

A Bucket Graph Based Labelling Algorithm for the Resource Constrained Shortest Path Problem with Applications to Vehicle Routing

Ruslan Sadykov^{1,2} **Artur Pessoa**³ **Eduardo Uchoa**³

¹ Inria Bordeaux,
France



² Université Bordeaux,
France



³ Universidade Federal
Fluminense, Brazil



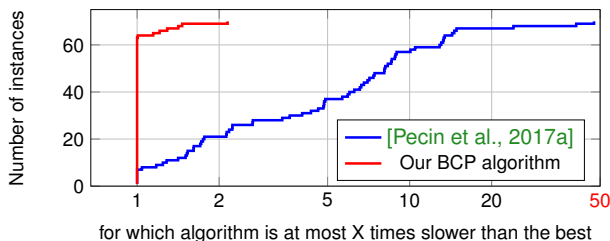
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Computational results for classic VRPTW instances

14 hardest [Solomon, 1987] instances with 100 customers

60 [Gehring and Homberger, 2002] instances with 200 customers

Algorithm	Solved	65 instances solved by both	
		Aver. time (m)	Geom. time (m)
[Pecin et al., 2017a]	65/74	217.8	32.6
Our BCP algorithm	70/74	72.5	8.3



Pecin, D., Contardo, C., Desaulniers, G., and Uchoa, E. (2017).

New enhancements for the exact solution of the vehicle routing problem with time windows.

INFORMS Journal on Computing, 29(3):489–502.

Branch-Cut-and-Price algorithm components contributing to improvement over [Pecin et al., 2017a]

- ▶ **New bucket graph based labelling algorithm** for the Resource Constrained Shortest Path pricing problem
- ▶ Automatic dual price smoothing stabilization [Pessoa et al., 2017]
- ▶ Dynamic *ng*-path relaxation [Roberti and Mingozzi, 2014]



Pessoa, A., Sadykov, R., Uchoa, E., and Vanderbeck, F. (2017).

Automation and combination of linear-programming based stabilization techniques in column generation.

INFORMS Journal on Computing, forthcoming.



Roberti, R. and Mingozzi, A. (2014).

Dynamic *ng*-path relaxation for the delivery man problem.

Transportation Science, 48(3):413–424.

Structure of RCSP instances we want to solve

- ▶ A **complete** directed **graph** $G = (V, A)$.
- ▶ **Unrestricted in sign** reduced **costs** \bar{c}_a on arcs $a \in A$
- ▶ Two global (capacity and time) resources with non-negative **non-integer resource consumption** $d_{a,1}, d_{a,2}$ on arcs $a \in A$
- ▶ Resource consumption bounds $[0, W]$ and $[l_v, u_v]$ on vertices $v \in V$
- ▶ Up to $\approx 500 - 1000$ of (more or less) **local** binary or (small) integer **resources**

We want to

Find a walk from the sources to the sink minimizing the total reduced cost respecting the resource constraints, as well as many other (up to 1000) different near-optimal feasible walks

Literature : “standalone” algorithms for the RCSP

Test instances with an **acyclic sparse graph** with global, but **few resources**, aim to find **one optimal solution**

- ▶ Heavy pre-processing and Lagrangian relaxation
[Dumitrescu and Boland, 2003]
- ▶ Transformation to the shortest path problem
[Zhu and Wilhelm, 2012] or the k -shortest paths problem
[Santos et al., 2007]
- ▶ **Pulse Algorithm** (limited dominance and depth-first search)
[Lozano and Medaglia, 2013]
 - ▶ **the best** “standalone” algorithm
 - ▶ **fails completely** for our hard instances [Pecin, 2014]



Lozano, L. and Medaglia, A. L. (2013).

On an exact method for the constrained shortest path problem.

Computers & Operations Research, 40(1):378 – 384.

