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**BENDERS'
 DECOMPOSITION**

of the capacitated plant location problem

author

- Which plant(s) should be built?
 "location"
- Which customers should be supplied by each plant?
 "allocation"

The Problem

Given: a set of **N** demand points, with
 D_j = annual demand of customer #j
 a set of **M** potential plant sites, with
 S_i = annual capacity of plant #i (if built)
 F_i = annual fixed cost of building & operating plant #i
 C_{ij} = unit cost of production at plant #i, plus cost of shipping to customer #j

Define the variables:

X_{ij} = annual quantity shipped from plant #i to customer #j
 $Y_i = \begin{cases} 1 & \text{if a plant is built at site \#i} \\ 0 & \text{otherwise} \end{cases}$

The mathematical model

Minimize $\sum_{i=1}^M F_i Y_i + \sum_{i=1}^M \sum_{j=1}^N C_{ij} X_{ij}$

subject to:

$\sum_{j=1}^N X_{ij} \leq S_i Y_i$ for all i ← if no plant is built at site #i, the total shipments from site #i must be zero!

$\sum_{i=1}^M X_{ij} \geq D_j$ for all j

$X_{ij} \geq 0, Y_i \in \{0,1\}$ for all i,j

Notice that if we had selected values for each variable Y_i , the problem of selecting X_{ij} is the classical **transportation problem!**

Define an optimal value function of this transportation problem:

$V(Y) = \sum_{i=1}^M F_i Y_i + \text{minimum} \sum_{i=1}^M \sum_{j=1}^N C_{ij} X_{ij}$

s.t.

$\sum_{j=1}^N X_{ij} \leq S_i Y_i$ for all i

$\sum_{i=1}^M X_{ij} \geq D_j$ for all j

$X_{ij} \geq 0$

That is, given a value for each Y_i , indicating whether a plant is to be built there, you can then solve a transportation problem to determine the quantities to be shipped from each of the plants to each customer.

The total annual fixed cost of the plants, plus the optimal transportation costs, is the value of the function V at the point Y .

Our original problem is therefore equivalent to

$$\text{Minimize } v(Y)$$

Unfortunately, the function V is difficult to characterize!

By Linear Programming duality theory, the optimal value of the transportation problem is equal to that of its dual LP:

$$v(Y) = \sum_{i=1}^M F_i Y_i + \text{maximum} \sum_{i=1}^M S_i Y_i u_i + \sum_{j=1}^N D_j v_j$$

s.t.

$$u_i + v_j \leq C_{ij} \quad \forall i \& j$$

$$u_i \geq 0 \quad v_j \geq 0$$

Suppose that all the basic solutions of the dual LP are enumerated, with (\hat{u}^k, \hat{v}^k) denoting basic solution number k . Then $v(Y)$ might be computed by evaluating the dual objective at each extreme point, and selecting that producing the largest value:

$$v(Y) = \sum_{i=1}^M F_i Y_i + \text{maximum}_k \left\{ \sum_{i=1}^M S_i Y_i \hat{u}_i^k + \sum_{j=1}^N D_j \hat{v}_j^k \right\}$$

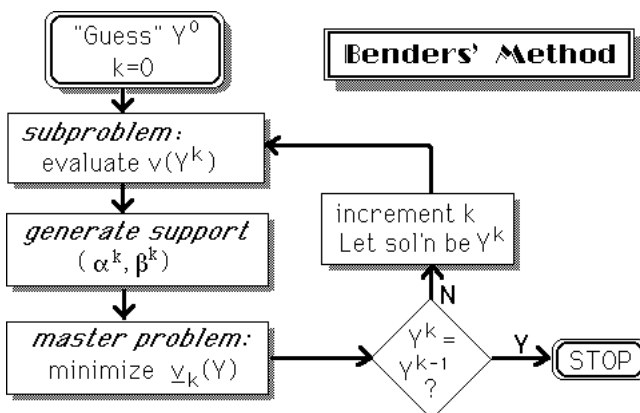
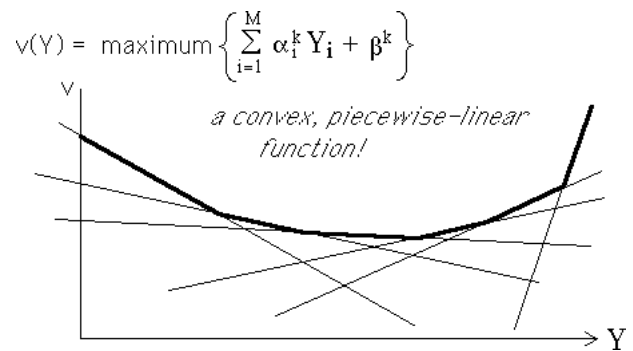
Define, for each dual basic solution (\hat{u}^k, \hat{v}^k) ,

