DM204 - Autumn 2013
Scheduling, Timetabling and Routing

## Lecture 11 <br> Vehicle Routing

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## Outline

1. Vehicle Routing
2. MILP Models

## Course Overview

$\checkmark$ Scheduling
$\checkmark$ Classification
$\checkmark$ Complexity issues
$\checkmark$ Single Machine

- Parallel Machine
- Flow Shop and Job Shop
- Resource Constrained Project Scheduling Model
- Timetabling
$\checkmark$ Crew/Vehicle Scheduling
- Public Transports
- Workforce scheduling
$\checkmark$ Reservations
$\checkmark$ Education
- Sport Timetabling
- Vehicle Routing
- MILP Approaches
- Construction Heuristics
- Local Search Algorithms


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1. Vehicle Routing

2. MILP Models

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## 2. MILP Models

## Problem Definition

Vehicle Routing: distribution of goods between depots and customers.
Delivery, collection, transportation.
Examples: solid waste collection, street cleaning, school bus routing, dial-a-ride systems, transportation of handicapped persons, routing of salespeople and maintenance unit.

Vehicle Routing Problems
Input: Vehicles, depots, road network, costs and customers requirements.
Output: Set of routes such that:

- requirement of customers are fulfilled,
- operational constraints are satisfied and
- a global transportation cost is minimized.



## Refinement

Road Network

- represented by a (directed or undirected) complete graph
- travel costs and travel times on the arcs obtained by shortest paths

Customers

- vertices of the graph
- collection or delivery demands
- time windows for service
- service time
- subset of vehicles that can serve them
- priority (if not obligatory visit)

Vehicles

- capacity
- types of goods
- subsets of arcs traversable
- fix costs associated to the use of a vehicle
- distance dependent costs
- a-priori partition of customers
- home depot in multi-depot systems
- drivers with union contracts

Operational Constraints

- vehicle capacity
- delivery or collection
- time windows
- working periods of the vehicle drivers
- precedence constraints on the customers


## Objectives

- minimization of global transportation cost (variable + fixed costs)
- minimization of the number of vehicles
- balancing of the routes
- minimization of penalties for un-served customers


## History:

Dantzig, Ramser "The truck dispatching problem", Management Science, 1959
Clark, Wright, "Scheduling of vehicles from a central depot to a number of delivery points". Operation Research. 1964

## Vehicle Routing Problems

- Capacitated (and Distance Constrained) VRP (CVRP and DCVRP)
- VRP with Time Windows (VRPTW)
- VRP with Backhauls (VRPB)
- VRP with Pickup and Delivery (VRPPD)
- Periodic VRP (PVRP)
- Multiple Depot VRP (MDVRP)
- Split Delivery VRP (SDVRP)
- VRP with Satellite Facilities (VRPSF)
- Site Dependent VRP
- Open VRP
- Stochastic VRP (SVRP)
- ...


## Capacitated Vehicle Routing (CVRP)

Input: (common to all VRPs)

- (di)graph (strongly connected, typically complete) $G(V, A)$, where $V=\{0, \ldots, n\}$ is a vertex set:
- 0 is the depot.
- $V^{\prime}=V \backslash\{0\}$ is the set of $n$ customers
- $A=\{(i, j): i, j \in V\}$ is a set of arcs
- $C$ a matrix of non-negative costs or distances $c_{i j}$ between customers $i$ and $j$ (shortest path or Euclidean distance) $\left(c_{i k}+c_{k j} \geq c_{i j} \quad \forall i, j \in V\right)$
- a non-negative vector of costumer demands $d_{i}$
- a set of $K$ (identical!) vehicles with capacity $Q, d_{i} \leq Q$


## Task:

Find collection of $K$ circuits with minimum cost, defined as the sum of the costs of the arcs of the circuits and such that:

- each circuit visits the depot vertex
- each customer vertex is visited by exactly one circuit; and
- the sum of the demands of the vertices visited by a circuit does not exceed the vehicle capacity $Q$.

Note: lower bound on $K$

- $\left\lceil d\left(V^{\prime}\right) / Q\right\rceil$
- number of bins in the associated Bin Packing Problem

A feasible solution is composed of:

- a partition $R_{1}, \ldots, R_{m}$ of $V$;
- a permutation $\pi^{i}$ of $R_{i} \bigcup\{0\}$ specifying the order of the customers on route $i$.