Basic labelling algorithm

$\mathcal{L} = \bigcup_{v \in V} \mathcal{L}_v$ — set of non-extended labels

$\mathcal{E} = \bigcup_{v \in V} \mathcal{E}_v$ — set of extended labels

$\mathcal{L} \rightarrow \{(\text{source}, 0, 0, 0, \{\text{source}\})\}$, $\mathcal{E} \leftarrow \emptyset$

while $\mathcal{L} \neq \emptyset$ **do**

 pick a label L in \mathcal{L} , $v^L \neq \text{sink}$

$\mathcal{L} \leftarrow \mathcal{L} \setminus \{L\}$, $\mathcal{E} \leftarrow \mathcal{E} \cup \{L\}$

foreach $v \in V \setminus v^L$ **do**

 extend L to L' along arc (v^L, v)

if L' is feasible and not dominated by a label in $\mathcal{L}_v \cup \mathcal{E}_v$ **then**

$\mathcal{L} \leftarrow \mathcal{L} \cup \{L'\}$

 remove from $\mathcal{L}_v \cup \mathcal{E}_v$ all labels dominated by L'

return a label in $\mathcal{L}_{\text{sink}}$ with the smallest reduced cost

Label-setting if labels are picked in a total order \leq_{lex} such that

L extends to $L' \Rightarrow L \leq_{\text{lex}} L'$, L dominates $L' \Rightarrow L \leq_{\text{lex}} L'$

Otherwise, it is **label-correcting** (for example, cycling over \mathcal{L}_v)

Literature: “embedded” algorithms for the RCSP

All approaches are variants of the **labelling algorithm**

- ▶ **Bi-directional search** [Righini and Salani, 2006]
- ▶ Enumeration of elementary routes using completion bounds from the **ng-path relaxation** [Baldacci et al., 2011]
- ▶ Completion bounds from **dynamic state-space relaxation** of the resources from non-robust cuts [Contardo and Martinelli, 2014]
- ▶ Limited dominance checks by **discretisation** of the resource consumption [Pecin et al., 2017b].



Righini, G. and Salani, M. (2006).

Symmetry helps: Bounded bi-directional dynamic programming for the elementary shortest path problem with resource constraints.

Discrete Optimization, 3(3):255 – 273.



Pecin, D., Pessoa, A., Poggi, M., and Uchoa, E. (2017).

Improved branch-cut-and-price for capacitated vehicle routing.

Mathematical Programming Computation, 9(1):61–100.

Our approach to improve the labelling algorithm

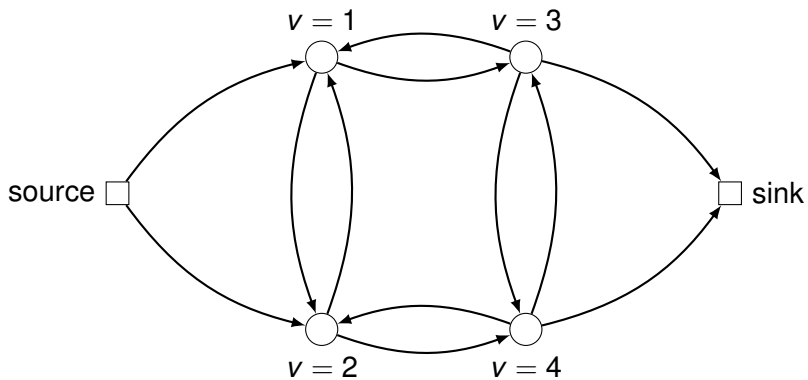
To our knowledge, no (published) attempts to

reduce the number of dominance checks

while keeping the dominance strength

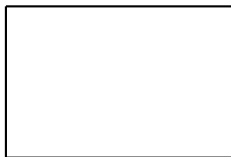
in a labelling algorithm

Original graph

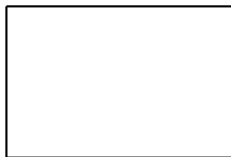


The bucket graph (with two main resources)

$v = 1$



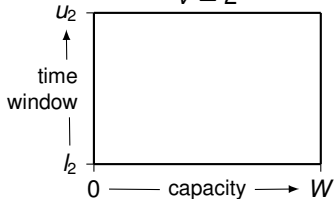
$v = 3$



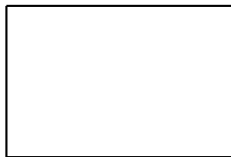
source

sink

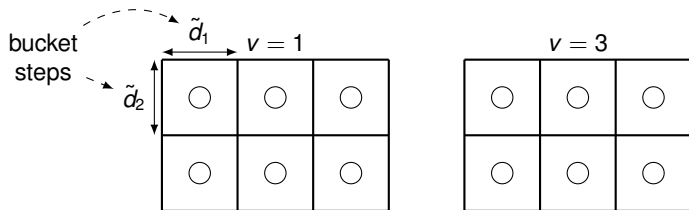
$v = 2$



$v = 4$

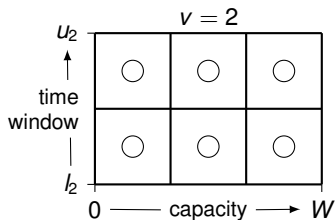


The bucket graph (with two main resources)

