

Capacitated Arc Routing Problem with Profits

Dominique Feillet

University of Avignon

(moving soon to Ecole des Mines de Saint-Etienne)

Co-authors: C. Archetti, A. Hertz, M.G. Speranza

How column generation can prove that France should have beat Italy

- The model
 - Maximize French score
 - Subject to: Italian score \leq French score -1
- No feasible solution found so far
- Duality can help
 - French staff should price out the referee as much as the Italian staff

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A simple approach

- Take your standard basic Branch and Price solver for VRP type problems
 - send an email to the good person if needed
- Change some lines of code

Topic: how to be (almost) as simple but efficient
(towards efficient generic solvers)

Outline

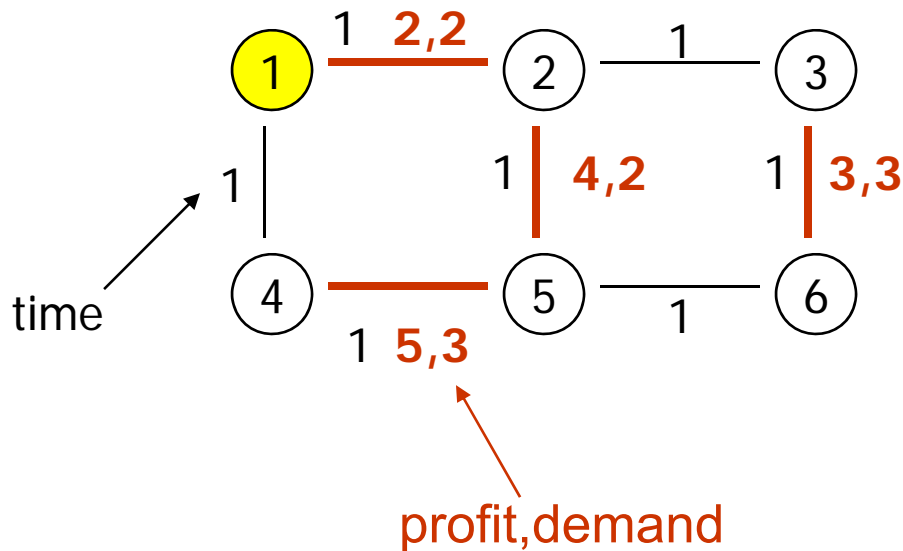
- The problem
- Model and basic algorithm
- Some robust acceleration methods
 - Initially developed for the VRPTW
- Some difficulties relative to the CARPP

The Capacitated Arc routing Problem with Profits

- Given
 - a graph $G=(V,E)$
 - a set Q of edges with specified demands and profits
 - a fleet of capacitated vehicles based on a depot
 - edge durations
 - a travel time limit
- Find a set of vehicle routes such that
 - the total profit is maximized
 - vehicle capacities and travel time limit are satisfied

The Capacitated Arc routing Problem with Profits

■ Illustration



Number of vehicles: 3
 Time limit: 5
 Capacity: 10

Example of solution:
 route 1=2=5=4-1

Profit: 11

Model and basic algorithm

- Model

$$\text{maximize } \sum_{r_k \in \Omega} p_k x_k$$

$$\sum_{r_k \in \Omega} a_{ek} x_k \leq 1 \quad (e \in Q),$$

$$\sum_{r_k \in \Omega} x_k \leq m,$$

$$x_k \in \{0, 1\} \quad (r_k \in \Omega).$$

Model and basic algorithm

- Master Problem

$$\begin{aligned} & \text{maximize } \sum_{r_k \in \Omega} p_k x_k \\ & \sum_{r_k \in \Omega} a_{ek} x_k \leq 1 \quad (e \in Q), \\ & \sum_{r_k \in \Omega} x_k \leq m, \\ & 0 \leq x_k \quad (r_k \in \Omega). \end{aligned}$$

- A route has a positive reduced cost if:

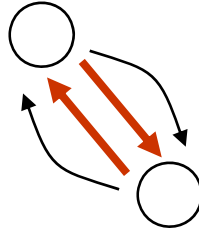
$$p_k - \sum_{e \in Q} a_{ek} \lambda_e - \mu > 0$$

Model and basic algorithm

- Subproblem:

- Shortest Path Problem with Resource Constraints

- modeling of the edges in Q:



- Service not performed, cost 0
- Service performed, cost $p_e - \lambda_e$

- Resources:

- time
- load
- one binary resource for every edge in Q (a service is performed at most once)

Some robust acceleration procedures

- Heuristic solution of the subproblem
 - Limited Discrepancy Search
 - Instead of:
 - problem dependant heuristic algorithm (tabu search...)
 - scaling algorithm (consider a subgraph)
- Use of the current primal master solution in the subproblem
 - MP provides the subproblem with routes having a zero reduced cost value

Limited Discrepancy Search (LDS)

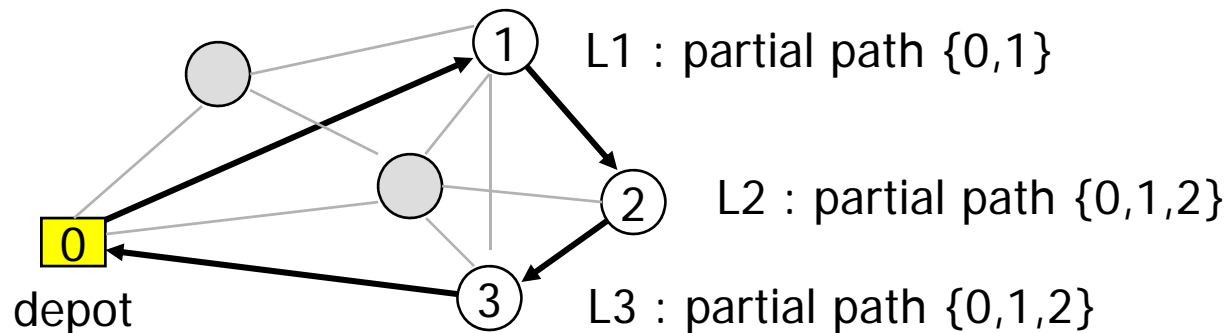
- Principle:
 - define good and bad extensions (arcs)
 - limit the number of bad extensions during the process of computing a route (discrepancies)
 - if no route of negative reduced cost is found, increase the value of the limit
- Implementation for the CARPP
 - count a discrepancy when traversing any edge unless it is in Q and
 - either you serve it
 - or it has already been served

Limited Discrepancy Search (LDS)

- Interest
 - simple
 - almost problem-independent
 - efficient
- Remark:
 - the rule for counting discrepancies is here adapted to the structure of the problem
 - a standard rule would be to count a discrepancy for all outgoing arcs but the two with the most positive reduced cost

Use of the current master primal solution

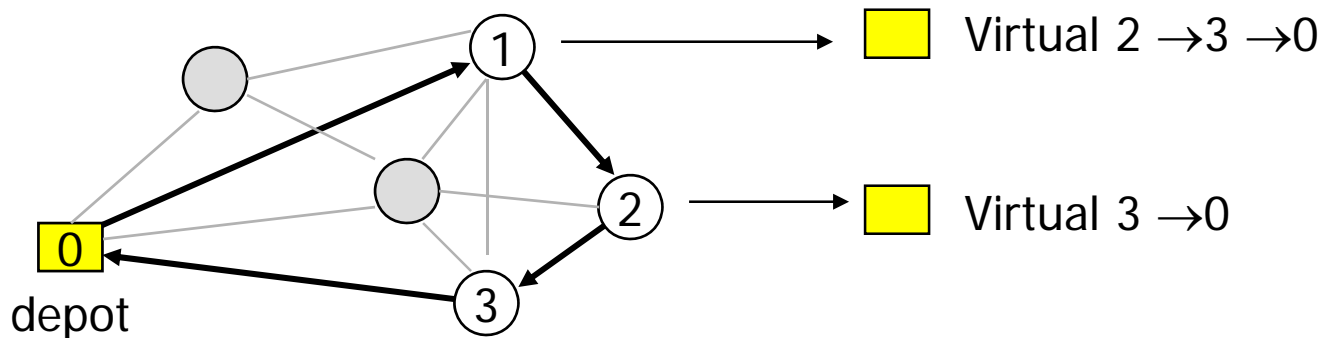
- Label Loading
 - consider the set of routes with a zero reduced cost value in the current solution of the Master Problem
 - generate labels corresponding to starting parts of these routes



