y = newX;
39        }
40    }
41
42    public Edge(int newX, int newY, double newValue) {
43        setEdge(newX, newY);
44        value = newValue;
```

```

45     }
46
47     public double getValue() {
48         return value;
49     }
50
51     public void setValue(double newValue) {
52         value = newValue;
53     }
54
55     public int getX() {
56         return x;
57     }
58
59     public int getY() {
60         return y;
61     }
62
63     public int hashCode() {
64         return x * Global.MaxGraphSize + y;
65     }
66
67     public String toString() {
68         return "(" + x + "," + y + ")=" + value;
69     }
70 }
71
72
73 public class Graph {
74     protected LinkedHashMap<Integer, Edge> edges;
75     protected LinkedHashSet<Integer> vertices;
76     public int liczIteracje = 0;
77     public int licz = 0;
78
79     public Graph() {
80         clear();
81     }
82
83     public int verticesNumber() {
84         return vertices.size();
85     }
86
87     public int edgesNumber() {
88         return edges.size();
89     }
90
91     public Graph(int size) throws Exception {
92         if(size < 1 || size > Global.MaxGraphSize) {
93             throw new Exception("graph size out of range");

```

```

94    }
95    edges = new LinkedHashMap<Integer, Edge>();
96    vertices = new LinkedHashSet<Integer>();
97    for(int i=0; i<size; i++) {
98        addVertex(i);
99    }
100}
101
102 public Graph(int newSize, double density, double lbound, double
103             ubound, boolean isInteger) throws Exception {
104     randomGraph(newSize, density, lbound, ubound, isInteger);
105 }
106
107 public void randomGraph(int size, double density, double lbound,
108                         double ubound, boolean isInteger) throws Exception {
109     clear();
110    if(density < 0.0 || density > 1.0) {
111        throw new Exception("graph density out of range");
112    }
113    if(size < 1 || size > Global.MaxGraphSize) {
114        throw new Exception("graph size out of range");
115    }
116    ArrayList<Edge> edgeList = new ArrayList<Edge>();
117    for(int i=0; i<size; i++) {
118        addVertex(i);
119        for(int j=i+1; j<size; j++) {
120            double value = Global.randomizer.nextDouble() *
121                (ubound - lbound) + lbound;
122            if(isInteger) {
123                value = (double) Math.round(value);
124            }
125            edgeList.add(new Edge(i, j, value));
126        }
127    }
128    for(int i=(int)(density*edgeList.size()); i>0; i--) {
129        int edgeNumber = (int) (edgeList.size() *
130                               Global.randomizer.nextDouble());
131        addEdge(edgeList.get(edgeNumber));
132        edgeList.remove(edgeNumber);
133    }
134    sortEdgesLexOrder();
135 }
136
137 public void sortEdgesLexOrder() {
138    Integer[] edgeHashesArray =
139        edges.keySet().toArray(new Integer[1]);
140    Arrays.sort(edgeHashesArray);
141    LinkedHashMap<Integer, Edge> newEdges =
142        new LinkedHashMap<Integer, Edge>();

```

```

143         for(int i=0; i<edgeHashesArray.length; i++) {
144             newEdges.put(edgeHashesArray[i],
145                         edges.get(edgeHashesArray[i]));
146         }
147         edges = newEdges;
148     }
149
150     public void sortVerticesLexOrder() {
151         Integer[] verticesArray =
152             vertices.toArray(new Integer[1]);
153         Arrays.sort(verticesArray);
154         LinkedHashSet<Integer> newVertices =
155             new LinkedHashSet<Integer>();
156         for(int i=0; i<verticesArray.length; i++) {
157             newVertices.add(verticesArray[i]);
158         }
159         vertices = newVertices;
160     }
161
162     public LinkedHashMap<Double, LinkedHashSet<Edge>>
163         sortEdgesByValues() {
164         sortEdgesLexOrder();
165         LinkedHashMap<Double, LinkedHashSet<Edge>>
166             edgesClassifiedByValues =
167                 new LinkedHashMap<Double, LinkedHashSet<Edge>>();
168         Iterator<Edge> edgesIter = edgesIterator();
169         while(edgesIter.hasNext()) {
170             Edge edge = edgesIter.next();
171             double value = edge.getValue();
172             if(!edgesClassifiedByValues.keySet().contains(value)) {
173                 edgesClassifiedByValues.put(value, new LinkedHashSet<Edge>());
174             }
175             edgesClassifiedByValues.get(value).add(edge);
176         }
177         Double[] sortedValues =
178             edgesClassifiedByValues.keySet().toArray(new Double[1]);
179         LinkedHashMap<Double, LinkedHashSet<Edge>> edgesClassifiedByValuesSorted
180             = new LinkedHashMap<Double, LinkedHashSet<Edge>>();
181         Arrays.sort(sortedValues);
182         for(int i=0; i<sortedValues.length; i++) {
183             double value = sortedValues[i];
184             edgesClassifiedByValuesSorted.put(value,
185                 edgesClassifiedByValues.get(value));
186         }
187         LinkedHashMap<Integer, Edge> edgesSortedByValues =
188             new LinkedHashMap<Integer, Edge>();
189         for(int i=0; i<sortedValues.length; i++) {
190             Iterator<Edge> edgesCategoryIter =
191                 edgesClassifiedByValuesSorted.get(

```