A route $R_{i}$ is feasible if $\sum_{i=\pi_{1}}^{\pi_{m}} d_{i} \leq Q$.

The cost of a given route $\left(R_{i}\right)$ is given by: $F\left(R_{i}\right)=\sum_{j=\pi_{0}^{i}}^{\pi_{m}^{i}} c_{j, j+1}$

The cost of the problem solution is: $F_{V R P}=\sum_{i=1}^{m} F\left(R_{i}\right)$.

Relation with TSP

- VRP with $K=1$, no limits, no (any) depot, customers with no demand $\rightarrow$ TSP
- VRP is a generalization of the Traveling Salesman Problem (TSP) $\rightarrow$ is NP-Hard.
- VRP with a depot, $K$ vehicles with no limits, customers with no demand $\rightarrow$ Multiple TSP $=$ one origin and $K$ salesman
- Multiple TSP is transformable in a TSP by adding $K$ identical copies of the origin and making costs between copies infinite.

Variants of CVRP:

- minimize number of vehicles
- different vehicles $Q_{k}, k=1, \ldots, K$
- Distance-Constrained VRP: length $t_{i j}$ on arcs and total duration of a route cannot exceed $T$ associated with each vehicle Generally $c_{i j}=t_{i j}$ (Service times $s_{i}$ can be added to the travel times of the arcs: $\left.t_{i j}^{\prime}=t_{i j}+s_{i} / 2+s_{j} / 2\right)$
- Distance constrained CVRP


## Vehicle Routing with Time Windows (Valk paidion)

## Further Input:

- each vertex is also associated with a time interval $\left[a_{i}, b_{j}\right]$.
- each arc is associated with a travel time $t_{i j}$
- each vertex is associated with a service time $s_{i}$


## Task:

Find a collection of $K$ simple circuits with minimum cost, such that:

- each circuit visit the depot vertex
- each customer vertex is visited by exactly one circuit; and
- the sum of the demands of the vertices visited by a circuit does not exceed the vehicle capacity $Q$.
- for each customer $i$, the service starts within the time windows [ $a_{i}, b_{i}$ ] (it is allowed to wait until $a_{i}$ if early arrive)


Time windows induce an orientation of the routes.

## Variants

- Minimize number of routes
- Minimize hierarchical objective function
- Makespan VRP with Time Windows (MPTW)
minimizing the completion time
- Delivery Man Problem with Time Windows (DMPTW) minimizing the sum of customers waiting times


## Solution Techniques for CVRP

- Integer Programming
- Construction Heuristics
- Local Search
- Metaheuristics
- Hybridization with Constraint Programming


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## Basic Models

- arc flow formulation integer variables on the edges counting the number of time it is traversed one, two or three index variables
- set partitioning formulation
- multi-commodity network flow formulation for VRPTW integer variables representing the flow of commodities along the paths traveled by the vehicles and integer variables representing times

Two index arc flow formulation

$$
\begin{array}{ll}
\min & \sum_{i \in V} \sum_{j \in V} c_{i j} x_{i j}  \tag{1}\\
\text { s.t. } & \sum_{i \in V} x_{i j}=1 \\
& \sum_{j \in V} x_{i j}=1 \\
& \sum_{i \in V} x_{i 0}=K \\
& \sum_{j \in V} x_{0 j}=K \\
& \sum_{i \in S} \sum_{j \notin S} x_{i j} \geq r(S) \\
& x_{i j} \in\{0,1\}
\end{array}
$$

$$
\begin{equation*}
\forall j \in V \backslash\{0\} \tag{2}
\end{equation*}
$$

$\forall i \in V \backslash\{0\}$
$\forall S \subseteq V \backslash\{0\}, S \neq \emptyset$
$\forall i, j \in V(7)$
$r(S)$ minimum number of vehicles needed to serve set $S$
(6): capacity-cut constraints

One index arc flow formulation

$$
\begin{array}{ll}
\min & \sum_{e \in E} c_{e} x_{e} \\
\text { s.t. } & \sum_{e \in \delta(i)} x_{e}=2 \\
& \sum_{e \in \delta(0)} x_{e}=2 K  \tag{10}\\
& \sum_{e \in \delta(S)} x_{e} \geq 2 r(S) \\
& x_{e} \in\{0,1\} \\
& x_{e} \in\{0,1,2\}
\end{array}
$$

