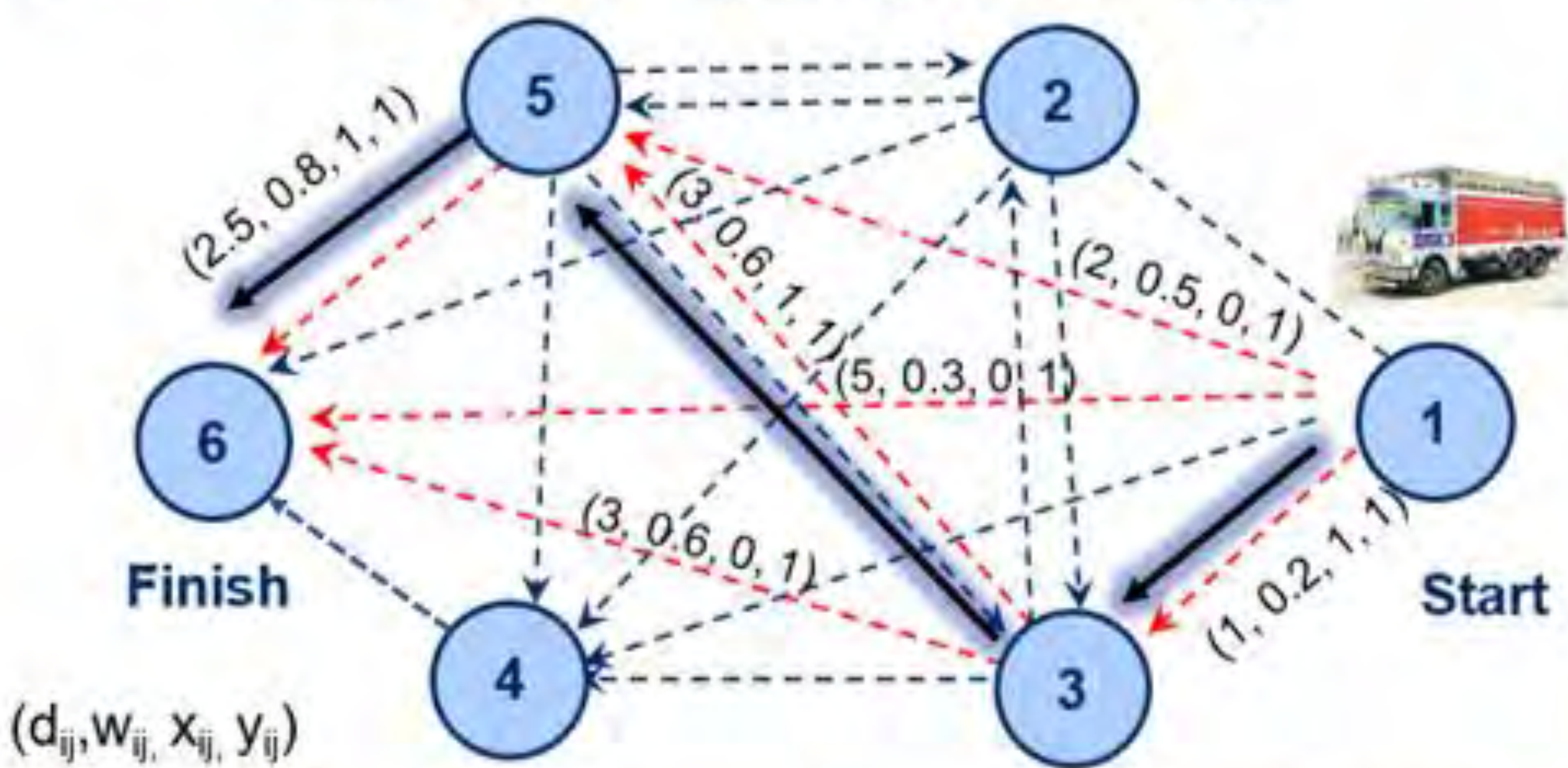


## Abstract

The Triples formulation is a compact formulation of multicommodity network flow problems that provides a different representation of flow than the traditional widely used node-arc and arc-path approaches. We have applied the Triples concept to the backhaul profit maximization problem (BPMP) and the fixed charge network flow problem (FCNF). Experimental analysis shows that the Triples formulations of BPMP and FCNF are more compact, stronger, and faster than the traditional node-arc models for most problem instances.