```

192             sortedValues[i]).iterator();
193         while(edgesCategoryIter.hasNext()) {
194             Edge edge = edgesCategoryIter.next();
195             edgesSortedByValues.put(edge.hashCode(), edge);
196         }
197     }
198     edges = edgesSortedByValues;
199     return edgesClassifiedByValuesSorted;
200 }
201
202 public Iterator<Edge> edgesIterator() {
203     return edges.values().iterator();
204 }
205
206 public Iterator<Integer> verticesIterator() {
207     return vertices.iterator();
208 }
209
210 public boolean equals(Object obj) {
211     if(obj.getClass() != Graph.class) return false;
212     Graph that = (Graph) obj;
213     if(vertices.size()!=that.vertices.size()) return false;
214     Iterator<Integer> verticesIter = vertices.iterator();
215     while(verticesIter.hasNext()) {
216         int vertex = verticesIter.next();
217         if(!that.vertices.contains(vertex)) return false;
218     }
219     if(edges.size()!=that.edges.size()) return false;
220     Iterator<Edge> edgesIter = edgesIterator();
221     while(edgesIter.hasNext()) {
222         Edge edge = edgesIter.next();
223         if(!that.edges.containsKey(edge.hashCode())) return false;
224     }
225     return true;
226 }
227
228 public boolean isClique() {
229     return edgesNumber()==(verticesNumber()*(verticesNumber()-1))/2;
230 }
231
232 public void addAllEdges(Graph g) throws Exception {
233     Iterator<Edge> gIter = g.edgesIterator();
234     while(gIter.hasNext()) {
235         Edge edge = gIter.next();
236         addEdge(edge);
237     }
238 }
239
240 public void addEdge(Edge edge) throws Exception {

```

```

241     if(containsEdge(edge)) {
242         throw new Exception("Edge " + edge.toString() +
243             " is already in graph");
244     }
245     addVertex(edge.getX());
246     addVertex(edge.getY());
247     edges.put(edge.hashCode(), edge);
248 }
249
250 public void addVertex(Integer v) {
251     vertices.add(v);
252 }
253
254 public void addEdge(int x, int y) throws Exception {
255     addEdge(x, y, 0.0);
256 }
257
258 public void addEdge(int x, int y, double value) throws Exception {
259     Edge edge = new Edge(x, y, value);
260     addEdge(edge);
261 }
262
263 public void removeEdge(int x, int y) {
264     Edge edge = new Edge(x, y);
265     edges.remove(edge.hashCode());
266 }
267
268 public void removeAllEdges(Graph g) {
269     Iterator<Edge> gIter = g.edgesIterator();
270     while(gIter.hasNext()) {
271         Edge edge = gIter.next();
272         this.removeEdge(edge.getX(), edge.getY());
273     }
274 }
275
276 public double getEdgeValue(Edge e) throws Exception {
277     if(!edges.keySet().contains(e.hashCode())) {
278         throw new Exception("Edge not in graph");
279     }
280     return edges.get(e.hashCode()).getValue();
281 }
282
283 public double getEdgeValue(int x, int y) throws Exception {
284     return getEdgeValue(new Edge(x,y));
285 }
286
287 public double getGraphValue() {
288     double value = 0.0;
289     Iterator<Edge> graphIter = edgesIterator();

```

