

Exercise Session 3

8.5.2014.

- Multi commodity flows (MCFP)
- Programming Exercise

MCFP: Given directed graph $G=(V,A)$, $c:A \rightarrow \mathbb{R}^+$, $u:A \rightarrow \mathbb{R}^+$
pair $(s_i, t_i)_{1 \leq i \leq k}$ with $d_i \in \mathbb{R}^+$
 $C \subset V^2$

solution: flows $x_{i,a}$ with minimal cost $\sum_{i=1}^k \sum_{a \in A} c(a) \cdot x_{i,a}$
flow value on arc a for commodity i

$$\bullet \sum_{a \in \mathcal{F}^-(v)} x_{i,a} - \sum_{a \in \mathcal{F}^+(v)} x_{i,a} = \begin{cases} -d_i & v=s_i \\ +d_i & v=t_i \\ 0 & \text{else} \end{cases} \quad \forall i \in [k]$$

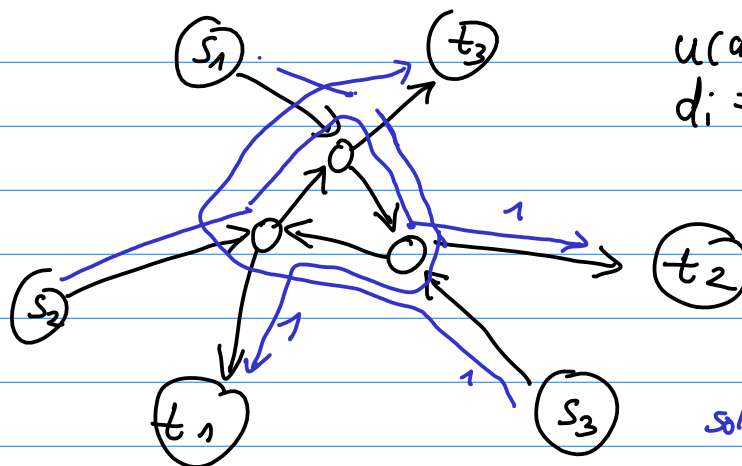
flow is conservative

$$\bullet \sum_{i \in [k]} x_{i,a} \leq u(a) \quad \forall a \in A$$

flow on arcs is below capacity

$$x_{i,a} \geq 0 \quad \forall i \in [k], a \in A$$

example:

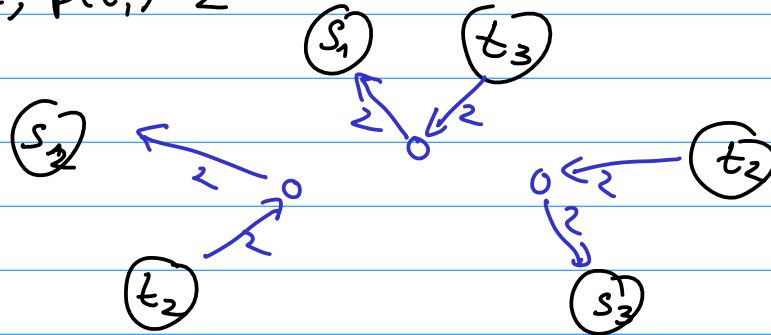


$$u(a) = 2 \quad \forall a \in A$$
$$d_i = 1, \quad c(a) = 1 \quad \forall a \in A$$

solution to MCFP

difference to b-Transshipment:

$$\beta(s_i) = -2, \beta(t_i) = 2$$



- can not use residual network as in b-Transshipment
- demands are coupled in pairs
- directed and undirected case not similar

