

Capacitated Vehicle Routing and Some Related Problems

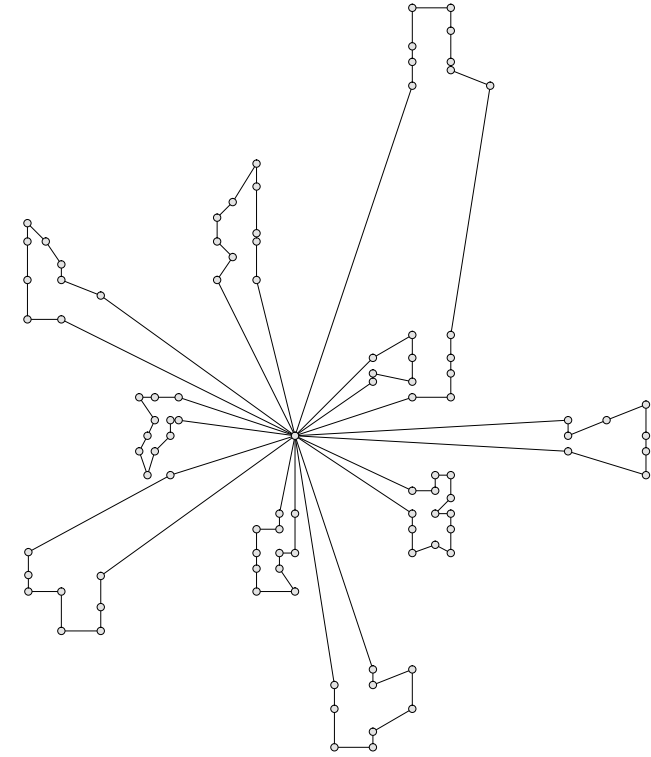
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Rutgers University, November 27, 2001

Outline of Talk

- Introduction and Motivation
- A New Approach
- Complexity and Special Cases
- Valid Inequalities
- Implementation
- Computational Issues and Results
- Future Directions



The Vehicle Routing Problem

The **VRP** is a combinatorial problem whose *ground set* is the edges of a graph $G(V, E)$. Notation:

- V is the set of customers and the depot (0).
- d is a vector of the customer **demands**.
- k is the number of **routes**.
- C is the **capacity** of a truck.

A **feasible solution** is composed of:

- a **partition** $\{R_1, \dots, R_k\}$ of V such that $\sum_{j \in R_i} d_j \leq C$, $1 \leq i \leq k$;
- a **permutation** σ_i of $R_i \cup \{0\}$ specifying the order of the customers on route i .

Classical Formulation for the VRP

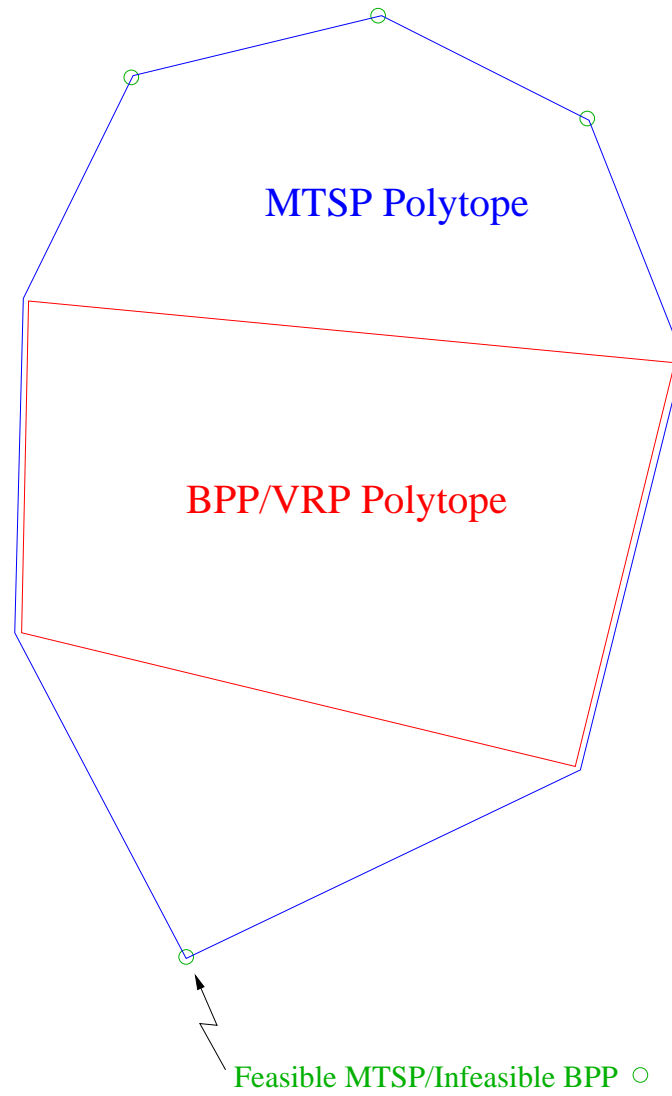
IP Formulation:

$$\begin{aligned}\sum_{j=1}^n x_{0j} &= 2k \\ \sum_{j=1}^n x_{ij} &= 2 \quad \forall i \in V \setminus \{0\} \\ \sum_{\substack{i \in S \\ j \notin S}} x_{ij} &\geq 2b(S) \quad \forall S \subset V \setminus \{0\}, |S| > 1.\end{aligned}$$

$b(S)$ = lower bound on the number of trucks required to service S (normally $\lceil (\sum_{i \in S} d_i) / C \rceil$).

If $C = \sum_{i \in S} d_i$, then we have the Multiple Traveling Salesman Problem.

Alternatively, if the edge costs are all zero, then we have the Bin Packing Problem.



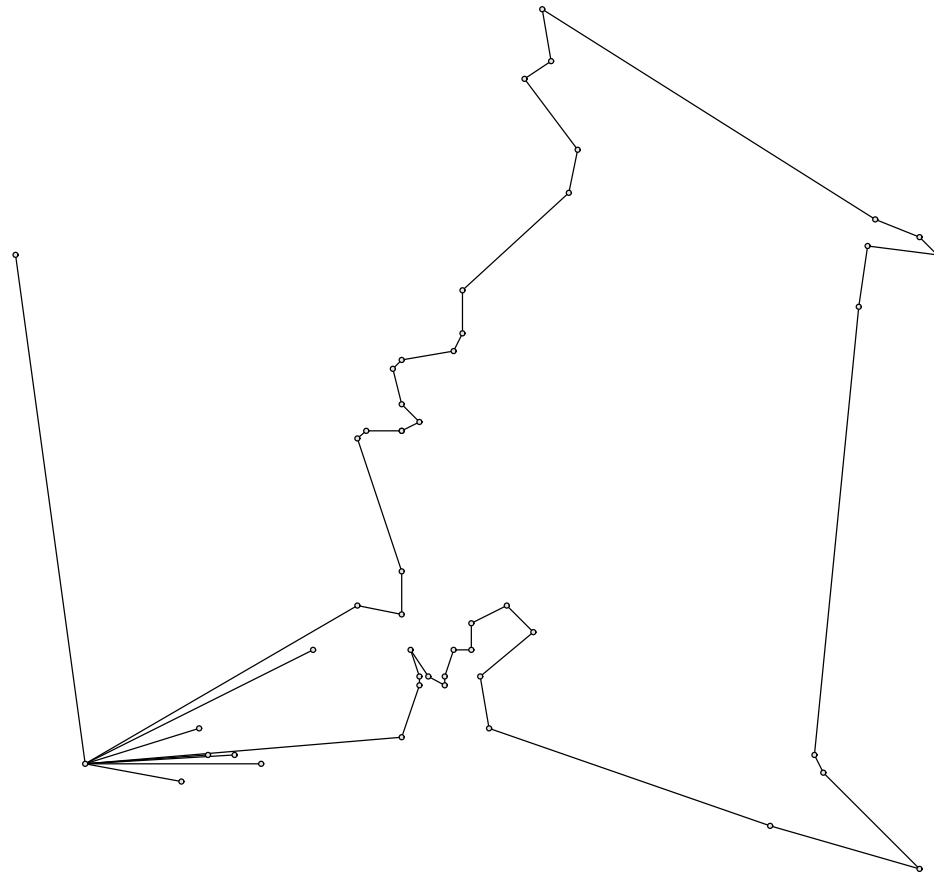
How Hard is the VRP?