```

290         while(graphIter.hasNext()) {
291             Edge edge = graphIter.next();
292             value += edge.getValue();
293         }
294         return value;
295     }
296
297     public void clear() {
298         this.edges = new LinkedHashMap<Integer, Edge>();
299         this.vertices = new LinkedHashSet<Integer>();
300     }
301
302     public boolean containsEdge(Edge edge) {
303         return edges.containsKey(edge.hashCode());
304     }
305
306     public boolean containsEdge(int x, int y) {
307         Edge edge = new Edge(x, y);
308         return containsEdge(edge);
309     }
310
311     public String toString() {
312         String string = "[V={";
313         Iterator<Integer> verticesIter = vertices.iterator();
314         while(verticesIter.hasNext()) {
315             Integer vertex = verticesIter.next();
316             string = string.concat(vertex.toString());
317             if(verticesIter.hasNext()) {
318                 string = string.concat(", ");
319             }
320         }
321         string = string.concat("}, E={");
322         Iterator<Edge> graphIter = this.edgesIterator();
323         while(graphIter.hasNext()) {
324             Edge edge = graphIter.next();
325             string = string.concat(edge.toString());
326             if(graphIter.hasNext()) {
327                 string = string.concat(", ");
328             }
329         }
330         string = string.concat("}]=") + getGraphValue();
331         return string;
332     }
333
334     public Graph copy() throws Exception {
335         Graph newGraph = new Graph();
336         Iterator<Integer> verticesIter = vertices.iterator();
337         while(verticesIter.hasNext()) {
338             int vertex = verticesIter.next();

```

```

339         newGraph.addVertex(vertex);
340     }
341     Iterator<Edge> edgesIter = edgesIterator();
342     while(edgesIter.hasNext()) {
343         Edge edge = edgesIter.next();
344         newGraph.addEdge(edge);
345     }
346     return newGraph;
347 }
348
349     public boolean isTreeHamilton() {
350         int[] stopnie = new int[verticesNumber()];
351     Iterator<Edge> drzewoIter = edgesIterator();
352     while(drzewoIter.hasNext()) {
353         Edge para = drzewoIter.next();
354         stopnie[para.getX()]++;
355         stopnie[para.getY()]++;
356     }
357     boolean hamiltonPath = true;
358     for(int i=0; i<verticesNumber(); i++) {
359         if(stopnie[i]>2) {
360             hamiltonPath = false;
361             break;
362         }
363     }
364     return hamiltonPath;
365 }
366
367     public static void main(String[] args) {
368         try {
369             Graph graph = new Graph(5, 0.9, 1.0, 1.0, false);
370             System.out.println(graph);
371         }
372         catch(Exception e) {
373             e.printStackTrace();
374         }
375     }
376 }
377
378
379     class Solution {
380         public String status;
381         public Double value;
382         public Double time;
383     }
384
385
386     public class Adjustment {
387         public Adjustment() {

```

```

388     }
389
390     static Graph graph = new Graph();
391     static String varname = "";
392     static LinkedHashMap<String, String> options;
393     static LinkedHashMap<String, LinkedHashSet<Solution>>
394         statsSolutionStatus;
395     static Random randomizer = null;
396
397     static void initStateSolutionStatus() {
398         statsSolutionStatus = new LinkedHashMap<String,
399             LinkedHashSet<Solution>>();
400         statsSolutionStatus.put("Bounded", new LinkedHashSet<Solution>());
401         statsSolutionStatus.put("Error", new LinkedHashSet<Solution>());
402         statsSolutionStatus.put("Bounded", new LinkedHashSet<Solution>());
403         statsSolutionStatus.put("Feasible", new LinkedHashSet<Solution>());
404         statsSolutionStatus.put("Infeasible", new LinkedHashSet<Solution>());
405         statsSolutionStatus.put("InfeasibleOrUnbounded",
406             new LinkedHashSet<Solution>());
407         statsSolutionStatus.put("Optimal", new LinkedHashSet<Solution>());
408         statsSolutionStatus.put("Unbounded", new LinkedHashSet<Solution>());
409         statsSolutionStatus.put("Unknown", new LinkedHashSet<Solution>());
410     }
411
412     static void createSampleGraph() throws Exception {
413         graph.addEdge(0, 1, 4.0);
414         graph.addEdge(0, 2, 1.0);
415         graph.addEdge(0, 3, 4.0);
416         graph.addEdge(1, 3, 5.0);
417         graph.addEdge(2, 3, 3.0);
418         graph.addEdge(2, 4, 7.0);
419         graph.addEdge(3, 4, 8.0);
420     }
421
422     static void randGraph(int newSize, double density) throws Exception {
423         double uBound = Double.parseDouble(options.get("ubound"));
424         double lBound = Double.parseDouble(options.get("lbound"));
425         boolean isInteger = false;
426         if(options.get("valtype").compareTo("integer")==0) {
427             isInteger = true;
428         }
429         graph.randomGraph(newSize, density, lBound, uBound, isInteger);
430     }
431
432     static void readGraph(String path) {
433         try {
434             File file = new File(path);
435             if(!file.exists()) {
436                 throw new Exception("File \\" + path +

```