$\forall S \subseteq V \backslash\{0\}, S \neq \emptyset(11)$
$\forall e \notin \delta(0)(12)$
$\forall e \in \delta(0)(13)$
$r(S)$ minimum number of vehicles needed to serve set $S$ $x_{e}=2$ if we allow single visit routes

Three index arc flow formulation

$$
\begin{array}{ll}
\min & \sum_{i \in V} \sum_{j \in V} c_{i j} \sum_{k=1}^{K} x_{i j k} \\
\text { s.t. } & \sum_{k=1}^{K} y_{i k}=1 \\
& \sum_{k=1}^{K} y_{0 k}=K \\
& \sum_{j \in V} x_{i j k}=\sum_{j \in V} x_{j i k}=y_{i k} \\
& \sum_{i \in V} d_{i} y_{i k} \leq C \\
& \sum_{i \in S} \sum_{j \neq S} x_{i j k} \geq y_{h k}  \tag{19}\\
& y_{i k} \in\{0,1\} \\
& x_{i j k} \in\{0,1\}
\end{array}
$$

$$
\forall S \subseteq V \backslash\{0\}, h \in S, k=1, \ldots, K
$$

$$
\begin{aligned}
\forall i \in V, k & =1, \ldots, K(20) \\
\forall i, j \in V, k & =1, \ldots, K(21)
\end{aligned}
$$

## Commodity Flow Models

$y_{i j}, y_{j i}$ two commodities and $x_{i j}$ if arc is in solution

$$
\begin{array}{ll}
\min & \sum_{(i, h) \in A^{\prime}} c_{i j} x_{i j}  \tag{22}\\
\text { s.t. } & \sum_{j \in V^{\prime}}\left(y_{j i}-y_{i j}\right)=2 d_{i} \\
& \sum_{j \in V^{\prime} \backslash\{0, n+1\}} y_{0 j}=d(V \backslash\{0, n+1\} \\
& \sum_{j \in V^{\prime} \backslash\{0, n+1\}} y_{j 0}=K C-d(V \backslash\{0, n+1\} \\
& \sum_{j \in V^{\prime} \backslash\{0, n+1\}} y_{n+1, j}=K C \\
& y_{i j}+y_{j i}=C x_{i j} \\
& \sum_{j \in V^{\prime}}\left(x_{i j}+x_{j i}\right)=2 \\
& y_{i j} \geq 0 \\
& x_{i j} \in\{0,1\}
\end{array}
$$

$\forall i \in V \backslash\{0, n+1\}$
$\forall i j \in A^{\prime} \quad$ (27)
$\forall i \in V \backslash\{0, n+1\}$
$\forall i j \in A^{\prime} \quad$ (29)
$\forall i j \in A^{\prime}$ (30)

## Set Partitioning Formulation

$$
\begin{align*}
& \mathcal{R}=\{1,2, \ldots, R\} \text { index set of routes } \\
& a_{i r}= \begin{cases}1 & \text { if costumer } i \text { is served by } r \\
0 & \text { otherwise }\end{cases} \\
& x_{r}= \begin{cases}1 & \text { if route } r \text { is selected } \\
0 & \text { otherwise }\end{cases} \\
& \text { min } \sum_{r \in \mathcal{R}} c_{r} x_{r}  \tag{31}\\
& \text { s.t. } \quad \sum_{r \in \mathcal{R}} a_{i r} x_{r}=1  \tag{32}\\
& \sum_{r \in \mathcal{R}} x_{r} \leq K  \tag{33}\\
& x_{r} \in\{0,1\} \tag{34}
\end{align*}
$$

$$
\forall i \in V \quad(32)
$$

What can we do with these integer programs?