Use of the current master primal solution

- Meta Extensions

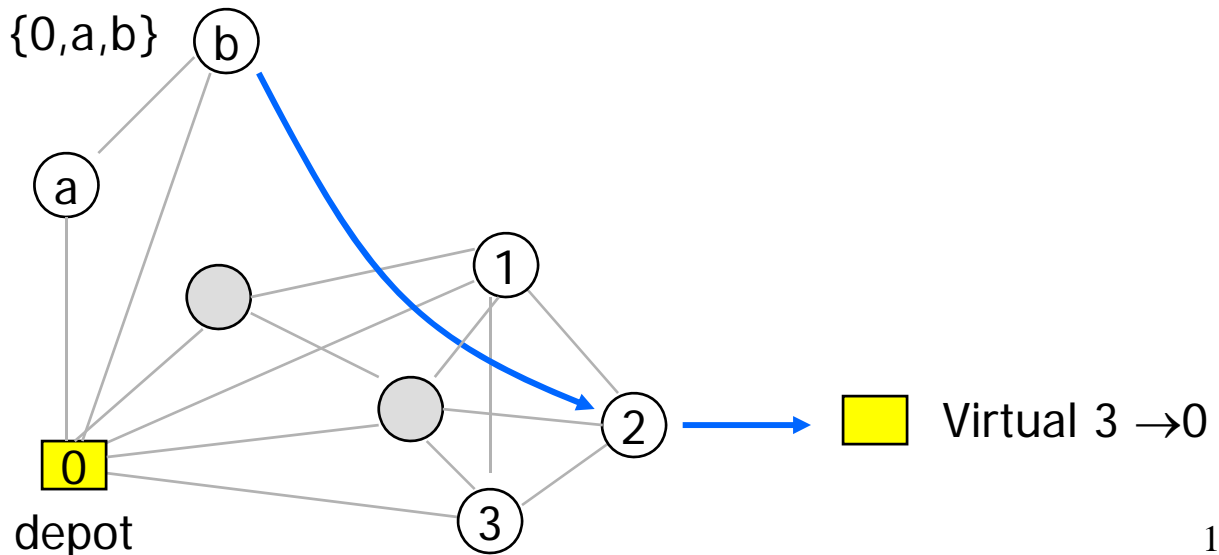
- consider the same set of routes
- generate virtual depots corresponding to ending parts of these routes
 - compute resource windows using resource consumption rules