## Backhaul Profit Maximization Problem

← Demand (optional pick up and delivery)



← Optimal vehicle route ← Demand satisfied

$d_{ij}$ : distance from  $i$  to  $j$ ;  $w_{ij}$ : demand from  $i$  to  $j$ ;  $x_{ij} = 1$  if the vehicle travels on arc  $(i, j)$ ;  $y_{ij} = 1$  if demand from  $i$  to  $j$  is satisfied  
Objective: maximize profit  
Decision variables:  $x_{ij}, y_{ij}$   
Constraints: time and weight.

## Flow: Node-Arc vs. Triples

### • Node-Arc Model

–  $z_{kl,ij} = 1$  if demand from  $k$  to  $l$  is shipped on arc  $(i, j)$

– Example solution (6 demands selected):

$$\bullet z_{13,13} = 1, z_{15,13} = z_{15,35} = 1, z_{16,13} = z_{16,35} = z_{16,56} = 1$$

$$\bullet z_{35,35} = 1, z_{36,35} = z_{36,56} = 1, z_{56,56} = 1$$

–  $\theta_{ij}$  = total flow (tons) on arc  $(i, j)$

$$\bullet \theta_{13} = w_{13} z_{13,13} + w_{15} z_{15,13} + w_{16} z_{16,13} = 1 \text{ ton}$$

$$\bullet \theta_{35} = w_{15} z_{15,35} + w_{16} z_{16,35} + w_{35} z_{35,35} + w_{36} z_{36,35} = 2 \text{ tons}$$

$$\bullet \theta_{56} = w_{16} z_{16,56} + w_{36} z_{36,56} + w_{56} z_{56,56} = 1.7 \text{ tons}$$

–  $O(n^4)$  binary variables and  $O(n^3)$  constraints

### • Triples Model

–  $u_{ij}^k$  = flow (tons) from  $i$  to  $j$  diverted through  $k$ :  
routed on arc  $(i, k)$  and a path from  $k$  to  $j$

– Example solution diverts 3 of 6 selected demands

$$\bullet u_{15}^3 = w_{15}, u_{16}^3 = w_{16}, \text{ and } u_{36}^5 = w_{16} + w_{36}$$

$$\bullet \theta_{13} = w_{13} + u_{15}^3 + u_{16}^3 = 1 \text{ ton}$$

$$\bullet \theta_{35} = w_{35} + u_{15}^3 + u_{36}^5 = 2 \text{ tons}$$

$$\bullet \theta_{56} = w_{56} + u_{36}^5 = 1.7 \text{ tons}$$

–  $O(n^3)$  binary variables and  $O(n^2)$  constraints

## BPMP: Node-Arc vs. Triples Models

Maximize  $p[\sum_{(k,l) \in E} d_{kl} w_{kl} y_{kl}] - c \sum_{(i,j) \in E} \theta_{ij} d_{ij} - cv \sum_{(i,j) \in E} d_{ij} x_{ij}$

Subject to:

$$\sum_{j \in V} z_{kl,kj} = y_{kl} \quad (k, l) \in A$$

$$\sum_{i \in V} z_{kl,il} = y_{kl} \quad (k, l) \in A$$

$$\sum_{i \in V, (i,a) \in A} z_{kl,ia} = \sum_{j \in V, (a,j) \in A} z_{kl,aj} \quad (k, l) \in A, a \in V \setminus \{k, l\}$$

$$\sum_{(k,l) \in A} z_{kl,ij} \leq (n^2 - 3n + 3) x_{ij} \quad (i, j) \in A$$

$$\sum_{(1,j) \in A} x_{1j} = 1$$

$$\sum_{(i,n) \in A} x_{in} = 1$$

$$\sum_{i \in V \setminus \{k,n\}} x_{ik} = \sum_{j \in V \setminus \{1,k\}} x_{kj} \quad k \in V \setminus \{1, n\}$$

$$\sum_{(i,j) \in A} d_{ij} x_{ij} \leq D$$

$$\theta_{ij} = \sum_{(k,l) \in A} w_{kl} z_{kl,ij} \quad (i, j) \in A$$

$$\theta_{ij} = w_{ij} y_{ij} + \sum_{(i,k) \in T} u_{ik}^j + \sum_{(k,j) \in T} u_{kj}^i - \sum_{(i,j,k) \in T} u_{ij}^k, \quad (i, j) \in A$$

