

### **VRP** with stochastic demands

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### **Capacitated vehicle routing problem static version**



### **Capacitated vehicle routing problem** (CVRP):

- •Visit all customers exactly once.
- •Each customer has a demand  $q_i$ .
- Each vehicle has a capacity *Q*.
- •Sum of demands on a route  $\leq$  vehicle capacity
- •**Objective:** Minimize the cost of edges used in the solution.





### VRP with stochastic demands (VRPSD)

### vehicle routing problem with stocastic demand (VRPSD)

•Visit all customers.

•The demand of each customer *i* is given by a random variable  $q_i$ . Its probability distribution is assumed to be known. We assume the variables are independent.

• Each vehicle has a capacity Q.

•We have to plan routes in advance. When executing the plan it may turn out that we the planned route violate the capacity constraint. In that case the vehicle returns to the depot to empty/restock the vehicle.

•Expected demand on each route cannot exceed *Q*.

•**Objective:** Minimize the expected cost of the solution.



### VRP with stochastic demands (VRPSD)



Assume all customers have expected demand 2.

One realization of the random variables: 3, 2, 2

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### **VRP** with stochastic demands





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### VRP with stochastic demands (VRPSD)



#### Vehicle capacity: 6

The extra trip back to the depot is the "recourse action".

We make decisions in stage 1 where we plan the routes for our vehicles.

In stage 2 the routes are followed blindly and any infeasibilities are "repaired" using the recourse action.

This type of problem is a *two-stage stochastic problem with recourse.* 

Cost of extra trips to the depot is called *recourse cost.* 



### VRP with stochastic demands (VRPSD)

What if realization of the random variables was: 3, 2, 2 ?

In that case we will still continue to node 3 even though the vehicle is full after visiting node 1.







Paper	Method	Test data	Demand <= Q
Gendreau, Laporte, Seguin	Integer L- shaped method (branch and cut)	Uniform distributions in different intervals	Yes
Laporte Louveaux, Van Hamme	Integer L- shaped method (branch and cut)	Poisson and normal distribution	Yes
Jabali, Rei, Gendreau, Laporte, 2013	Integer L- shaped method (branch and cut)	Truncated normal distribution	yes
Christiansen & Lysgaard 2007	Branch and price	Poisson distribution	Yes

### Formulating VRPSD using a set partitioning model



- $V_c = \{1, \ldots, n\}$ . Set of customers.
- $\Omega$  : set of all feasible VRPSD routes.
- $c_p$  : expected cost of route  $p \in \Omega$ .
- *a<sub>ip</sub>*: constant that is 1 if customer *i* is visited by route *p* and 0 othervise.
- $y_p$ : binary variable that is 1 if and only if path  $p \in \Omega$  is used in the solution.  $\min \sum c_n u_n$

$$\min\sum_{p\in\Omega}c_py_p$$

Subject to:

$$\sum_{p \in \Omega} a_{ip} y_p = 1 \hspace{1cm} orall i \in V_c$$
 $y_p \in \{0,1\} \hspace{1cm} orall p \in \Omega$ 

### Formulating VRPSD using a set partitioning model



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- $a_{ip}$ : constant that is 1 if customer i is visited by route p and 0 othervise.
- $y_p$ : binary variable that is 1 if and only if path  $p \in \Omega$  is used in the solution. min **Expected cost consist of:** 
  - Normal cost of traversed arc +
  - Expected cost of recourse actions

$$\sum_{p \in \Omega} a_{ip} y_p = 1 \qquad \forall i \in V_c$$

$$y_p \in \{0,1\}$$
  $orall p \in \Omega$ 

Subject to:

### **Properties of set partitioning approach**



- LP relaxation of set partitioning problem is solved by column generation
- ... and the entire problem is solved by branch-(and-cut)-and-price.
- Stochasticity is "removed" from master problem and "hidden" away in definition of variables.
- If we can compute expected recourse cost correctly when solving the pricing problem then we can also solve the entire problem.
- We can reuse many techniques developed for static CVRP since stochasticity is hidden away in sub-problem.
- Approach was first proposed in:
- C.H. Christiansen and J. Lysgaard. *A branch-and-price algorithm for the capacitated vehicle routing problem with stochastic demands*. Operations Research Letters, 35:773?781, 2007