$$\alpha_i^k = F_i + S_i \hat{u}_i^k$$

$$\beta^k = \sum_{j=1}^N D_j \hat{v}_j^k$$

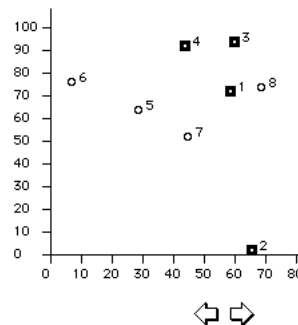
so that $v(Y) = \text{maximum} \left\{ \sum_{i=1}^M \alpha_i^k Y_i + \beta^k \right\}$

Thus, $v(Y)$ is the maximum of a large number of linear functions of Y .



Example

Plant Location Problem



4 plant sites,
8 demand points
(4 demand points are also potential plant sites)

■ plant site & demand pt
○ demand pt only

Random Problem (Seed = 94294)

Number of sources = M = 4
 Number of destinations = N = 8
 Total demand: 29

Costs, Supplies, Demands

i \ j	1	2	3	4	5	6	7	8	K	F
1	0	70	22	25	31	52	24	10	13	300
2	70	0	92	93	72	95	54	72	13	400
3	22	92	0	16	43	56	45	22	10	250
4	25	93	16	0	32	40	40	31	9	200
Demand	4	2	10	5	1	1	5	1	45	

K = capacity,
F = fixed cost



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To initiate the search, we "guess" that all the plants are opened, i.e.,

$$Y_i = 1 \text{ for } i=1,2,3,4$$

The first step is then to solve the subproblem to evaluate $V(1,1,1,1)$, i.e., the transportation problem with all four plants opened.



Next, we must solve the (partial) master problem, namely

$$\text{Minimize } \underline{v}_1(Y) \\ Y_i \in \{0,1\}$$

where

$$\underline{v}_1(Y) = 300 Y_1 + 400 Y_2 + 250 Y_3 + 200 Y_4 + 201$$



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(A constraint might have been added to the master problem which would guarantee that only feasible sets of plants were selected.... that is,

$$\sum_{i=1}^M S_i Y_i \geq \sum_{j=1}^N D_j,$$

but in this case no such constraint was used.)



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Options

Optimizing the Master Problem

Each Master Problem minimizes $v(Y)$, requiring a complete search of the enumeration tree.

Suboptimizing the Master Problem

A solution Y with $v(Y) < \text{incumbent}$ is found by the Master Problem; only one "pass" through the enumeration tree is required.



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Subproblem Solution

Plants opened: # 1 2 3 4
 Minimum transport cost = 201
 Fixed cost of plants = 1150
 Total = 1351
 CPU time = 9.05 sec.
 Generated support is $\alpha Y + b$, where
 $\alpha = 300 \ 400 \ 250 \ 200$
 $\& b = 201$
 That is, $v(Y) \geq \alpha Y + b$
 This is support # 1

*** New incumbent! ***

The cost of (1,1,1,1) is 1351, our initial "incumbent"



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Master Problem

Open: # , estimated cost: 201

Optimum of Master Problem

Optimal set of plants: <empty>
 with estimated cost 201
 CPU time: 0.55 sec.