$$\theta_{ij} \leq Q, \quad (i, j) \in A$$

$$\theta_{ij} \leq Q x_{ij}, \quad (i, j) \in A$$

$$s_i - s_j + (n) x_{ij} \leq n - 1 \quad (i, j) \in A$$

$$x_{ij} = 0, 1 \quad (i, j) \in A$$

$$y_{kl} = 0, 1 \quad (k, l) \in A$$

$$u_{ij}^k \geq 0, \quad (i, j, k) \in T$$

$$z_{kl,ij} = 0, 1 \quad (k, l) \in A, (i, j) \in A$$

Black: common for both models

Blue: unique to Node-arc model

Red: unique to Triples model

## Restricted Triples Heuristic

Algorithm: Restricted Triples Heuristic

Input: A BPMP instance

Output: A feasible triples solution

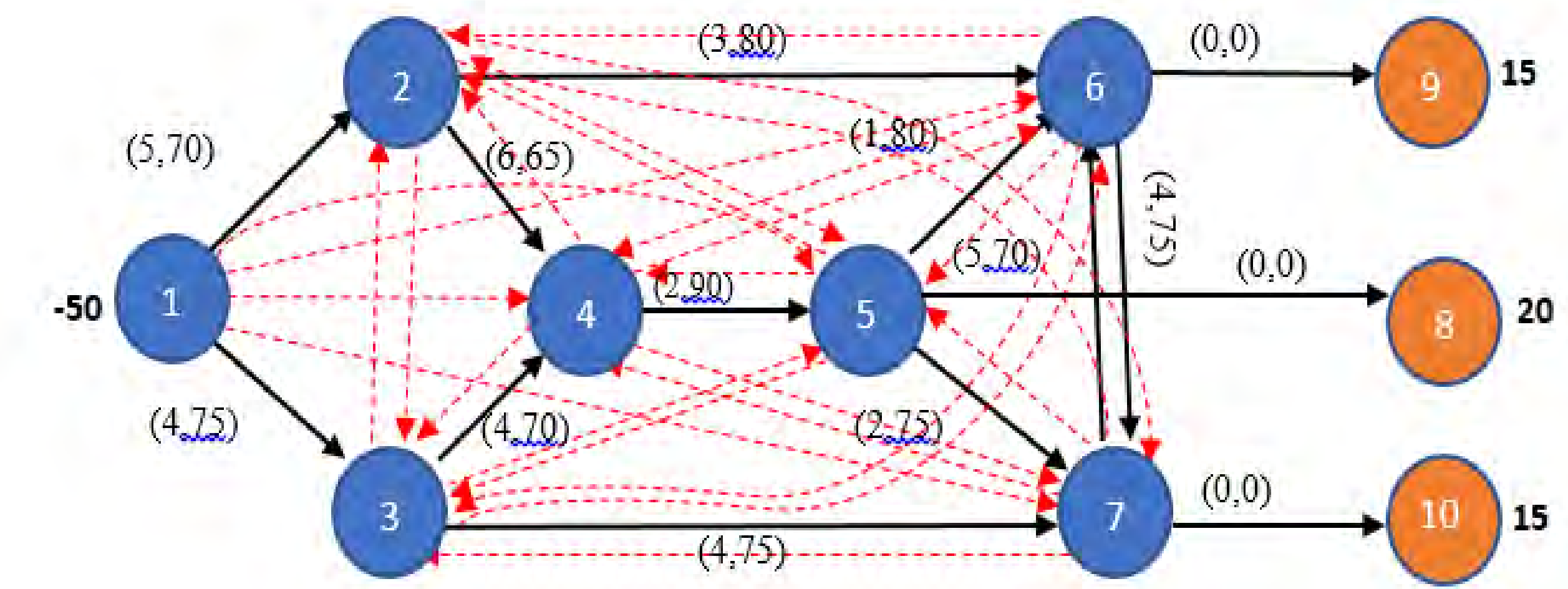
- (1)  $\hat{T} \leftarrow \emptyset$
- (2) For  $\forall (i, j, k) \in T$
- (3)  $\rho_{ij}^k \leftarrow p d_{ij} w_{ij} - c(d_{ik} + d_{kj})(v + w_{ij})$
- (4) If  $w_{ij} + w_{ik} \leq Q$  Then  $\rho_{ij}^k \leftarrow \rho_{ij}^k + (p - c) d_{ik} w_{ik}$
- (5) If  $w_{ij} + w_{kj} \leq Q$  Then  $\rho_{ij}^k \leftarrow \rho_{ij}^k + (p - c) d_{kj} w_{kj}$
- (6) If  $\rho_{ij}^k \geq 0$  Then  $\hat{T} \leftarrow \hat{T} \cup \{(i, j, k)\}$
- (7) End For
- (8) Solve enhanced triples formulation with  $\hat{T}$  (i.e., fix  $u_{ij}^k = 0$  for  $(i, j, k) \in T \setminus \hat{T}$ )
- (9)  $A_x \leftarrow \{(i, j) \in A : x_{ij} = 1\}$
- (10)  $R_y \leftarrow \{(i, j) \in R : y_{ij} = 1\}$
- (11) Solve enhanced triples formulation with  $\hat{T}$  subject to
 
$$x_{ij} = 1 \quad \forall (i, j) \in A_x,$$

$$y_{ij} = 1 \quad \forall (i, j) \in R_y.$$
- (12) Return Triples solution  $(x, y, u, s)$