### Calculating expected recourse cost of a route

- Let p(s) be the customer at position s in the route and let  $\xi_i$  be the random variable associated with the demand of customer i.
- The probability of having the first failure at customer j is:

$$\mathbb{P}\Big(\sum_{s=1}^{j-1}\xi_{p(s)} \le Q < \sum_{s=1}^{j}\xi_{p(s)}\Big)$$

The probability of having the l'th failure at customer j is:

$$\mathbb{P}\Big(\sum_{s=1}^{j-1}\xi_{p(s)} \le lQ < \sum_{s=1}^{j}\xi_{p(s)}\Big)$$

the expected recourse cost associated with customer j is

$$\sum_{l=1}^{\infty} \mathbb{P}\Big(\sum_{s=1}^{j-1} \xi_{p(s)} \le lQ < \sum_{s=1}^{j} \xi_{p(s)}\Big) 2c_{0,p(j)}$$

the expected recourse cost of the route is (the number of customers on the route is t)

$$Q_r = \sum_{j=1}^t \sum_{l=1}^\infty \mathbb{P}\Big(\sum_{s=1}^{j-1} \xi_{p(s)} \le lQ < \sum_{s=1}^j \xi_{p(s)}\Big) 2c_{0,p(j)}$$

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### Calculating expected recourse cost of a route

(from last slide): The expected recourse cost of the route is (the number of customers on the route is t)

$$Q_r = \sum_{j=1}^t \sum_{l=1}^\infty \mathbb{P}\left(\sum_{s=1}^{j-1} \xi_{p(s)} \le lQ < \sum_{s=1}^j \xi_{p(s)}\right) 2c_{0,p(j)}$$

Define

$$F^{j}(x) = \mathbb{P}\left(\sum_{s=1}^{j} \xi_{p(s)} \le x\right)$$

Notice that  $F^{j}$  is the cumultative distribution function of the random variable  $\sum_{s=1}^{j} \xi_{p(s)}$ . The expected recourse cost of the route can be rewritten

$$Q_r = 2\sum_{j=1}^{t} \sum_{l=1}^{\infty} \left( F^{j-1}(lQ) - F^j(lQ) \right) c_{0i_j}$$

### **Assumptions needed**



- Customer demands are independent random variables
- $\bullet$  All demands should follow the same distribution  $\Psi.$
- $\bullet$  The distribution ( $\Psi$ ) should have the accumulative property
  - That is, the sum of two or more  $\Psi$ -distributed random variables is itself a  $\Psi$  –distributed random variable
  - We should be able to describe the distribution of each random variable by a number of parameters
- Examples of distributions that have the accumulative property: normal, Gamma, poisson, exponential distribution.
- In the following we use the poison distribution

### **Poisson distribution**





if  $\xi_i$  and  $\xi_j$  have a Poisson distribution with parameter  $\lambda_i$  and  $\lambda_j$ , respectively then  $\xi_i + \xi_j$  is Poisson distributed with parameter  $\lambda_i + \lambda_j$ 

### Christiansen/Lysgaard insight



- Let  $P_1$  and  $P_2$  be two different partial routes from the depot to customer *i*.
- Let  $\xi(P_j)$  be the random variable corresponding to the sum of the variables corresponding customers visited by route  $P_j$ .
- Assume that  $\xi(P_1)$  and  $\xi(P_2)$  have the same distribution.
- Assume further that the expected cost of  $P_1$  is lower than that of  $P_2$  and that any extension of  $P_2$  also is feasible for  $P_1$
- In this case  $P_1$  dominates  $P_2$ .

# Christiansen/Lysgaard pricing algorithm (Poisson distribution)

- Shortest path computation in an expanded graph.
- Special care taken to eliminates two-cycles (ordinary shortst path computation would allow them.
- Higher order cycles are possible.