LP formulation: arc based

$$\min \sum_{i \in [K]} \sum_{a \in A} c(a) \cdot x_{i,a}$$

$$\bullet \sum_{a \in \mathcal{F}^-(v)} x_{i,a} - \sum_{a \in \mathcal{F}^+(v)} x_{i,a} = \begin{cases} -d_i & v = s_i \\ +d_i & v = t_i \\ 0 & \text{else} \end{cases} \quad \forall i \in [K], \forall v \in V$$

flow is conservative

$$\bullet \sum_{i \in [K]} x_{i,a} \leq u(a) \quad \forall a \in A$$

flow on arcs is below capacity

$$x_{i,a} \geq 0 \quad \forall i \in [K], a \in A$$

Problems: LP is solvable via simplex, # constr, # variables is polynomial
 \leadsto possibly HUGE

$$\bullet \text{example for real life data set: } |V| = 106 \quad K = 11 \cdot 130 \approx |V|^2 \\ |A| = 904$$

$$\Rightarrow |E| \cdot K \geq 10 \text{ million variables}$$

$$\Rightarrow \geq |V| \cdot K \geq 1 \text{ million constraints} \Rightarrow \text{huge Matrix } A$$

Using Dantzig Wolfe Decomposition:

Path-based LP formulation

→ dualize LP

$$\min \sum_{p \in \mathcal{P}} c_p \cdot x_p$$

$$\sum_{p \in \mathcal{P}: a \in p} x_p \leq u(a) \quad \forall a \in A \quad \rightsquigarrow \gamma_a \leq 0$$

$$\sum_{p \in \mathcal{P}_i} x_p \geq d_i \quad \forall i \in [K] \quad \rightsquigarrow z_i \geq 0$$

$$x_p \geq 0 \quad \forall p \in \mathcal{P} \quad \rightsquigarrow \text{constr. for all } p \in \mathcal{P}$$

dual LP: $\max \sum_{a \in A} u(a) \cdot \gamma_a + \sum_{i \in [K]} d_i \cdot z_i$

$$z_i + \sum_{a \in p} \gamma_a \leq c \quad \forall p \in \mathcal{P}_i, i \in [K]$$

$$\gamma_a \leq 0 \quad \forall a \in A$$

// arc prices

$$z_i \geq 0 \quad \forall i \in [K]$$

// potential for commodities

Idea: expon. many paths $s_i - t_i$, only a couple of them are interesting

→ start with a few paths variables

→ solve primal LP

→ use dual LP to find variable (path) with neg. red. costs

→ add path to primal LP and repeat

reduced cost of path p : $\bar{c} := c - \underbrace{c_B^T \cdot B^{-1}}_{\text{dual variable}} \cdot A$

using: red. cost is negative \Leftrightarrow corresponding dual constr. is violated

\Rightarrow looking for s_i - t_i path p s.t:

$$z_i + \sum_{a \in p} y_a > c_p \Leftrightarrow z_i > \sum_{a \in p} c(a) - y_a =: \sum_{a \in p} c'(a) = c'_p$$

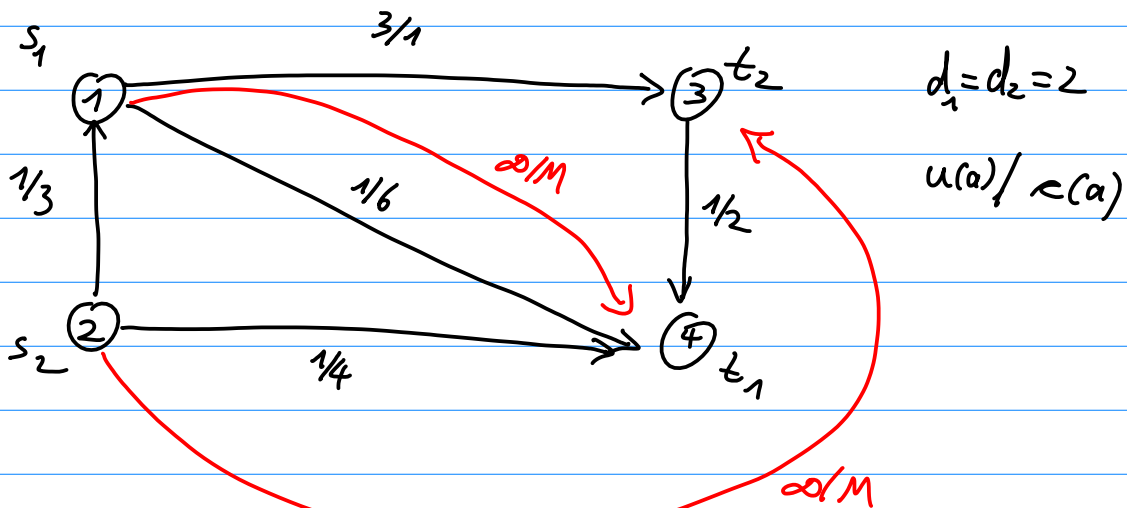
find s_i - t_i path in G with edge cost $< z_i$
 \rightarrow can be done by min s_i - t_i path computation