```

437             "\" does not exist");
438         }
439         BufferedReader stream =
440             new BufferedReader(new FileReader(path));
441         String row;
442         row = stream.readLine();
443         int size = Integer.parseInt(row);
444         graph = new Graph();
445         int rowcnt = 0;
446         while((row = stream.readLine()) != null) {
447             String[] numbers = row.split(" ");
448             if(numbers.length < size) {
449                 throw new Exception("File error");
450             }
451             for(int i=0; i<size; i++) {
452                 double value = Double.parseDouble(numbers[i]);
453                 if(!graph.containsEdge(rowcnt,i) &&
454                     value<Global.inf) {
455                     graph.addEdge(rowcnt,i,value);
456                 }
457             }
458             rowcnt++;
459         }
460         stream.close();
461         graph.sortVerticesLexOrder();
462         graph.sortEdgesLexOrder();
463     }
464     catch(Exception e) {
465         e.printStackTrace();
466     }
467 }
468
469     static void saveGraph(String path) {
470         try {
471             File file = new File(path);
472             BufferedWriter stream =
473                 new BufferedWriter(new FileWriter(path));
474             int size = graph.verticesNumber();
475             stream.write(Integer.toString(size) + "\n");
476             for(int i=0; i<size; i++) {
477                 for(int j=0; j<size; j++) {
478                     if(graph.containsEdge(i,j)) {
479                         stream.write(Double.toString(
480                             graph.getEdgeValue(i,j)) + " ");
481                     }
482                     else {
483                         stream.write(Double.toString(Global.inf) + " ");
484                     }
485                 }
486             }
487         }

```

```

486         stream.write("\n");
487     }
488     stream.close();
489   }
490   catch(Exception e) {
491     e.printStackTrace();
492   }
493 }
494
495 static void printGraph() {
496   System.out.println(graph.toString());
497 }
498
499 static void populateByRow(IloMPModeler model,
500   LinkedHashMap<String, IloNumVar> variables, ArrayList<IloRange>
501   range, boolean adjustment) throws Exception {
502   double lb = 0.0;
503   double ub = 0.0;
504   IloNumVarType type;
505   IloLinearNumExpr obj = model.linearNumExpr();
506   double M = 0.0;
507   int size = graph.verticesNumber();
508   for(int i=0; i<size; i++) {
509     for(int j=0; j<size; j++) {
510       if(graph.containsEdge(i,j)) {
511         M += Math.abs(graph.getEdgeValue(i,j));
512       }
513     }
514   }
515   for(int i=0; i<size; i++) {
516     for(int j=0; j<size; j++) {
517       if(graph.containsEdge(i,j)) {
518         type = IloNumVarType.Bool;
519         lb = 0.0;
520         ub = 1.0;
521         varname = "x_" + i + "_" + j;
522         variables.put(varname, model.numVar(lb, ub,
523           type, varname));
524         if(!adjustment) {
525           obj.addTerm(variables.get(varname),
526             graph.getEdgeValue(i,j));
527         }
528       }
529     }
530   }
531   for(int k=1; k<size; k++) {
532     for(int i=0; i<size; i++) {
533       for(int j=0; j<size; j++) {
534         if(graph.containsEdge(i,j)) {

```