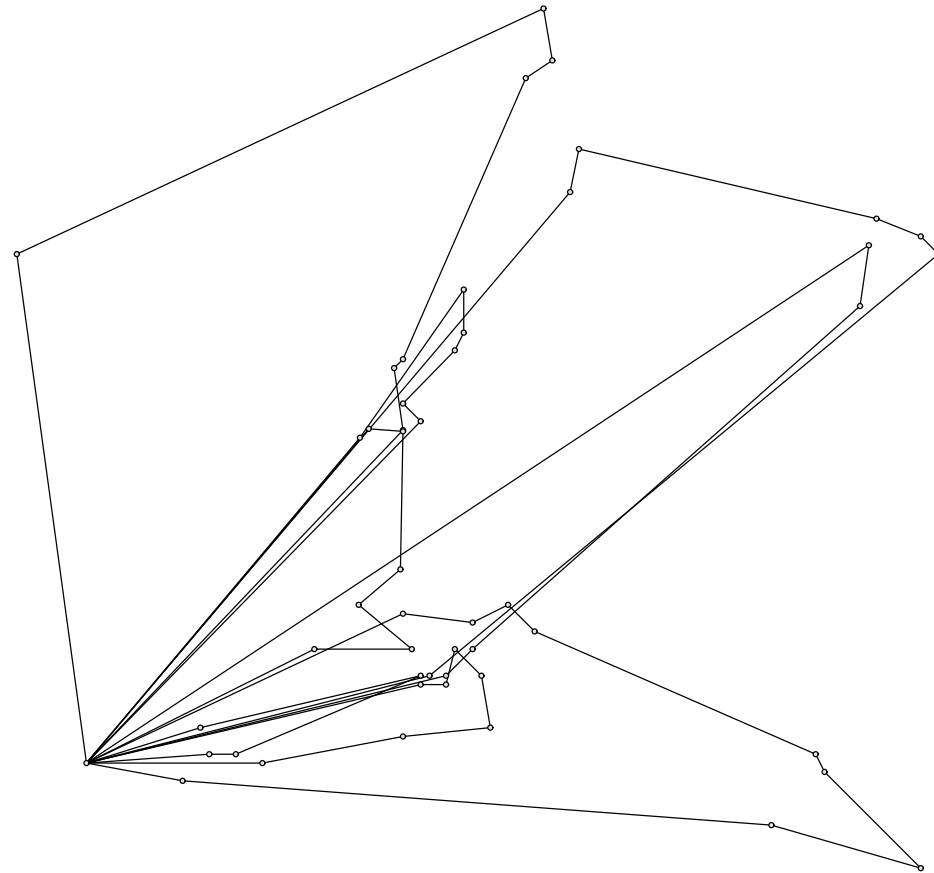
- Test Set
 - TSPLIB/VRPLIB
 - Augerat's repository
 - Available at BranchAndCut.org/VRP
- Largest VRP instance solved: F-n135-k7
- Smallest VRP instance unsolved: B-n50-k8
- Largest TSP instance solved: usa13509
- Time to solve B-n50-k8 as an MTSP: .1 sec
- Why the gap?