$c' \geq 0$ because $c(a) \geq 0$ and $y_a \leq 0$

\rightarrow Dijkstra algorithm can be used

- procedure:
- ① setup LP with paths such that it contains a solution
 add s_i - t_i arcs with $c(a) = M, u(a) = 0$
 - ② solve LP and get dual sol. z_i, y_a
 - ③ for $i=1$ to k
 - find shortest s_i - t_i path p in G with $c'_a = c_a - y_a$
 - add path variable to LP if $c'_p < z_i$
 - ④ goto ② until no add. paths were found

example:



choice for M : big enough $M = \sum_{a \in A} c(a) \cdot \sum_{i=1}^k d_i$

Programming exercise:

- read file format network/demands
- build graph
- build primal path based LP
- solve LP via CPLEX and get dual sol.
- search paths with length $\leq z_i$; \leadsto Dijkstra...
- repeat until no further path found

path calculation:

- Bellman: $O(n^4) / O(n^3 \log n)$

- Floydwarshall: $O(n^3)$

- Dijkstra: $O(n^2) \cdot k$

Listen

$O(m \log n) \cdot k$

Fibb. Heap

$O((m+n) \log n) \cdot k$

AVL tree

CPLEX:

```
package examples;

import ilog.concert.*;
import ilog.cplex.*;

/**
 * Example of the usage of the CPLEX API with Java.
 * The code shows how to build an LP for a multi commodity
 * flow problem with four given paths.
 *
 * @author Torsten Gellert
 */

public class SolvingLpWithCplex{

    /** Contant for the cost of artifical arcs for this particular e;
    public static final double M = 1000.0;
    /** Hide constructor, since this class is an simple example. */
    private SolvingLpWithCplex() {
    }
}
```

```

/**
 * Creates a path based LP formulation of the multi commodity flow problem
 * for the instance shown in the exercise session.
 * Solves and prints all variables and additional available information.
 *
 * @param args The argument list . Not used by the method.
 */

public static void main( String[] args ) {
    try {
        IloCplex cplex = new IloCplex();

        IloNumVar[][] var = new IloNumVar[1][];
        IloRange[][] rng = new IloRange[1][];

        populateByColumn(cplex, var, rng);

        if ( cplex.solve() ) {
            double[] x      = cplex.getValues(var[0]);
            double[] dj     = cplex.getReducedCosts(var[0]);
            double[] pi     = cplex.getDuals(rng[0]);
            double[] slack  = cplex.getSlacks(rng[0]);

            cplex.output().println("Solution status = " + cplex.getStatus());
            cplex.output().println("Solution value = " + cplex.getObjValue());

            int nvars = x.length;
            for (int j = 0; j < nvars; ++j) {
                cplex.output().println(var[0][j].getName() +
                    " = " + x[j] +
                    "\tReduced cost = " + dj[j]);
            }

            int ncons = slack.length;
            for (int i = 0; i < ncons; ++i) {
                cplex.output().println(rng[0][i].getName() +
                    ":\tSlack = " + slack[i] +
                    "\tPi = " + pi[i]);
            }
        }
        cplex.end();
    } catch (IloException exc) {
        System.err.println("Concert exception '" + exc + "' caught");
    }
}

/**
 * Builds all variables and constraints(ranges). These are
 * stored in a arrays and these array have to be placed in the
 * first position of the given two dimentional arrays.
 *
 * @param model The Cplex model for handling with LPs.
 * @param var   The array to store all created variables.
 * @param rng   The array to store all ranges/constraints.
 * @throws IloException if there occurs an error while build the LP model.
 */