- plug them into a commercial solver and try to solve them
- preprocess them
- determine lower bounds
- solve the linear relaxation
- combinatorial relaxations relax some constraints and get an easy solvable problem
- Lagrangian relaxation
- polyhedral study to tighten the formulations
- upper bounds via heuristics
- branch and bound
- cutting plane
- branch and cut
- Dantzig Wolfe decomposition
- column generation (via reformulation)
- branch and price


## Combinatorial Relaxations

Relax: capacity cut constraints (CCC) or generalized subtour elimination constraints (GSEC)
Consider both ACVRP and SCVRP

- Relax CCC in 2-index formulation obtain a transportation problem Solution may contain isolated circuits and exceed vertex capacity
- Relax CCC in 1-index formulation obtain a b-matching problem

$$
\begin{array}{lr}
\min & \sum_{e \in E} c_{e} x_{e} \\
\text { s.t. } & \sum_{e \in \delta(i)} x_{e}=b_{i} \\
& x_{e} \in\{0,1\} \\
& \forall i \in V, b_{0}=2 K, b_{i}=2 \forall i \neq 0 \\
& \forall e \notin \delta(0) \\
& \forall 0,1,2\}
\end{array}
$$

Solution has same problems as above

- relax in 2-index flow formulation:

$$
\begin{array}{llr}
\min & \sum_{i \in V} \sum_{j \in V} c_{i j} x_{i j} & \\
\text { s.t. } & \sum_{i \in V} x_{i j}=1 & \forall j \in V \backslash\{0\} \\
& \sum_{j \in V} x_{i j}=1 & \forall i \in V \backslash\{0\} \\
& \sum_{i \in V} x_{i 0}=K & \\
& \sum_{j \in V} x_{0 j}=K & \forall S \subseteq V \backslash\{0\}, S \neq \emptyset \\
& \sum_{i \in S} \sum_{i \notin S} x_{i j} \geq r(S) 1 & \forall i, j \in V
\end{array}
$$

K-shortest spanning arborescence problem

- relax in 1-index formulation

$$
\left.\begin{array}{ll}
\min & \sum_{e \in E} c_{e} x_{e} \\
\text { s.t. } & \sum_{e \in \delta(i)} x_{e}=2 \\
& \sum_{e \in \delta(0)} x_{e}=2 K \\
& \forall i \in V \backslash\{0\} \\
& \sum_{e \in \delta(S)} x_{e} \geq 2 r(S) \\
& x_{e} \in\{0,1\}
\end{array} \forall S \subseteq V \backslash\{0\}, S \neq \emptyset\right\}
$$

K-tree: min cost set of $n+K$ edges spanning the graph with degree $2 K$ at the depot.

- Lagrangian relaxation of the sub tour constraints iteratively added after violation recognized by separation procedure. Subgradient optimization for the multipliers.


## Branch and Cut

$$
\begin{align*}
& \min \sum_{e \in E} c_{e} x_{e}  \tag{35}\\
& \text { s.t. } \sum_{e \in \delta(i)} x_{e}=2  \tag{36}\\
& \sum_{e \in \delta(0)} x_{e}=2 K  \tag{37}\\
& \sum_{e \in \delta(S)} x_{e} \geq 2\left\lceil\frac{d(S)}{C}\right\rceil \\
& x_{e} \in\{0,1\} \\
& x_{e} \in\{0,1,2\} \\
& \forall S \subseteq V \backslash\{0\}, S \neq \emptyset \text { (38) } \\
& \forall e \notin \delta(0) \quad(39) \\
& \forall e \in \delta(0) \text { (40) }
\end{align*}
$$

## Branch and Cut

- Let $L P(\infty)$ be linear relaxation of IP (with all cuts)
- $z_{L P(\infty)} \leq z_{I P}$
- Problems if many constraints
- Solve $L P(h)$ instead and add constraints later ( $h$ cuts included)
- If $L P(h)$ has integer solution and no constraint unsatisfied then we are done, that is optimal If not, then $z_{L P(h)} \leq z_{L P(h+1)} \leq z_{L P(\infty)} \leq z_{I P}$
- Crucial step: separation algorithm given a solution to $L P(h)$ it tell us if some constraints are violated.