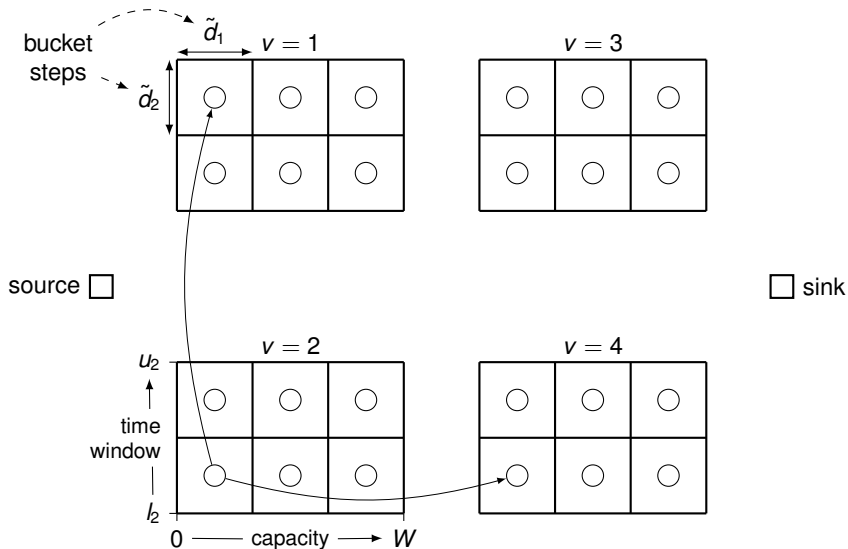


source

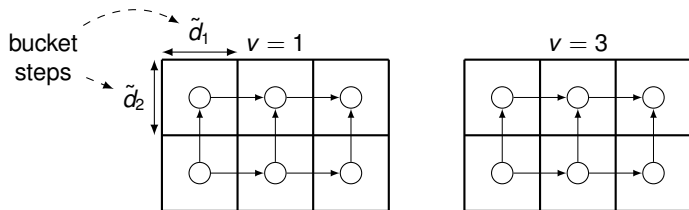
sink



The bucket graph (with two main resources)

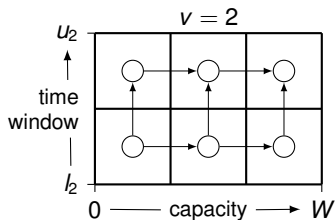


The bucket graph (with two main resources)

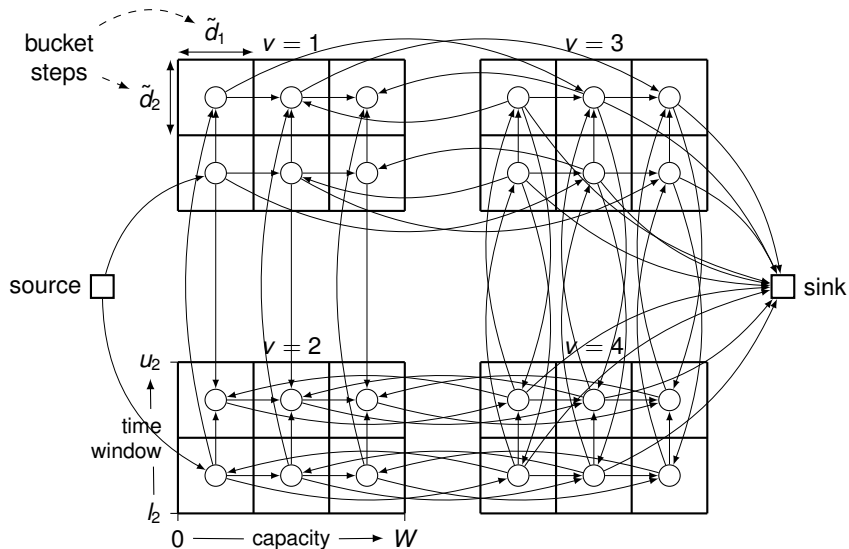


source

sink



The bucket graph (with two main resources)



Extension order of labels

Extend labels according to a **topological order of strongly connected components** in the bucket graph.