```

```

static void populateByColumn(IloMPModeler model,
                             IloNumVar[][] var,
                             IloRange[][] rng) throws IloException {
    IloObjective obj = model.addMinimize();

    //create all constraints
    rng[0] = new IloRange[7];
    //rhs for arcs
    rng[0][0] = model.addRange(-Double.MAX_VALUE, 3.0, "arc_1_3");
    rng[0][1] = model.addRange(-Double.MAX_VALUE, 1.0, "arc_1_4");
    rng[0][2] = model.addRange(-Double.MAX_VALUE, 1.0, "arc_2_1");
    rng[0][3] = model.addRange(-Double.MAX_VALUE, 1.0, "arc_2_4");
    rng[0][4] = model.addRange(-Double.MAX_VALUE, 1.0, "arc_3_4");
    //rhs for commodities
    rng[0][5] = model.addRange(2, Double.MAX_VALUE, "commodity_1");
    rng[0][6] = model.addRange(2, Double.MAX_VALUE, "commodity_2");

    //add variables
    IloRange c1 = rng[0][5];
    IloRange c2 = rng[0][6];

    IloRange arc_1_3 = rng[0][0];
    IloRange arc_1_4 = rng[0][1];
    IloRange arc_2_1 = rng[0][2];
    IloRange arc_2_4 = rng[0][3];
    IloRange arc_3_4 = rng[0][4];

    var[0] = new IloNumVar[4];
    var[0][0] = model.numVar(model.column(obj, M).and(
        model.column(c1, 1.0)),
        0.0, Double.MAX_VALUE, "p_1_1");
    var[0][1] = model.numVar(model.column(obj, M).and(
        model.column(c2, 1.0)),
        0.0, Double.MAX_VALUE, "p_2_1");
    var[0][2] = model.numVar(model.column(obj, 4).and(
        model.column(arc_1_3, 1.0).and(
        model.column(arc_3_4, 1.0).and(
        model.column(c1, 1.0))),
        0.0, Double.MAX_VALUE, "p_1_2");
    var[0][3] = model.numVar(model.column(obj, 4.0).and(
        model.column(arc_2_1, 1.0).and(
        model.column(arc_1_3, 1.0).and(
        model.column(c2, 1.0))),
        0.0, Double.MAX_VALUE, "p_2_2");
}
}

```

```
/*
```

```
Output:
```

```
LP Presolve eliminated 4 rows and 0 columns.  
Reduced LP has 3 rows, 4 columns, and 6 nonzeros.  
Presolve time = 0.00 sec.
```

```
Iteration log . . .
```

```
Iteration: 1 Dual objective = 8.000000
```

```
Solution status = Optimal
```

```
Solution value = 1012.0
```

```
p_1_1 = 1.0 Reduced cost = 0.0
```

```
p_2_1 = 0.0 Reduced cost = 0.0
```

```
p_1_2 = 1.0 Reduced cost = 0.0
```

```
p_2_2 = 2.0 Reduced cost = 0.0
```

```
arc_1_3: Slack = 0.0 Pi = -996.0
```

```
arc_1_4: Slack = 1.0 Pi = 0.0
```

```
arc_2_1: Slack = 1.0 Pi = 0.0
```

```
arc_2_4: Slack = 1.0 Pi = 0.0
```

```
arc_3_4: Slack = 1.0 Pi = 0.0
```

```
commodity_1: Slack = 0.0 Pi = 1000.0
```

```
commodity_2: Slack = 0.0 Pi = 1000.0
```

```
*/
```

→ see web page for code and running instructions

Also: checkstyle compliant code!