```

535 lb = 0.0;
536 ub = Double.MAX_VALUE;
537 type = IloNumVarType.Float;
538 varname = "f_" + i + "_" + j + "_" + k;
539 variables.put(varname, model.numVar(lb, ub, type, varname));
540 }
541 }
542 }
543 }
544 for(int k=0; k<2; k++) {
545     for(int i=0; i<size; i++) {
546         for(int j=0; j<size; j++) {
547             if(graph.containsEdge(i,j)) {
548                 type = IloNumVarType.Float;
549                 lb = 0.0;
550                 ub = Double.MAX_VALUE;
551                 if(k==0) varname = "u_";
552                 if(k==1) varname = "l_";
553                 varname += i + "_" + j;
554                 variables.put(varname, model.numVar(lb, ub, type, varname));
555                 if(adjustment) {
556                     obj.addTerm(variables.get(varname), 1.0);
557                 }
558             }
559         }
560     }
561 }
562 for(int k=0; k<2; k++) {
563     for(int i=0; i<size; i++) {
564         for(int j=0; j<size; j++) {
565             if(graph.containsEdge(i,j)) {
566                 type = IloNumVarType.Float;
567                 lb = 0.0;
568                 ub = Double.MAX_VALUE;
569                 if(k==0) varname = "zu_";
570                 if(k==1) varname = "zl_";
571                 varname += i + "_" + j;
572                 variables.put(varname, model.numVar(lb, ub, type, varname));
573             }
574         }
575     }
576 }
577 for(int i=0; i<size; i++) {
578     for(int k=1; k<size; k++) {
579         type = IloNumVarType.Float;
580         lb = -Double.MAX_VALUE;
581         ub = Double.MAX_VALUE;
582         varname = "v_" + i + "_" + k;
583         variables.put(varname, model.numVar(lb, ub, type, varname));

```

```

584         }
585     }
586     for(int k=1; k<size; k++) {
587         for(int i=0; i<size; i++) {
588             for(int j=0; j<size; j++) {
589                 if(graph.containsEdge(i,j)) {
590                     lb = 0.0;
591                     ub = Double.MAX_VALUE;
592                     type = IloNumVarType.Float;
593                     varname = "w_" + i + "_" + j + "_" + k;
594                     variables.put(varname,model.numVar(lb,ub,type,varname));
595                 }
596             }
597         }
598     }
599     type = IloNumVarType.Float;
600     lb = -Double.MAX_VALUE;
601     ub = Double.MAX_VALUE;
602     varname = "l";
603     variables.put(varname, model.numVar(lb, ub, type, varname));
604     model.addMinimize(obj);
605     int rngcount = 0;
606     IloNumExpr expr = model.linearNumExpr();
607     for(int i=0; i<size; i++) {
608         for(int j=0; j<size; j++) {
609             if(graph.containsEdge(i,j)) {
610                 expr = model.sum(expr,
611                                   variables.get("x_" + i + "_" + j));
612             }
613         }
614     }
615     range.add(model.addEq(expr, size-1, "c" + rngcount));
616     rngcount++;
617     for(int k=1; k<size; k++) {
618         expr = model.linearNumExpr();
619         for(int j=0; j<size; j++) {
620             if(graph.containsEdge(j,0)) {
621                 expr = model.sum(expr,
622                                   variables.get("f_" + j + "_" + 0 + "_" + k));
623             }
624         }
625         for(int j=0; j<size; j++) {
626             if(graph.containsEdge(0,j)) {
627                 expr = model.diff(expr,
628                                   variables.get("f_" + 0 + "_" + j + "_" + k));
629             }
630         }
631     range.add(model.addEq(expr, -1.0, "c" + rngcount));
632     rngcount++;

```

```

633         }
634         for(int k=1; k<size; k++) {
635             expr = model.linearNumExpr();
636             for(int j=0; j<size; j++) {
637                 if(graph.containsEdge(j,k)) {
638                     expr = model.sum(expr,
639                         variables.get("f_" + j + "_" + k + "_" + k));
640                 }
641                 if(graph.containsEdge(k,j)) {
642                     expr = model.diff(expr,
643                         variables.get("f_" + k + "_" + j + "_" + k));
644                 }
645             }
646             range.add(model.addEq(expr, 1.0, "c" + rngcount));
647             rngcount++;
648         }
649         for(int k=1; k<size; k++) {
650             for(int i=1; i<size; i++) {
651                 if(i==k) {
652                     continue;
653                 }
654                 else {
655                     expr = model.linearNumExpr();
656                     for(int j=0; j<size; j++) {
657                         if(graph.containsEdge(i,j)) {
658                             expr = model.sum(expr, model.diff(
659                                 variables.get("f_" + j + "_" +
660                                     i + "_" + k), variables.get(
661                                     "f_" + i + "_" + j + "_" + k)));
662                         }
663                         range.add(model.addEq(expr, 0.0, "c" + rngcount));
664                         rngcount++;
665                     }
666                 }
667             }
668         }
669         for(int k=1; k<size; k++) {
670             for(int i=0; i<size; i++) {
671                 for(int j=0; j<size; j++) {
672                     expr = model.linearNumExpr();
673                     if(graph.containsEdge(i,j)) {
674                         expr = model.diff(variables.get("x_" + i + "_" + j),
675                             variables.get("f_" + i + "_" + j + "_" + k));
676                     range.add(model.addGe(expr,0.0,"c"+rngcount));
677                     rngcount++;
678                 }
679             }
680         }
681     }

```