Because the approximating function $\underline{v}_1(Y)$ is such a poor approximation, the solution to the master problem is to open NO plants!



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Subproblem Solution

Plants opened: # (none)
 Minimum transport cost = 290000
 Fixed cost of plants = 0
 Total = 290000
 CPU time = 14.45 sec.
 Generated support is $\alpha Y + b$, where
 $\alpha = -129700 \ -129600 \ -99750 \ -89800$
 $\& b = 290000$
 That is, $v(Y) \geq \alpha Y + b$
 This is support # 2

A "dummy" source with very large "shipping" costs was included, to guarantee feasibility.



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Master Problem

Open: # 1 2 3, estimated cost: 1151
 Open: # 1 2 4, estimated cost: 1101
 Open: # 1 3 4, estimated cost: 951

Optimum of Master Problem

Optimal set of plants: 1 3 4
 with estimated cost 951

CPU time: 4.8 sec.

y_2 is minimized at
 $Y=(1, 0, 1, 1)$



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Subproblem Solution

Plants opened: # 1 3 4

*** New incumbent! ***
(replaces 1351)

Minimum transport cost = 341
 Fixed cost of plants = 750
 Total = 1091

CPU time = 11.2 sec.

Generated support is $\alpha Y + b$, where
 $\alpha = 1210 \ 400 \ 790 \ 830$
 $\& b = -1739$

That is, $v(Y) \geq \alpha Y + b$

This is support # 3

$y_2(1,0,1,1) = 951 < 1091 = v(1,0,1,1)$



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Master Problem

Open: # 1 3 4, estimated cost: 1091
 Open: # 2 3 4, estimated cost: 1051

Optimum of Master Problem

Optimal set of plants: 2 3 4
 with estimated cost 1051

CPU time: 4.8 sec.

Minimum of y_3 is 1051, at
 $Y = (0, 1, 1, 1)$



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Subproblem Solution

Plants opened: # 2 3 4

Minimum transport cost = 599
 Fixed cost of plants = 850
 Total = 1449

CPU time = 21.75 sec.

While the estimated cost of
 $Y=(0,1,1,1)$ was lower than the
incumbent's cost, its actual
cost is considerably higher!

$y_3(0,1,1,1) = 1051 < 1449 = v(0,1,1,1)$



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Master Problem

Open: # 1 3 4, estimated cost: 1091

Optimum of Master Problem

Optimal set of plants: 1 3 4 *← the incumbent*
 with estimated cost 1091

CPU time: 4.85 sec.

$v(Y) = \underline{y}(Y)$

termination criterion is satisfied!

The Y which minimizes
 $y_4(Y)$ happens to be the
incumbent!



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Current List of Supports of $v(Y)$

Current approximation of $v(Y)$ is $\text{Maximum} \{ \alpha[i]Y + b[i] \}$
 where α & b are:

α_1	α_2	α_3	α_4	β
300	400	250	200	201
-129700	-129600	-99750	-89800	290000
1210	400	790	830	-1739
300	1310	340	425	-626

Current incumbent: 1091



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Suboptimizing the Master Problem

Again, we begin with the "guess"

$Y = (1,1,1,1)$,

i.e., that all four plants are open.



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Initial "guess":
all plants open

Subproblem Solution

Plants opened: # 1 2 3 4

*** New incumbent! ***

Minimum transport cost = 201
 Fixed cost of plants = 1150
 Total = 1351

CPU time = 9.05 sec.

Generated support is $\alpha Y + b$, where
 $\alpha = 300 \ 400 \ 250 \ 200$
 $\& b = 201$

That is, $v(Y) \geq \alpha Y + b$

This is support # 1

Initial subproblem



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Solution of 1st subproblem

Optimal Shipments

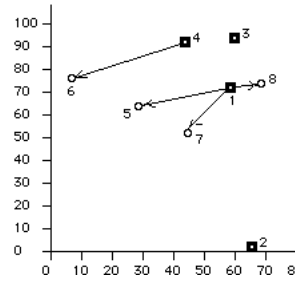
to	1	2	3	4	5	6	7	8	9	
from	1	4	0	0	0	1	0	5	1	2
	2	0	2	0	0	0	0	0	0	11
	3	0	0	10	0	0	0	0	0	0
	4	0	0	0	5	0	1	0	0	3

(Demand pt #9 is dummy demand for excess capacity.)
NOTE: Solution is degenerate!