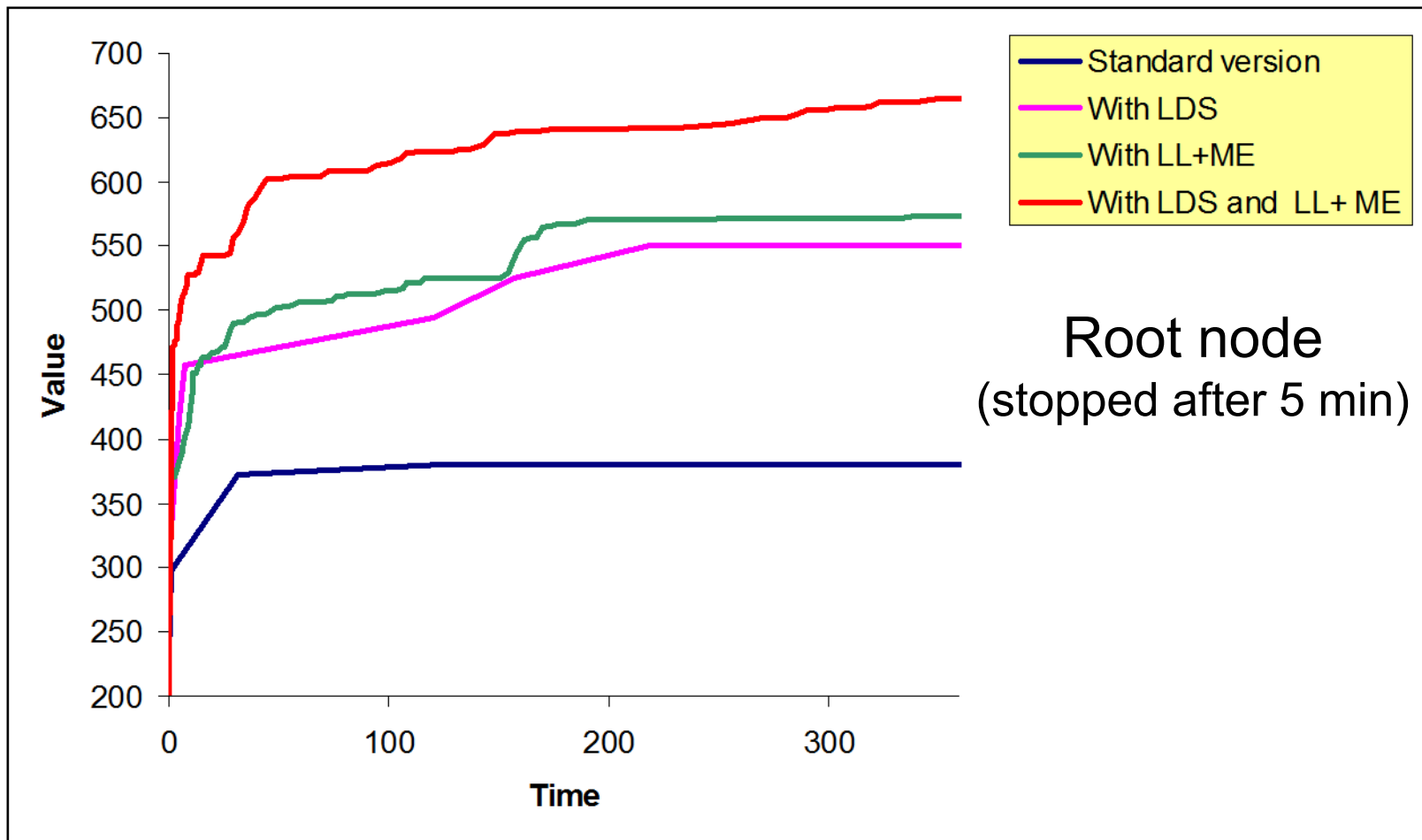
Use of the current master primal solution

- Interest
 - Amounts to a very powerful local search
 - Simple, efficient, generic