```

682     for(int i=0; i<size; i++) {
683         expr = model.linearNumExpr();
684         for(int j=0; j<size; j++) {
685             if(graph.containsEdge(i,j)) {
686                 expr = model.sum(expr, model.sum(variables.get(
687                     "x_" + i + "_" + j), variables.get("x_" + j + "_" + i)));
688             }
689         }
690         range.add(model.addLe(expr, 2.0, "c" + rngcount));
691         rngcount++;
692     }
693     for(int k=1; k<size; k++) {
694         for(int i=0; i<size; i++) {
695             for(int j=0; j<size; j++) {
696                 expr = model.linearNumExpr();
697                 if(graph.containsEdge(i,j)) {
698                     expr = model.sum(expr, model.diff(
699                         model.diff(variables.get("v_" + j + "_" + k),
700                             variables.get("v_" + i + "_" + k)),
701                             variables.get("w_" + i + "_" + j + "_" + k)));
702             range.add(model.addLe(expr, 0, "c" + rngcount));
703             rngcount++;
704         }
705     }
706 }
707 }
708 for(int i=0; i<size; i++) {
709     for(int j=0; j<size; j++) {
710         expr = model.linearNumExpr();
711         double rh = 0.0;
712         if(i < j && graph.containsEdge(i,j)) {
713             for(int k=1; k<size; k++) {
714                 expr = model.sum(expr, variables.get("w_"
715                     + i + "_" + j + "_" + k));
716             }
717             expr = model.sum(expr, variables.get("l"));
718             expr = model.diff(expr, variables.get("u_"
719                     + i + "_" + j));
720             expr = model.sum(expr, variables.get("l_"
721                     + i + "_" + j));
722             rh = graph.getEdgeValue(i,j);
723             range.add(model.addLe(expr, rh, "c" + rngcount));
724             rngcount++;
725         }
726     }
727 }
728 for(int i=0; i<size; i++) {
729     for(int j=0; j<size; j++) {
730         expr = model.linearNumExpr();

```

```

731     double rh = 0.0;
732     if(i<j && graph.containsEdge(i,j)) {
733         for(int k=1; k<size; k++) {
734             expr = model.sum(expr, variables.get("v_"
735                               + j + "_" + i + "_" + k));
736         }
737         expr = model.sum(expr, variables.get("l"));
738         expr = model.diff(expr, variables.get("u_"
739                           + j + "_" + i));
740         expr = model.sum(expr, variables.get("l_"
741                           + j + "_" + i));
742         rh = graph.getEdgeValue(j,i);
743         range.add(model.addLe(expr,rh,"c"+rngcount));
744         rngcount++;
745     }
746     }
747 }
748 expr = model.linearNumExpr();
749 for(int k=1; k<size; k++) {
750     expr = model.sum(expr, model.diff(variables.get("v_"
751                               + k+"_" + k), variables.get("v_" + 0 + "_" + k)));
752 }
753 expr = model.sum(expr, model.prod(variables.get("l"),size-1));
754 for(int i=0; i<size; i++) {
755     for(int j=0; j<size; j++) {
756         double c = 0.0;
757         if(graph.containsEdge(i,j)) {
758             c = graph.getEdgeValue(i,j);
759             expr = model.diff(expr, model.prod(variables.get(
760                               "zu_" + i + "_" + j), 1.0));
761             expr = model.sum(expr, model.prod(variables.get(
762                               "zl_" + i + "_" + j), 1.0));
763             expr = model.diff(expr, model.prod(c,
764                               variables.get("x_" + i + "_" + j)));
765         }
766     }
767 }
768 range.add(model.addEq(expr, 0, "c" + rngcount));
769 rngcount++;
770 for(int i=0; i<size; i++) {
771     for(int j=0; j<size; j++) {
772         expr = model.linearNumExpr();
773         if(graph.containsEdge(i,j)) {
774             expr = model.diff(variables.get("zu_" + i +
775                               "_" + j), model.prod(variables.get(
776                               "x_" + i + "_" + j), M));
777             range.add(model.addLe(expr, 0.0, "c" + rngcount));
778             rngcount++;
779     }

```