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Optimal shipments (to non-local customers)



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Dual Solution of Transportation Problem

Supply constraints

i=	1	2	3	4
U[i]	0	0	0	0

Demand constraints

j=	1	2	3	4	5	6	7	8
V[j]	0	0	0	0	31	40	24	10

Reduced costs: COST - U^o.+V

	0	70	22	25	0	12	0	0
70	0	92	93	41	55	30	62	
22	92	0	16	12	16	21	12	
25	93	16	0	1	0	16	21	



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$$\alpha_i^k = F_i + S_i \hat{u}_i^k$$

$$\beta^k = \sum_{j=1}^N D_j \hat{v}_j^k$$

Generating the first support for v(Y)

Supply constraints

i=	1	2	3	4
U[i]	0	0	0	0
S[i]	13	13	10	9
F[i]	300	400	250	200

Demand constraints

j=	1	2	3	4	5	6	7	8
V[j]	0	0	0	0	31	40	24	10
D[j]	4	2	10	5	1	1	5	1

$$\alpha_i^0 = F_i \Rightarrow \alpha^0 = (300, 400, 250, 200)$$

$$\beta^0 = 31 + 40 + 120 + 10 = 201$$



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Current List of Supports of v(Y)

Current approximation of v(Y) is Maximum { $\alpha[i]Y + b[i]$ } where α & b are:

α_1	α_2	α_3	α_4	β
300	400	250	200	201

Current incumbent: 1351

$$\Rightarrow \underline{v}_1(Y) = 300Y_1 + 400Y_2 + 250Y_3 + 200Y_4 + 201$$



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First master problem solution

Master Problem

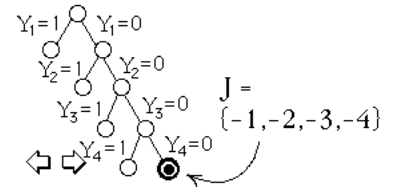
(suboptimized, i.e., a solution Y such that v(Y) < incumbent.)

Trial set of plants : <empty> with estimated cost 201 < incumbent (= 1351)

Current status vectors for Balas' additive algorithm:

\underline{j} : $\underline{0} \quad \underline{-1} \quad \underline{-2} \quad \underline{-3} \quad \underline{-4}$
underline: 0 0 0 0

CPU time: 0.55 sec.



1997

Subproblem Solution

Plants opened: # (none)

Minimum transport cost = 290000
Fixed cost of plants = 0
Total = 290000

CPU time = 14.45 sec.

Generated support is $\alpha Y + b$, where
 $\alpha = -129700 \quad -129600 \quad -99750 \quad -89800$
& $b = 290000$

That is, $v(Y) \geq \alpha Y + b$

This is support # 2

(all demand is supplied from dummy plant with high shipping cost, 10000/unit)



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Master Problem

(suboptimized, i.e., a solution Y such that v(Y) < incumbent.)

Trial set of plants: 2 3 4 with estimated cost 1051 < incumbent (= 1351)

Current status vectors for Balas' additive algorithm:

\underline{j} : $\underline{1} \quad \underline{2} \quad \underline{3} \quad \underline{4}$
underline: 0 1 0 0

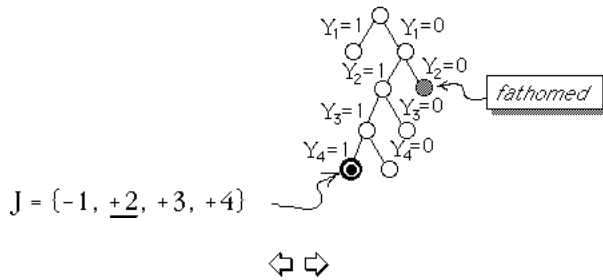
CPU time: 1.6 sec.