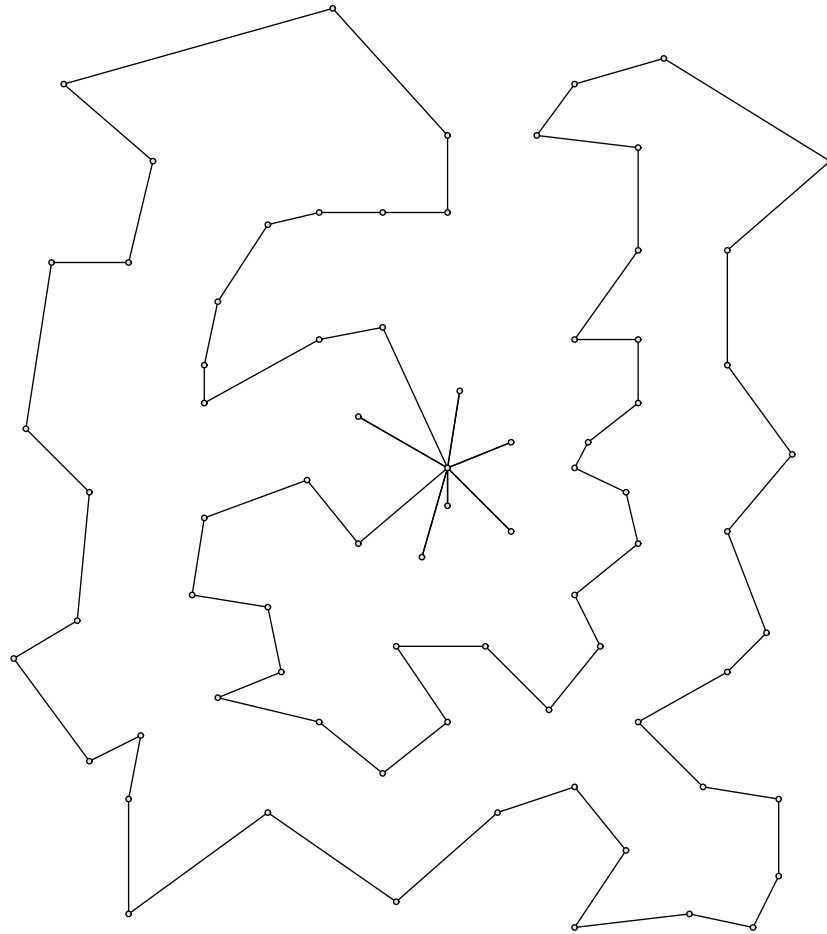
Standard Approach

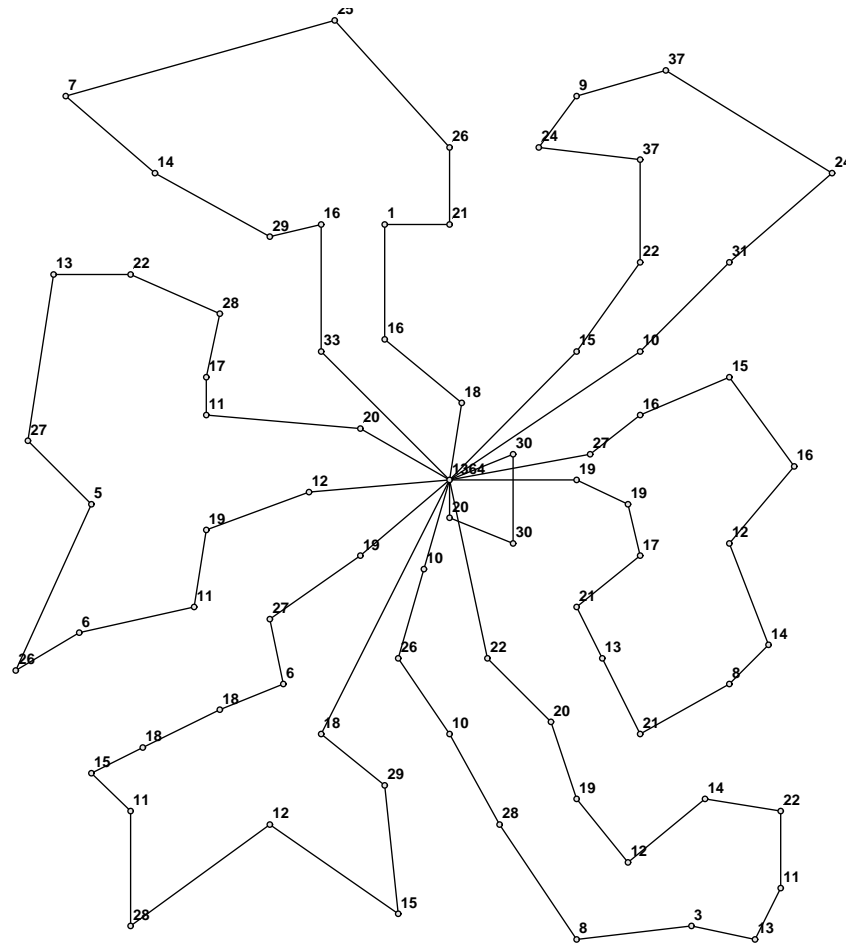
- Standard approaches treat the **VRP** in much the same way as the **TSP**.
 - Most known valid inequalities are generalizations from the **TSP**.
 - Branching rules are also generalizations from the **TSP**.
- However, the TSP does not seem to be the right template.
- It is the packing, not the routing that makes the problem difficult.