Impact of bucket steps

Large enough bucket steps produce the standard **label-correcting** algorithm

- ▶ One bucket per vertex
- ▶ Bucket graph reduces to the original graph
- ▶ One strongly connected component (for our instances)

Small enough bucket steps produce a **label-setting** algorithm

- ▶ Acyclic bucket graph
- ▶ Guarantee that only non-dominated labels are extended

Optimization of dominance checks

Practical observation

Higher dominance probability between labels with similar global resource consumption

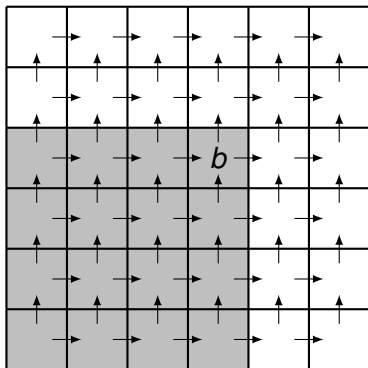
After the label's creation


check dominance with labels in the **same bucket only!**

Before the label's extension

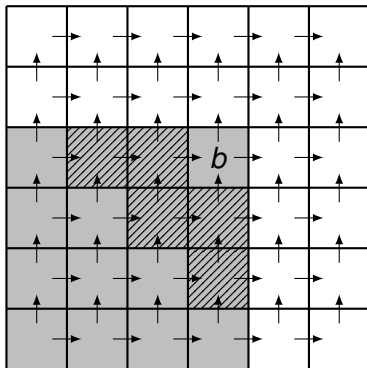
check dominance with labels in **other buckets using bounds**


Using bounds to reduce dominance checks between buckets




\bar{c}_b^{best} — minimum reduced cost of labels in buckets $b' \preceq b$ (area )

Using bounds to reduce dominance checks between buckets



\bar{c}_b^{best} — minimum reduced cost of labels in buckets $b' \preceq b$ (area )

Label L may be dominated in buckets $b' \preceq b$ only if $\bar{c}^L \geq \bar{c}_b^{best}$

(only buckets in area  are tested)

Bi-directional variant

- ▶ Pick a global resource (f.e. capacity) and a **threshold w^***
- ▶ In the forward labelling, keep **only labels \vec{L} with $w^{\vec{L}} \leq w^*$**
- ▶ In the backward labelling, keep **only labels \vec{L} with $w^{\vec{L}} > w^*$**
- ▶ Perform the **concatenation step**: a forward label \vec{L} and a backward label \vec{L} can be concatenated along arc $(v^{\vec{L}}, v^{\vec{L}})$
- ▶ Concatenation is **accelerated using bounds $\bar{c}_{\vec{b}}^{best}$** : if

$$\bar{c}^{\vec{L}} + \bar{c}_{(v^{\vec{L}}, v^{\vec{L}})} + \bar{c}_{\vec{b}}^{best} \geq UB(\bar{c}^*)$$

then we can skip backward buckets $\vec{b}' \preceq \vec{b}$ while searching for a concatenation pair for label \vec{L}

- ▶ **Exploiting symmetry**: if all time windows are the same, the backward labelling is equivalent to the forward, we use only forward buckets and labels in concatenation

Computational impact of buckets steps

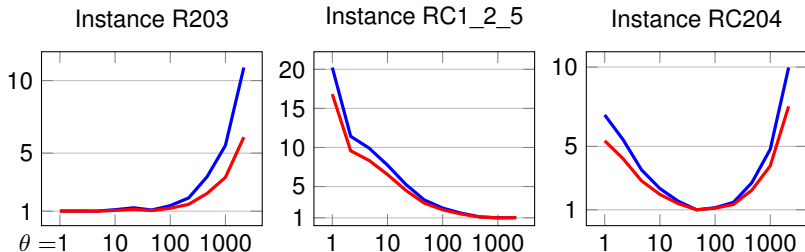
- ▶ Same **VRPTW instances**
- ▶ A full-blown **state-of-the-art column-and-cut generation** at the root (stop when the target lower bound is reached)
- ▶ We test the parameter θ — **the maximum number of buckets per vertex**:

$$\tilde{d}_1 = \frac{W}{\sqrt{\theta}}, \quad \tilde{d}_2 = \frac{U_{depot} - l_{depot}}{\sqrt{\theta}} \quad (\text{two global resources})$$

$$\tilde{d} = \frac{U_{depot} - l_{depot}}{\theta} \quad (\text{one global resource})$$