```

780         }
781     }
782     for(int i=0; i<size; i++) {
783         for(int j=0; j<size; j++) {
784             expr = model.linearNumExpr();
785             if(graph.containsEdge(i,j)) {
786                 expr = model.diff(variables.get("zu_" + i +
787                     "_" + j), variables.get("u_" + i + "_" + j));
788                 range.add(model.addLe(expr, 0.0, "c" + rngcount));
789                 rngcount++;
790             }
791         }
792     }
793     for(int i=0; i<size; i++) {
794         for(int j=0; j<size; j++) {
795             expr = model.linearNumExpr();
796             if(graph.containsEdge(i,j)) {
797                 expr = model.sum(model.diff(variables.get("u_"
798                     + i + "_" + j),variables.get("zu_" + i + "_" + j)),
799                     model.prod(variables.get("x_" + i + "_" + j),M));
800                 range.add(model.addLe(expr, M, "c" + rngcount));
801                 rngcount++;
802             }
803         }
804     }
805     for(int i=0; i<size; i++) {
806         for(int j=0; j<size; j++) {
807             expr = model.linearNumExpr();
808             if(graph.containsEdge(i,j)) {
809                 expr = model.diff(variables.get("zl_" + i + "_" + j),
810                     model.prod(variables.get("x_" + i + "_" + j),M));
811                 range.add(model.addLe(expr,0.0,"c"+rngcount));
812                 rngcount++;
813             }
814         }
815     }
816     for(int i=0; i<size; i++) {
817         for(int j=0; j<size; j++) {
818             expr = model.linearNumExpr();
819             if(graph.containsEdge(i,j)) {
820                 expr = model.diff(variables.get("zl_" + i + "_" + j),
821                     variables.get("l_" + i + "_" + j));
822                 range.add(model.addLe(expr,0.0,"c"+rngcount));
823                 rngcount++;
824             }
825         }
826     }
827     for(int i=0; i<size; i++) {
828         for(int j=0; j<size; j++) {

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829         expr = model.linearNumExpr();
830         if(graph.containsEdge(i,j)) {
831             expr = model.sum(model.diff(variables.get("l_"
832                             +i+"_"+j), variables.get("zl_" +i+"_"+j)),
833                             model.prod(variables.get("x_" +i+"_"+j), M));
834             range.add(model.addLe(expr, M, "c" + rngcount));
835             rngcount++;
836         }
837     }
838 }
839 }
840
841 public static void readTest(String path) {
842     try {
843         File plik = new File(path);
844         if(!plik.exists()) {
845             throw new Exception("File \\" + path + "\ does
846 not exist");
847         }
848         BufferedReader strumien = new BufferedReader(new
849             FileReader(path));
850         String wiersz;
851         options = new LinkedHashMap<String, String>();
852         while((wiersz = strumien.readLine()) != null) {
853             String[] entry = wiersz.split("\s=\s");
854             if(entry.length != 2) {
855                 throw new Exception("Error in: " + wiersz);
856             }
857             options.put(entry[0], entry[1]);
858         }
859         strumien.close();
860     }
861     catch(Exception e) {
862         e.printStackTrace();
863     }
864 }
865
866 public static Solution solve(IloCplex cplex, LinkedHashMap<String,
867     IloNumVar> variables, ArrayList<IloRange> range)
868 throws Exception {
869     IloRange[] rng = new IloRange[range.size()];
870     for(int i=0; i<range.size(); i++) {
871         rng[i] = range.get(i);
872     }
873     IloNumVar[] var = new IloNumVar[variables.size()];
874     String[] varnames = new String[variables.size()];
875     Iterator<IloNumVar> iter = variables.values().iterator();
876     int i=0;
877     while(iter.hasNext()) {