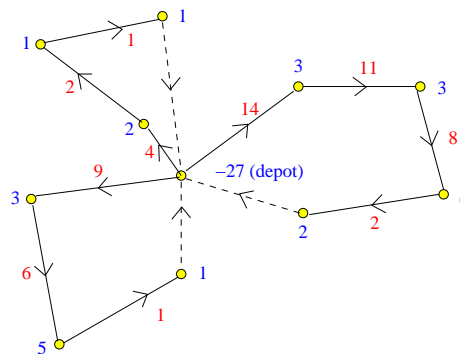






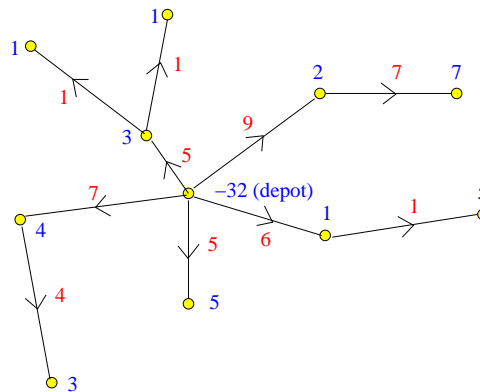
What Makes the VRP Difficult?

- It is the intersection of two difficult problems.
 - Traveling Salesman Problem (Routing)
 - Bin Packing Problem (Packing)
- We don't have an effective, polynomially sized relaxation.
- Current approaches treat it as a routing problem.
- We know very little about the packing aspect.
- We need a different template.
- Idea: Consider flow-based formulations.



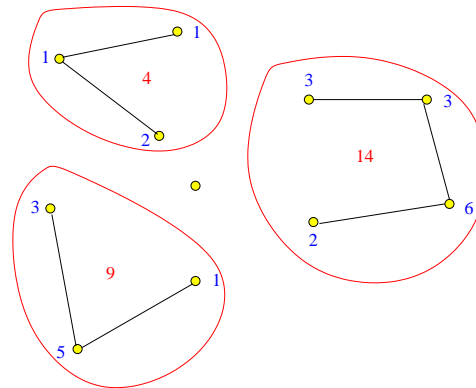
Node Routing

- We are given an undirected graph $G = (V, E)$.
 - The nodes represent **supply/demand** points.
- We consider problems with one supply point (the *depot*).
- A *node routing* is a directed subgraph G' of G satisfying the following properties:
 - G' is (weakly) connected.
 - The **in-degree** of each non-depot node is 1.



Capacitated Node Routing

- A *capacitated node routing* is one in which the demand in each component of $G' \setminus \{0\}$ is $\leq C$.
- Feasible solutions are bin packings.
- This restriction is easily modeled using a flow-based formulation.
- With capacities, we can model the **VRP** and the **Capacitated Spanning Tree Problem (CSTP)**.

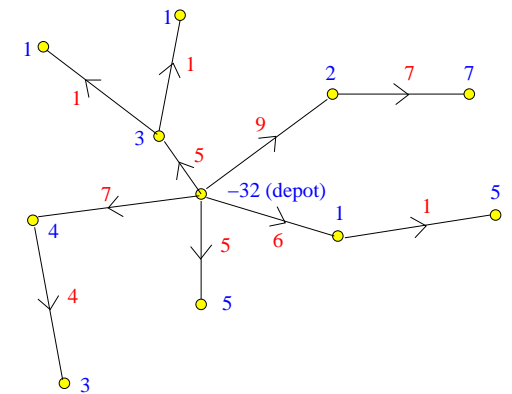


Optimal Node Routing

- Properties of a **node routing**.
 - It is a **spanning arborescence** plus (possibly) some edges returning to the depot.
 - There is a **unique path** from the depot to each demand point.
- We wish to construct a **least cost routing**.

- Cost Measures

- Lengths of all edges in G' .
- Length of all paths from the depot.
- Linear combination of these two.