L : partial path {0,a,b}



Some robust acceleration procedures

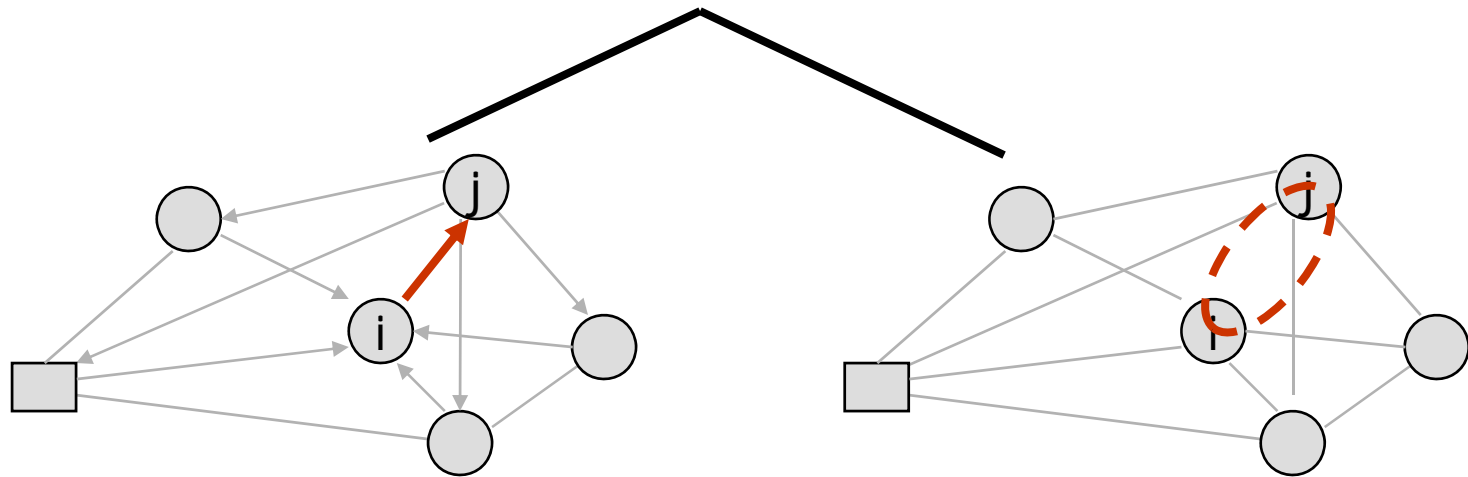


Some difficulties relative to the CARPP

- Presence of long negative cost cycles
 - routes typically are about 20 edges long (in problem that we are able to solve), with vertices traversed several times
 - routes with non-elementary services of edges cannot have a value 1 in the MP
- Rounded capacity cuts are not valid
 - edges are not necessarily served

Some difficulties relative to the CARPP

- Usual branching rule when applying column generation to vehicle routing problems is not valid

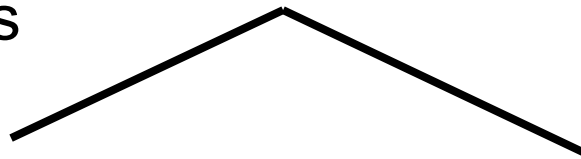


- Valid on condition that every vertex is visited exactly once in optimal solutions, which is not true here

Branching rule

- First level:

- search for a service strictly performed between 0 and 1 times

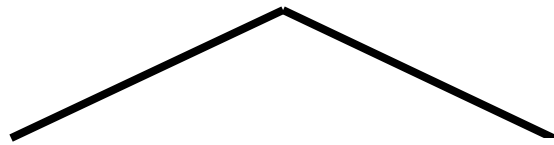


Change \leq to $=$ in the MP
formulation

Remove from Q
(update Ω_1 and the subproblem)