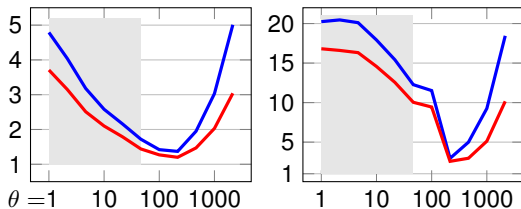
- ▶ $\theta = 1$ — standard label-correcting algorithm

Computational impact of buckets steps



Average

Maximum

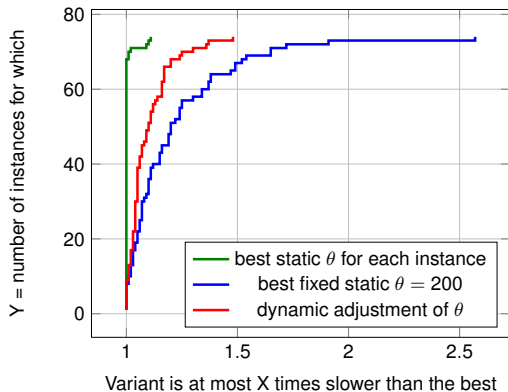


— Pricing time ratio to best θ — Total time ratio to best θ

Dynamic adjustment of bucket steps

- ▶ Start with $\theta = 25$
- ▶ Multiply θ by 2 each time this ratio is above a threshold

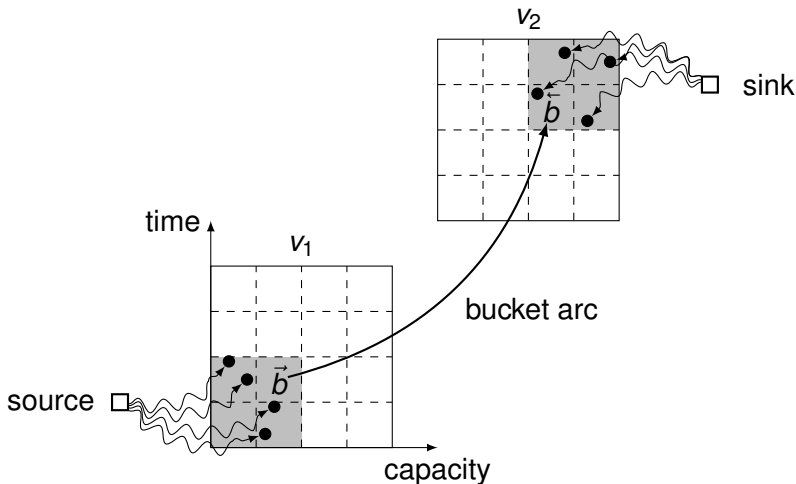
$$\frac{\text{\# of dominance checks inside buckets}}{\text{\# of non-dominated labels}}$$



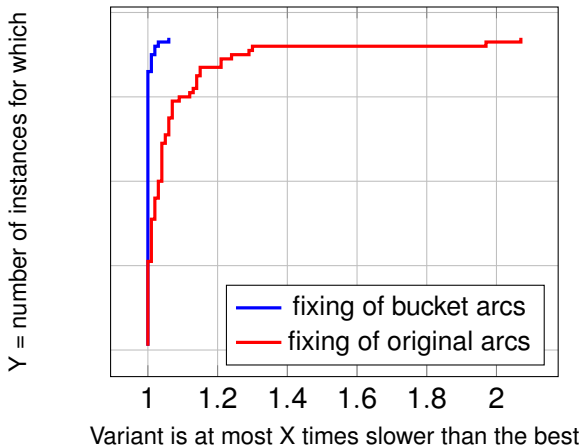
Fixing of bucket arcs by reduced cost

A sufficient condition to fix a bucket arc $(\vec{b}, (v_1, v_2), \vec{b})$

No pair of labels $(\vec{L}, \vec{L}), v^{\vec{L}} = v_1, v^{\vec{L}} = v_2, \vec{b}^{\vec{L}} \preceq \vec{b}, \vec{b}^{\vec{L}} \preceq \vec{b}$, producing a path by concatenation along arc (v_1, v_2) with reduced cost smaller than the current primal-dual gap.