IP Formulation

IP formulation for this routing problem:

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

$$\text{s.t.} \quad x(\delta(V \setminus \{i\})) = 1 \quad \forall i \in V \setminus \{0\}$$

$$f(\delta(V \setminus \{i\})) - f(\delta(\{i\})) = d_i \quad \forall i \in V \setminus \{0\}$$

$$0 \leq f_{ij} \leq C x_{ij} \quad \forall (i, j) \in A$$

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

where:

- x_{ij}, x_{ji} (*fixed-charge variables*) indicate whether $\{i, j\}$ is in the routing *and its orientation*.
- f_{ij} (*flow variable*) represents demand flow from i to j .

Complexity

- This node routing problem is **NP-complete** even in the uncapacitated case (fixed-charge network flow problem).
- Polynomially solvable special cases.
 - $\tau = 0 \Rightarrow$ **Minimum Spanning Tree Problem.**
 - $\gamma = 0 \Rightarrow$ **Shortest Paths Tree Problem.**
 - Note that **demands are irrelevant.**
- Other special cases.
 - $\tau = 0 \Rightarrow$ **Capacitated Spanning Tree Problem.**
 - $\tau, \gamma > 0 \Rightarrow$ **Cable-Trench Problem.**
 - $\tau = 0$ and $x(\delta(\{i\})) = 1 \Rightarrow$ **Traveling Salesman Problem.**
 - $\tau > 0$ and $x(\delta(\{i\})) = 1 \Rightarrow$ **Variable Cost TSP.**
 - $x(\delta(V \setminus \{0\})) = x(\delta(\{0\})) = k \Rightarrow$ **VRP.**