## Empirical CPU Comparison

| Average CPU (seconds) |          |         |          |
|-----------------------|----------|---------|----------|
| Nodes                 | Node Arc | Triples | Speed Up |
| 10                    | 14.4     | 2.4     | 1.5      |
| 20                    | 13,644   | 20      | 758      |
| 30                    | N/A      | 480     | NA       |

## Fixed Charge Network Flow Problem



Legend:  $(c_{ij}, f_{ij})$ , default values: (1000,1000) if missing.

### Base

$$\text{Min } \sum_{(i,j) \in A} c_{ij} x_{ij} + \sum_{(i,j) \in A} f_{ij} y_{ij}$$

$$\sum_{(i,k) \in A} x_{ik} - \sum_{(k,j) \in A} x_{kj} = b_k, \quad k \in V$$

$$x_{ij} \leq (\sum_{d \in D} b_d) y_{ij}, \quad (i, j) \in A$$

$$x_{ij} \geq 0, \quad (i, j) \in A$$

$$y_{ij} \in \{0, 1\}$$

### MCE

$$\text{Min } \sum_{(i,j) \in A} x_{ij} c_{ij} + \sum_{(i,j) \in A} f_{ij} y_{ij}$$

$$\sum_{(i,k) \in A} w_{ik}^d - \sum_{(k,j) \in A} w_{kj}^d = \begin{cases} b_d & k = d \in D \\ 0 & k \in V \setminus \{1, d \in D, d \neq k\} \\ -b_d & k = 1, d \in D \end{cases}$$

$$x_{ij} \leq (\sum_{d \in D} b_d) y_{ij}, \quad (i, j) \in A$$

$$x_{ij} \geq \sum_{d \in D} w_{ij}^d, \quad (i, j) \in A$$

$$w_{ij}^d \leq b_d y_{ij}, \quad (i, j) \in A, d \in D$$

$$w_{ij}^d \geq 0, \quad (i, j) \in A, d \in D$$

$$y_{ij} \in \{0, 1\}$$

### Triples

$$\text{Minimize } \sum_{(i,j) \in A} x_{ij} c_{ij} + \sum_{(i,j) \in A} f_{ij} y_{ij}$$

$$x_{ij} = d_{ij} + \sum_{(i,k) \in T} u_{ik}^j + \sum_{(k,j) \in T} u_{kj}^i - \sum_{(i,j,k) \in T} u_{ij}^k \quad (i, j) : i \in V, j \in V \setminus \{i\}$$

$$T: \{(i, j, k) : i \in V, j \in V \setminus \{i\}, k \in V \setminus \{i, j\}\}$$

$$x_{ij} \leq s_{ij} * y_{ij}, \quad (i, j) : i \in V, j \in V \setminus \{i\}$$

$$u_{ij}^k \geq 0 \quad \forall (i, j, k) \in T$$

$$y_{ij} \in \{0, 1\} \quad \forall (i, j) \in A$$

## Experimental Comparison

| Node    | MCE vs. Base |                |                   |                       | Triples vs. Base |               |                   |                       |
|---------|--------------|----------------|-------------------|-----------------------|------------------|---------------|-------------------|-----------------------|
|         | CPU speed up | ticks speed up | Real time speedup | LP lower improve ment | CPU speed up     | ticks speedup | Real time speedup | LP lower improve ment |
| 20      | 6.45         | 1.34           | 1.91              | 36.8%                 | 6.41             | 1.34          | 1.94              | 36.8%                 |
| 30      | 6.15         | 1.12           | 1.56              | 26.6%                 | 5.85             | 1.12          | 1.51              | 26.6%                 |
| 100     | 7.02         | 1.44           | 1.62              | 89.1%                 | 6.79             | 1.44          | 1.59              | 89.1%                 |
| 500     | 1.41         | 16.29          | 0.12              | 320.1%                | 1.38             | 16.29         | 0.11              | 320.1%                |
| Average | 5.26         | 5.05           | 1.30              | 118.1%                | 5.10             | 5.05          | 1.29              | 118.1%                |

## Conclusions and Future Work

- The Triples model is more compact and faster to solve than node-arc model for BPMP.
- The Triples model is at least as good as the traditional advanced MCE formulation for FCNF.
- In the future, we will explore singleton-selection strategies to further accelerate our solution approach.