### Christiansen/Lysgaard pricing algorithm - interpretted as labeling algorithm



• The Chrisitansen / Lysgaard approach can be implemented with a labeling algorithm with labels

$$L = (i, c, \lambda)$$

with *i* being node, *c* expected cost of the partial path represented by label,  $\lambda$  parameter for distribution. A label is feasible if  $\lambda \leq Q$ .

• Dominance: Label  $L_1$  dominates  $L_2$  if

$$i_1 = i_2 \land c_1 \le c_2 \land \lambda_1 = \lambda_2$$

• We show that this can be improved to:

$$i_1 = i_2 \land c_1 \le c_2 \land \lambda_1 \le \lambda_2$$

for the poisson distribution and

$$i_1 = i_2 \wedge c_1 \le c_2 \wedge \frac{1}{\lambda_1} \le \frac{1}{\lambda_2}$$

for the exponential distribution.

#### **Improved dominance for Poisson distribution**

- Remember, the expected recourse cost of a route can be written

$$Q_r = 2\sum_{j=1}^t \sum_{l=1}^\infty \left( F^{j-1}(lQ) - F^j(lQ) \right) c_{0i_j}$$

with

$$F^{j}(x) = \mathbb{P}\left(\sum_{s=1}^{j} \xi_{p(s)} \le x\right)$$

## Improved dominance for Poisson distribution

• New notation: 
$$F^{poi}(x, \lambda)$$
 is the cumulative distribution function for Poisson distribution with parameter  $\lambda$ .

• We extend  $L_1$  by a partial route with total expected demand  $d + \Delta$  with  $\Delta$  being the difference in expected demand between last and second to last customer on extension. Probability of *l*'th failure at last customer of extension:

$$F^{poi}(lQ,\lambda_1+d) - F^{poi}(lQ,\lambda_1+d+\Delta)$$

Similar for label  $L_2$ :

$$F^{poi}(lQ, \lambda_2 + d) - F^{poi}(lQ, \lambda_2 + d + \Delta)$$

If we can show

 $F^{poi}(lQ,\lambda_1+d) - F^{poi}(lQ,\lambda_1+d+\Delta) \le F^{poi}(lQ,\lambda_2+d) - F^{poi}(lQ,\lambda_2+d+\Delta)$ 

for all  $l \in \mathbb{N}, d \geq 0, \Delta > 0, \lambda_2 + d + \Delta \leq Q$  then probability of failure for all feasible extensions of  $L_2$  is higher than the same extension of  $L_1$ .

# Improved dominance for Poisson distribution

- New notation: I son distribution
- We extend  $L_1$  b  $\Delta$  being the dif last customer or extension:

this follows from  $f(\lambda) = F^{poi}(x, \lambda)$  (x is a constant here ) being concave and nondecreasing in the interval ]0,x] on for Pois-

 $d + \Delta$  with d second to customer of

$$F^{por}(lQ,\lambda_1+d) - F^{por}(lQ,\lambda_1+d+\Delta)$$

Similar for label  $L_2$ :

$$F^{poi}(lQ, \lambda_2 + d) - F^{poi}(lQ, \lambda_2 + d + \Delta)$$

If we can show

$$F^{poi}(lQ,\lambda_1+d) - F^{poi}(lQ,\lambda_1+d+\Delta) \le F^{poi}(lQ,\lambda_2+d) - F^{poi}(lQ,\lambda_2+d+\Delta)$$

for all  $l \in \mathbb{N}$ ,  $d \ge 0$ ,  $\Delta > 0$ ,  $\lambda_2 + d + \Delta \le Q$  then probability of failure for all feasible extensions of  $L_2$  is higher than the same extension of  $L_1$ .