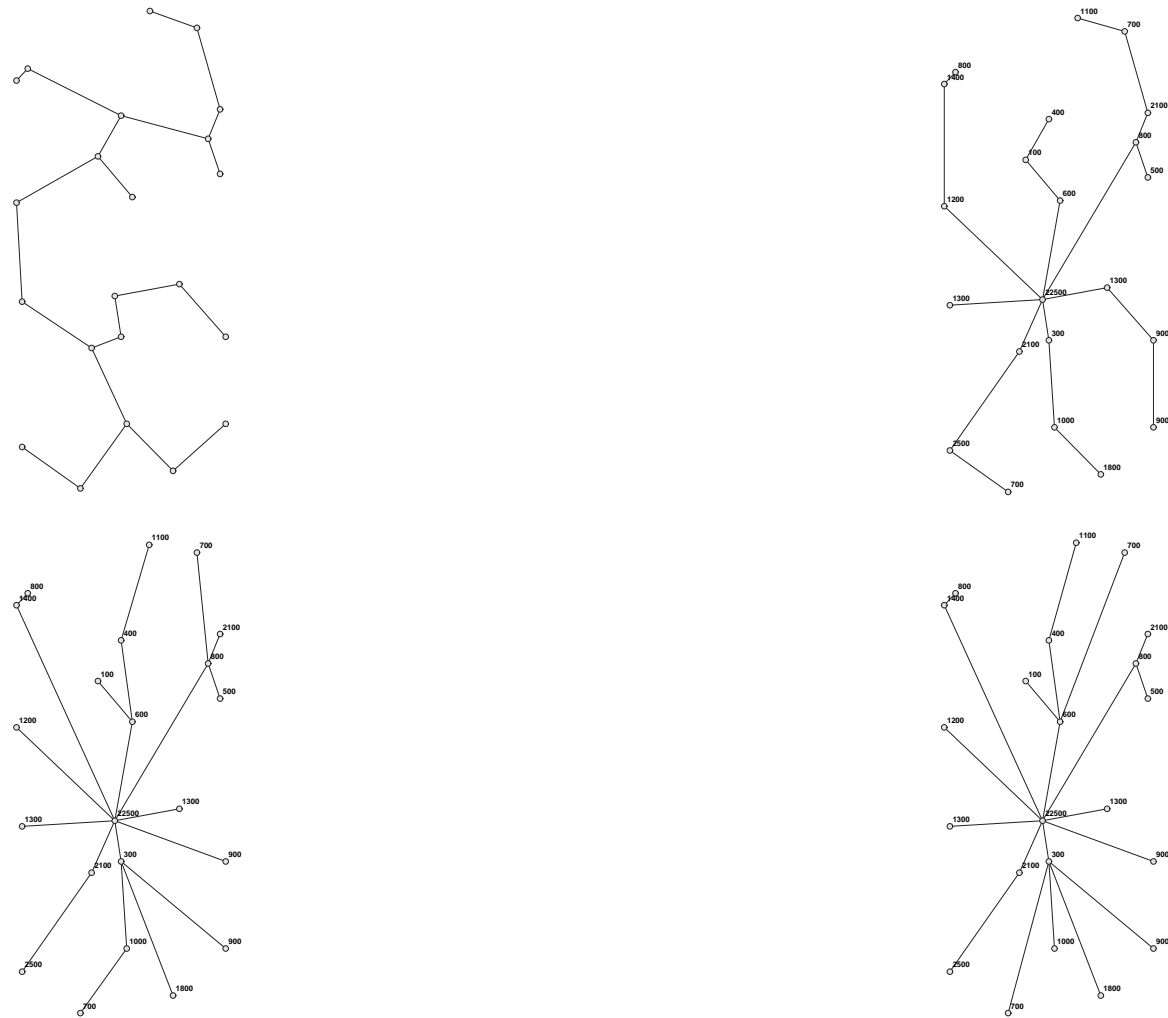


Figure 1: Optimal uncapacitated spanning trees with increasing τ/γ ratios



Figure 2: Uncapacitated vs. capacitated spanning trees ($\tau = 0$)

Connection to Other Models

- There are connections to many well-studied models that may provide better templates.
- The basic model can be seen as an instance of the **Fixed-charge Network Flow Problem**.
- Removing the upper bounds on the fixed-charge variables yields the **Capacitated Network Design Problem**.
- We have already mentioned several other related combinatorial models.
- We are looking to make stronger connections among these varied areas of the literature.

Valid Inequalities

- Note that any inequalities valid for the **TSP**, **VRP**, or **CSTP** have counterparts here.
- Many can be strengthened by taking advantage of the directed formulation.
- Fractional Capacity Constraints

$$\sum_{i \notin S, j \in S} x_{ij} \geq d(S)/C, \quad 0 \notin S$$

- Multi-star Inequalities

$$\sum_{i \notin S, j \in S} x_{ij} \geq d(S)/C + \frac{\sum_{i \notin S, j \in S} x_{ji} d_i}{C}, \quad 0 \notin S$$

Valid Inequalities

- Rounded Capacity Constraints

$$\sum_{i \notin S, j \in S} x_{ij} \geq \lceil d(S)/C \rceil$$

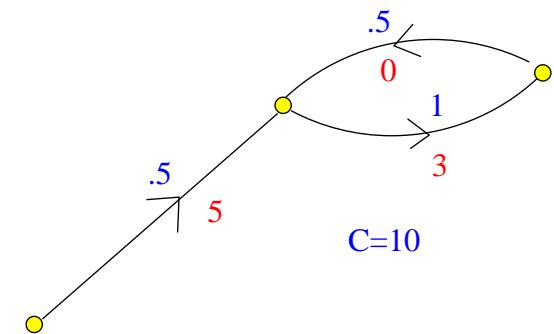
- Generalized, framed capacity constraints
- Combs, Hypo-tours, Clique Clusters
- Path-bin inequalities

Flow Linking

- Note that only the edge variables are required to be integral.
- We use the flow variables to force integrality of the edge variables through *flow linking constraints*.
- Flow Linking Constraints

$$f_{ij} \leq (C - d_i)x_{ij} \Leftrightarrow x_{ij} \geq \frac{f_{ij}}{C - d_i}$$

$$f_{ij} - \sum_{k \neq j} f_{jk} \leq x_{ij}d_j$$

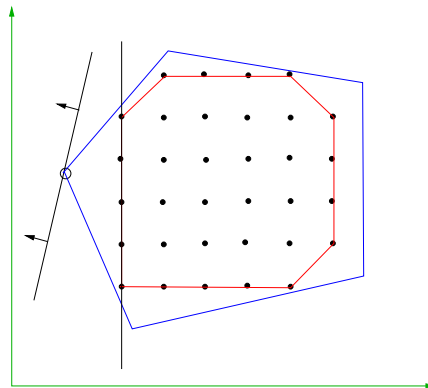


- Edge Cuts

$$x_{ij} + x_{ji} \leq 1$$

Separation

- The **fractional capacity constraints** and **multi-star inequalities** are automatically satisfied.
- **Flow linking constraints** and **edge cuts** can be included explicitly or separated in **polynomial time**.
- Separating **rounded capacity constraints** is **NP-complete**, but can be done effectively.
- Heuristic procedures for other classes have not yet been implemented.