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878     IloNumVar numvar = iter.next();
879     var[i] = numvar;
880     varnames[i] = numvar.getName();
881     i++;
882 }
883 if(options.get("solve").compareTo("true")==0) {
884     Solution solution = new Solution();
885     long timerStart = java.lang.System.currentTimeMillis();
886     if(cplex.solve()) {
887         double[] x = cplex.getValues(var);
888         int ncols = cplex.getNumCols();
889         for (int j = 0; j < ncols; ++j) {
890             if(x[j]!=0.0 && j<var.length) {
891                 cplex.output().println(" " + varnames[j] +
892                     "\t=\t" + x[j]);
893             }
894         }
895     }
896     solution.status = cplex.getStatus().toString();
897     solution.time = (double)
898         (java.lang.System.currentTimeMillis()
899          - timerStart) / 1000.0;
900     if(cplex.getStatus() == IloCplex.Status.Feasible ||
901        cplex.getStatus() == IloCplex.Status.Optimal) {
902         solution.value = cplex.getObjectiveValue();
903     }
904     statsSolutionStatus.get(cplex.getStatus().toString())
905         .add(solution);
906     return solution;
907 }
908 return null;
909 }

910 public static void main(String[] args) {
911     try {
912         if(args.length != 1) {
913             throw new Exception("Incorrect number of arguments");
914         }
915         initStatsSolutionStatus();
916         readTest(args[0]);
917         long seed = 0;
918         int maxTests = Integer.parseInt(options.get("testcount"));
919         if(options.get("source").compareTo("random")==0) {
920             if(options.get("seed").compareTo("time")==0) {
921                 seed = java.lang.System.currentTimeMillis();
922             }
923         }
924     else {
925         seed = Integer.parseInt(options.get("seed"));
926     }

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    }
    Global.initRandomizer(seed);
    BufferedWriter stream = null;
    if(options.get("name").compareTo("none")!=0) {
        stream = new BufferedWriter(new
            FileWriter(options.get("name") + ".log"));
        stream.write("") + maxTests + "\n\n");
    }
    for(int countTests=0; countTests<maxTests; countTests++) {
        if(options.get("source").compareTo("random")==0) {
            randGraph(Integer.parseInt(options.get("size")),
                Double.parseDouble(options.get("density")));
        }
        if(options.get("source").compareTo("file")==0) {
            readGraph(options.get("inpath"));
        }
        printGraph();
        LinkedHashMap<String, IloNumVar> variables = new
            LinkedHashMap<String, IloNumVar>();
        ArrayList<IloRange> range = new ArrayList<IloRange>();
        IloCplex cplex = new IloCplex();
        populateByRow(cplex, variables, range, true);
        Solution solution = solve(cplex, variables, range);
        if(options.get("name").compareTo("none")!=0) {
            String path = "";
            path += options.get("name")+"_"+countTests+".lp";
            cplex.exportModel(path);
            stream.write("number\t= " + countTests + "\n");
            stream.write("status\t= " + cplex.getStatus() + "\n");
            stream.write("value\t= " + solution.value + "\n");
            stream.write("time\t= " + solution.time + "\n");
            stream.write("\n");
            stream.flush();
            saveGraph(options.get("name")+"_"+countTests
                +"_graph");
        }
        cplex.end();
    }
    if(stream!=null) {
        Iterator<Solution> iter =
            statsSolutionStatus.get("Optimal").iterator();
        double optimalSolutions =
            statsSolutionStatus.get("Optimal").size();
        double timeavg = 0.0;
        double timemax = 0.0;
        double timemin = 999999999.0;
        while(iter.hasNext()) {
            Solution solution = iter.next();
            timeavg += solution.time / optimalSolutions;

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976         if(timemax < solution.time) {
977             timemax = solution.time;
978         }
979         if(timemin > solution.time) {
980             timemin = solution.time;
981         }
982     }
983     stream.write("\n");
984     stream.write("optimal\t= " +(int)optimalSolutions+"\n");
985     stream.write("timemin\t= " + timemin + "\n");
986     stream.write("timeavg\t= " + timeavg + "\n");
987     stream.write("timemax\t= " + timemax + "\n");
988     stream.close();
989 }
990 }
991 catch(IloCplex.UnknownObjectException iloException) {
992     System.out.println(iloException.getObject().toString());
993 }
994 catch(Exception e) {
995     e.printStackTrace();
996 }
997 }
998 }
```