Branching rule

- Second level:
 - search for an arc visited a fractional number of times



$$\sum_{r_k \in \Omega_1} a_{ijk} x_k \leq \lfloor \alpha \rfloor$$

include the new dual
variable in the subproblem

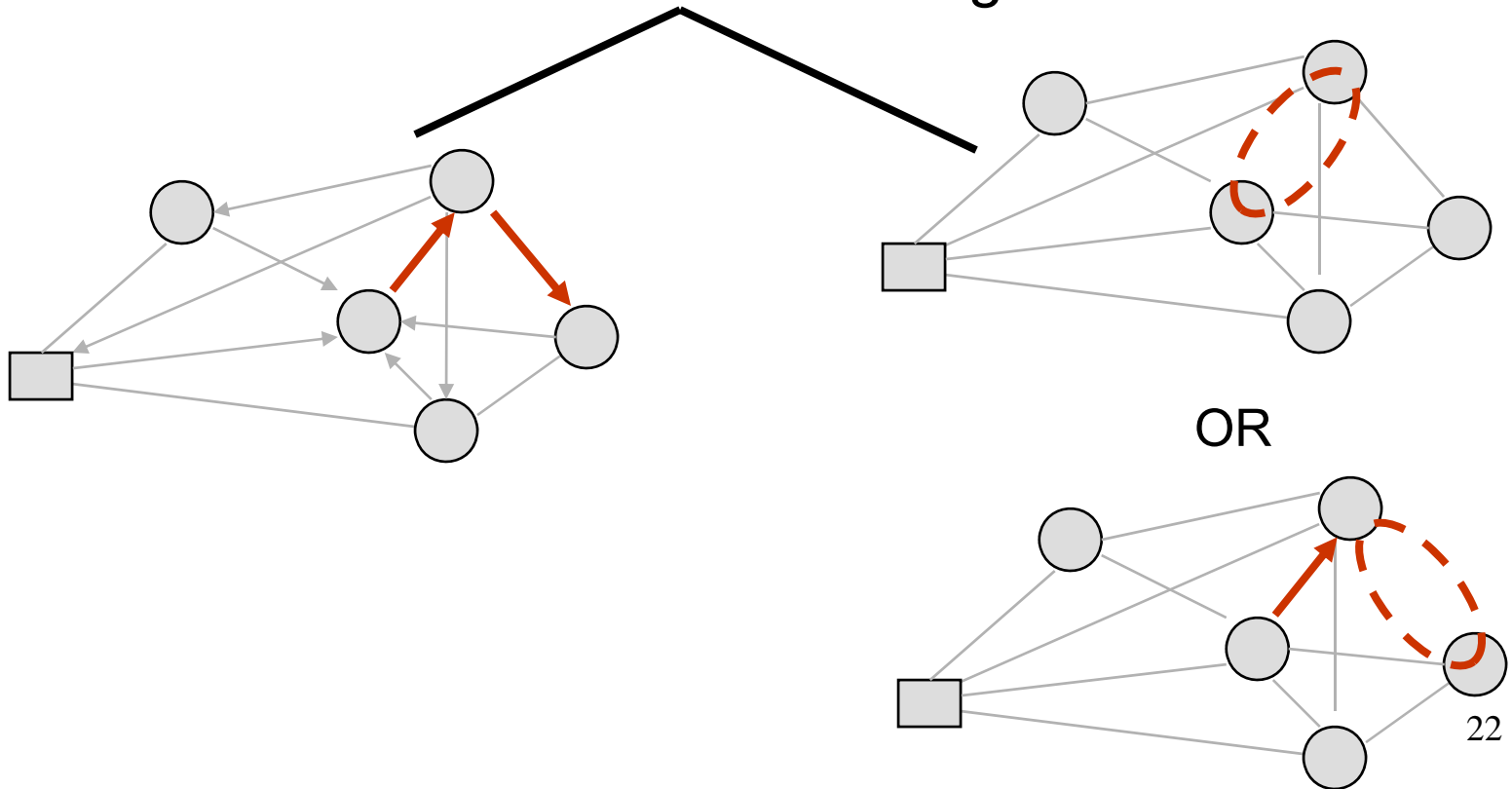
$$\sum_{r_k \in \Omega_1} a_{ijk} x_k \geq \lfloor \alpha \rfloor + 1$$

include the new dual
variable in the subproblem

- note that the time limit constraint becomes essential in case of negative cost cycles with no edge serviced

Branching rule

- Third level: Flow splitting method
 - use when the flow matrix is integer



Conclusions

- Acceleration techniques have proved to be simple to adapt and efficient
- One has to be careful with two common tricks
 - (partial) relaxation of the condition that services are performed at most once
 - management of the branching with removal of variable (MP) and arcs (subproblem)

Questions?

- No questions on the football game allowed

Conclusions

- Perspectives
 - How the acceleration procedures could also tend to find integer solutions during the process?
 - this was often seen during the computational experiments
 - interesting for the use of column generation methods in heuristic fashions

Limited Discrepancy Search (LDS)