Solver Implementation

- The implementation uses **SYMPHONY**, a parallel framework for branch, cut, and price (relative of COIN/BCP).
- In **SYMPHONY**, the user supplies:
 - the initial LP relaxation,
 - separation subroutines,
 - feasibility checker, and
 - other optional subroutines.
- **SYMPHONY** handles everything else.
- The source code and documentation are available from www.BranchAndCut.org

Preliminary Computation: Formulation Issues

- The new formulation is **polynomial** and yields **stronger relaxations** initially, but there are drawbacks.
- For the **VRP**, the formulation creates **symmetry**.
- It also seems to make branching less effective.
- There is a related “undirected” formulation which uses one fixed-charge variable per edge.
 - This formulation is smaller and performs much better for the **VRP**.
 - For the **CSTP** and **CTP**, however, the undirected formulation is extremely weak.

Preliminary Computation: Results So Far

- So far, the presence of the flow variables does not seem to help.
- **Capacitating** the model does increase difficulty significantly.
- Consider relaxations of the **VRP**.
 - The **TSP** is very easy relative to the **VRP**.
 - The **CSTP** is not much easier than the **VRP**.
- Versions of these models with positive variable (flow) costs are extremely difficult.
 - Is this due to the **upper bound** or **lower bound**?
 - The flow linking constraints are important for these models.

problem	<i>TSP</i>		<i>CSTP</i>		<i>VRP</i>	
	Tree Size	CPU sec	Tree Size	CPU sec	Tree Size	CPU sec
eil13	1	0.00	13	0.09	1	0.00
eil22	1	0.11	2	0.10	1	0.02
eil33	1	0.02	69	3.97	2	0.44
bayg29	1	0.12	1	0.04	4	0.32
bays29	1	0.17	15	1.12	5	0.55
ulysses16.tsp	1	0.00	1	0.03	1	0.01
ulysses22.tsp	1	0.00	1	0.06	1	0.03
gr17	1	0.01	5	0.05	1	0.01
gr21	1	0.00	1	0.02	1	0.03
gr24	1	0.02	5	0.27	4	0.40
fri26	1	0.02	1	0.07	8	0.39
swiss42	1	0.02	35	3.66	10	2.45
att48	2	0.30	92	5.04	193	30.10
gr48	2	1.38	1	0.07	16	4.17
hk48	1	0.19	209	22.88	45	21.19
eil51	1	0.16	77	15.11	11	10.79
A – n32 – k5	1	0.02	1	0.07	2	0.20
A – n33 – k5	3	0.81	3	0.21	7	0.90
A – n34 – k5	6	2.06	4	0.40	9	2.63
A – n36 – k5	1	0.03	52	5.17	51	7.95
A – n37 – k5	1	0.03	5	0.22	11	0.97
A – n38 – k5	1	0.10	1	0.13	111	21.80
A – n39 – k5	1	0.30	11	0.99	480	310.92
A – n44 – k6	3	1.72	586	84.08	1185	1525.78
A – n45 – k6	2	0.27	47	6.19	133	145.59
A – n46 – k7	1	1.25	3	0.20	2	1.95
A – n48 – k7	2	2.01	775	507.41	1949	1620.57
A – n53 – k7	1	0.62	115	19.99	619	881.05
B – n31 – k5	1	0.01	3	0.63	1	0.08
B – n38 – k6	1	0.04	5	0.56	14	1.73
B – n39 – k5	1	0.03	188	9.67	1	0.05
B – n41 – k6	1	0.08	216	18.96	20	2.89
B – n43 – k6	1	0.09	1	0.36	138	34.92
B – n45 – k5	1	0.09	22	1.13	18	5.81
B – n51 – k7	1	0.36	1	0.13	199	39.48

Conclusions and Future Directions

- We have established interesting connections to other well-studied models.
- The TSP does not seem to be the right template to follow.
- We have yet to take full advantage of the information provided by the flow variables.
- **Better flow linking** seems to be the key.
- We also need some **new branching rules**.
- The connection to the network design literature needs to be explored.
- We are also considering **decomposition-based methods**.