## Other improvements compared to Christiansen/Lysgaard approach



- Eliminating more cycles using ng-routes
- Adding valid inequalities known from the CVRP.
- Strong branching

#### **Results (all with Poisson distributed demands)**



	BCP			Christiansen & Lysgaard				
	Time (s)	Opt	Root LB	BB	Time (s)	Opt	LB root	BB
A-32-5	7	Х	843.80	853.60	282	Х	817.31	853.6
A-33-5	2	Х	702.64	704.20	8	Х	700.01	704.2
A-33-6	2	Х	792.81	793.90	49	Х	775	793.9
A-34-5	5	Х	822.78	826.87	-		803.26	825.26
A-36-5	9	Х	851.74	858.71	-		838.83	852.09
A-37-5	27	Х	700.06	708.34	-		687.4	707.54
A-37-6	6	Х	1025.66	1031.16	-		1007.98	1030.44
A-38-5	24	Х	768.30	775.14	-		739.19	761.12
A-39-5	1	Х	869.18	869.18	3	Х	866.92	869.18
A-39-6	9	Х	873.24	876.60	279	Х	850.09	876.6
A-44-6	52	Х	1019.01	1025.48	-		1007.55	1021.29
A-45-6	20	Х	1017.18	1026.73	-		984.38	1006.88
A-45-7	19	Х	1261.44	1264.83	882	Х	1254.23	1264.83
A-46-7	9	Х	999.47	1002.22	-		986.39	999.87
A-48-7	14	Х	1183.99	1187.14	-		1160.52	1180.22
A-53-7	123	Х	1112.89	1124.27	-		1093.64	1109.34
A-54-7	137	Х	1280.69	1287.07	-		1262.49	1279.93
A-55-9	15	Х	1175.88	1179.11	-		1148.4	1173.56

BCP. TL 7200 sec, Intel Xeon 2.66Ghz

C&L. TL 1200 sec, Pentium Centrino 1.5Ghz. Guestimate: 3 times slower than Xeon.

#### **Results (all with Poisson distributed demands)**



	BCP			Christiansen & Lysgaard				
	Time (s)	Opt	Root LB	BB	Time (s)	Opt	LB root	BB
A-60-9	7215	-	1508.45	1522.83	-		1489.82	1503.65
E-22-4	0	Х	411.57	411.57	1	Х	409.86	411.57
E-33-4	5	Х	850.27	850.27	86	Х	844.35	850.27
E-51-5	1298	Х	542.12	553.26	-		538.75	544.63
P-16-8	0	Х	512.819	512.819	0	Х	511.27	512.82
P-19-2	6	Х	222.564	224.062	153	Х	210.9	224.06
P-20-2	46	Х	225.279	233.054	352	Х	221.11	233.05
P-21-2	0	Х	218.962	218.962	5	Х	217.75	218.96
P-22-2	27	Х	225.757	231.341	219	Х	223.67	231.26
P-22-8	0	Х	681.06	681.06	0	Х	677.97	681.06
P-23-8	0	Х	619.527	619.527	1	Х	619.52	619.52
P-40-5	2	Х	472.497	472.497	9	Х	471.47	472.5
P-45-5	203	Х	525.948	533.522	-		519.03	527.77
P-50-7	17	Х	578.419	582.371	-		573.66	581.19
P-50-8	40	Х	663.973	669.23	-		659.67	666.9
P-50-10	21	Х	752.082	758.764	-		750.27	756.52
P-51-10	6	Х	807.477	809.7	430	Х	802.58	809.7
P-55-7	73	Х	585.134	588.563	-		582.12	585.47
P-55-10	42	Х	736.556	742.41	-		734.69	740.02
P-55-15	10	Х	1064.57	1068.92	792	Х	1062.67	1068.05
P-60-10	220	Х	796.201	803.604	-		793.71	798.63
P-60-15	8	Х	1083.9	1085.49	_	Х	1080.85	1085.12

BCP. TL 7200 sec, Intel Xeon 2.66Ghz

C&L. TL 1200 sec, Pentium Centrino 1.5Ghz. Guestimate: 3 times slower than Xeon.

#### **More tests**

- Tested on 91 instances based on standard CVRP instances.
- 69 could be solved to optimality within 2 hours.
- Largest solved instance: 100 customers.
- Smallest unsolved instance: 33 customers.

### Conclusion

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- Significant improvement compared to existing branch-and-price algorithm.
- Approach is promising at least for some distributions
- Future work (to name a few directions)
- Compare to integer L-shaped methods (branch and cut).
- Check if improved dominance work for more distributions
- Understand why some small instances are difficult (poor lower bounds).
- Understand the role of each improvement suggested.