If the procedure does not return an integer solution we proceed by branch and bound

## Problems with B\&C:

i) no good algorithm for the separation phase it may be not exact in which case bounds relations still hold and we can go on with branching
ii) number of iterations for cutting phase is too high
iii) program unsolvable because of size
iv) the tree explodes

The main problem is (iv).
Worth trying to strengthen the linear relaxation by inequalities that although unnecessary in the integer formulation force the optimal solution of LP and IP to get closer. Polyhedral studies

## On the Set Covering Formulation

Solving the SCP integer program
Branch and bound

- generate routes such that:
- they are good in terms of cost
- they reduce the potential for fractional solutions
- constraint branching [Ryan, Foster, 1981]

$$
\exists \text { constraints } r_{1}, r_{2}: 0<\sum_{j \in J\left(r_{1}, r_{2}\right)} x_{j}<1
$$

$J\left(r_{1}, r_{2}\right)$ all columns covering $r_{1}, r_{2}$ simultaneously. Branch on:

$$
\sum_{j \in J\left(r_{1}, r_{2}\right)} x_{j} \leq 0 \quad \quad \sum_{j \in J\left(r_{1}, r_{2}\right)} x_{j} \geq 1
$$

Solving the SCP linear relaxation
Column Generation Algorithm
Step 1 Generate an initial set of columns $\mathcal{R}^{\prime}$
Step 2 Solve problem $P^{\prime}$ and get optimal primal variables, $\bar{x}$, and optimal dual variables, $(\bar{\pi}, \bar{\theta})$
Step 3 Solve problem CG, or identify routes $r \in \mathcal{R}$ satisfying $\bar{c}_{r}<0$
Step 4 For every $r \in \mathcal{R}$ with $\bar{c}_{r}<0$ add the column $r$ to $\mathcal{R}^{\prime}$ and go to Step 2
Step 5 If no routes $r$ have $\bar{c}_{r}<0$, i.e., $\bar{c}_{\text {min }} \geq 0$ then stop.
In most of the cases we are left with a fractional solution

Solving the SCP integer program:

- cutting plane
- branch and price


## Cutting Plane Algorithm

Step 1 Generate an initial set $\mathcal{R}^{\prime}$ of columns
Step 2 Solve, using column generation, the problem $P^{\prime}$ (linear programming relaxation of $P$ )
Step 3 If the optimal solution to $P^{\prime}$ is integer stop.
Else, generate cutting plane separating this fractional solution.
Add these cutting planes to the linear program $P^{\prime}$
Step 4 Solve the linear program $P^{\prime}$. Goto Step 3.
Is the solution to this procedure overall optimal?

## Cuts

Intersection graph $G=(V, E): V$ are the routes and $E$ is made by links between routes that have at least a customer in common (Independence set in $G$ is a collection of routes where each customer is visited only once.)

Clique constraints

$$
\sum_{r \in K} \bar{x}_{r} \leq 1 \quad \forall \text { cliques } K \text { of } G
$$

Cliques searched heuristically
Odd holes
Odd hole: odd cycle with no chord

$$
\sum_{r \in H} \bar{x}_{r} \leq \frac{|H|-1}{2} \quad \forall \text { odd holes } H
$$

Generation via layered graphs

$$
\left[\begin{array}{llllll}
1 & 1 & 1 & & & \\
1 & & 1 & 1 & & \\
1 & & & 1 & 1 & \\
1 & & & & 1 & 1 \\
1 & 1 & & & & 1
\end{array}\right]\left[\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3} \\
x_{4} \\
x_{5} \\
x_{6}
\end{array}\right] \leq\left[\begin{array}{l}
1 \\
1 \\
1 \\
1 \\
1
\end{array}\right]
$$


$x^{\prime}=[0 ; 0.5 ; 0.5 ; 0.5 ; 0.5]$ is an extreme point of the polytope and clique constraints would not prevent it (although they define facets they do not define all the facets). The constraint on the odd hole $H=\{2,3,4,5,6\}$ :

$$
x_{2}+x_{3}+x_{4}+x_{5}+x_{6} \leq 2
$$

cuts off solution $x^{\prime}$. (this is facet defining for $H$ but not for $G$ )

Branch and price

- it must be possible to incorporate information on the node in the column generation procedure
- easy to incorporate $x_{r}=1$, just omit nodes in $S_{r}$ from generation; but not clear how to impose $x_{r}=0$.
- different branching: on the edges: $x_{i j}=1$ then in column generation set $c_{i j}=-\infty$; if $x_{i j}=0$ then $c_{i j}=\infty$