Computational impact of fixing bucket arcs by reduced cost (the root node only)



Computational results for the MDVRP instances

Classic **distance constrained multi-depot** instances by [Cordeau et al., 1997] with **up to 288 customers**.

Algorithm	Solved	10 inst. solved by both	
		Aver. time	Geom. time
[Contardo and Martinelli, 2014]	10/13	269.8	8.4
Our algorithm	22/22	2.5	0.5

One improved BKS (instance “pr10”) over [Vidal et al., 2012]



Contardo, C. and Martinelli, R. (2014).

A new exact algorithm for the multi-depot vehicle routing problem under capacity and route length constraints.

Discrete Optimization, 12:129 – 146.

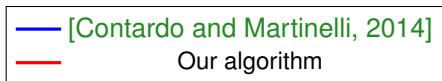
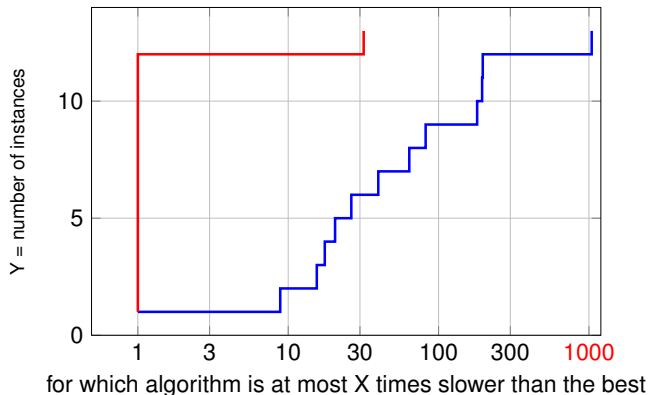


Vidal, T., Crainic, T. G., Gendreau, M., Lahrichi, N., and Rei, W. (2012).

A hybrid genetic algorithm for multidepot and periodic vehicle routing problems.

Operations Research, 60(3):611–624.

Computational results for the MDVRP instances: performance profile



Computational results for other problems

First exact algorithm for these vehicle routing variants

DCVRP Classic distance-constrained CVRP instances
[Christofides et al., 1979]

SDVRP Standard distance-constrained site-dependent instances [Cordeau and Laporte, 2001]

HFVRP “Nightmare” heterogeneous fleet VRP instances (very large capacities) [Duhamel et al., 2011]

Class	Solved	Largest solved n	Smallest unsolved n	Geomean time	Improv. BKS
DCVRP	6/7	200	120	16m44s	0/7
SDVRP	7/10	216	240	11m26s	4/10
HFVRP	56/96	186	107	23m07s	43/96



Christofides, N., Mingozzi, A., and Toth, P. (1979).

Combinatorial Optimization, chapter “The vehicle routing problem”, p. 315–338.

Wiley, Chichester.

Conclusions

- ▶ **No universally best algorithm** for the RCSP, very different instances are considered in the literature
- ▶ **Our approach is good for RCSP instances** coming from state-of-the-art Branch-Cut-and-Price algorithms **for vehicle routing**
- ▶ **Bucket steps size is a critical** instance-dependent **parameter** for the labelling algorithm
- ▶ **Fixing bucket arcs** by reduced costs **is possible** and may be used by default (does not hurt)
- ▶ **Significant computational improvement over the state-of-the-art** for exact solution of important vehicle routing problems

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New route relaxation and pricing strategies for the vehicle routing problem.

Operations Research, 59(5):1269–1283.



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A tabu search heuristic for periodic and multi-depot vehicle routing problems.

Networks, 30(2):105–119.

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Duhamel, C., Lacomme, P., and Prodhon, C. (2011).

Efficient frameworks for greedy split and new depth first search split procedures for routing problems.

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Dumitrescu, I. and Boland, N. (2003).

Improved preprocessing, labeling and scaling algorithms for the weight-constrained shortest path problem.

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Gehring, H. and Homberger, J. (2002).

Parallelization of a two-phase metaheuristic for routing problems with time windows.

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On an exact method for the constrained shortest path problem.

Computers & Operations Research, 40(1):378 – 384.



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References IV



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Automation and combination of linear-programming based stabilization
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A hybrid genetic algorithm for multidepot and periodic vehicle routing problems.

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Zhu, X. and Wilhelm, W. E. (2012).

A three-stage approach for the resource-constrained shortest path as a sub-problem in column generation.

Computers & Operations Research, 39(2):164 – 178.