$$J = [-1, +2, +3, +4]$$



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The status of the search tree is currently:



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Subproblem Solution

Plants opened: # 2 3 4
 Minimum transport cost = 599
 Fixed cost of plants = 850
 Total = 1449
 CPU time = 21.8 sec.
 Generated support is $\alpha Y + b$, where
 $\alpha = 300 \ 1310 \ 340 \ 425$
 $\& b = -626$
 That is, $v(Y) \geq \alpha Y + b$
 This is support # 3



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Master Problem

(suboptimized, i.e., a solution Y such that $v(Y) < \text{incumbent}$.)

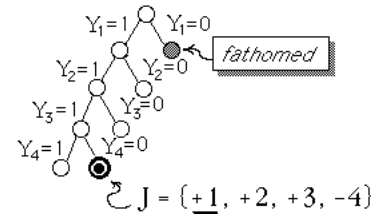
Trial set of plants: 1 2 3
 with estimated cost 1324 < incumbent (= 1351)
 Current status vectors for Balas' additive algorithm:
 $j: \quad 1 \ 2 \ 3 \ -4$
 underline: 1 0 0 0
 CPU time: 1.7 sec.

$J = [+1, +2, +3, -4]$



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The status of the search tree is currently:



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Subproblem Solution

Plants opened: # 1 2 3
 Minimum transport cost = 458
 Fixed cost of plants = 950
 Total = 1408
 CPU time = 16.3 sec.
 Generated support is $\alpha Y + b$, where
 $\alpha = 625 \ 1115 \ 280 \ 200$
 $\& b = -612$
 That is, $v(Y) \geq \alpha Y + b$
 This is support # 4



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Optimal Shipments

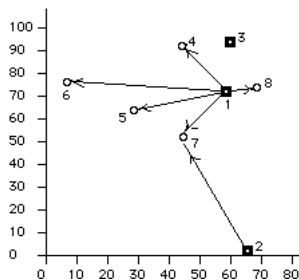
	to	1	2	3	4	5	6	7	8	9
from	1	4	0	0	5	1	1	1	0	0
	2	0	2	0	0	0	0	4	0	7
	3	0	0	10	0	0	0	0	0	0

(Demand pt #9 is dummy demand for excess capacity.)
 NOTE: Solution is degenerate!



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Optimal shipments (to non-local customers)



Plant #3 serves only the local customer at that location

Customer #7 is supplied by two different plants!



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Dual Solution of Transportation Problem

Supply constraints					Demand constraints								
U_i	$i=1$	2	3	4	V_j	$j=1$	2	3	4	5	6	7	8
	25	55	3	0		-25	-55	-3	0	6	27	-1	-15

Reduced costs: $COST - U_i + V_j$

0	100	0	0	0	0	0	0
40	0	40	38	11	13	0	32
44	144	0	13	34	26	43	34
50	148	19	0	26	13	41	46



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Master Problem

(suboptimized, i.e., a solution Y such that $v(Y) < \text{incumbent}$.)

Trial set of plants: 1 3 4
 with estimated cost 951 < incumbent (= 1351)
 Current status vectors for Balas' additive algorithm:
 j: 1 2 3 4
 underline: 1 1 0 0
 CPU time: 2.3 sec.

$$J = \{+1, -2, +3, +4\}$$



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Subproblem Solution

Plants opened: # 1 3 4
 Minimum transport cost = 341
 Fixed cost of plants = 750
 Total = 1091
 CPU time = 11.2 sec.
 Generated support is $\alpha Y + b$, where
 $\alpha = 1210 \ 400 \ 790 \ 830$
 $\& b = -1739$
 That is, $v(Y) \geq \alpha Y + b$
 This is support # 5



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When the master problem was optimized at each iteration, a total of FOUR subproblems were necessary, while we required FIVE subproblems when we suboptimized the master problem...

One more subproblem was required than in the algorithm which the Master Problem was optimized at each iteration!

However, the savings in computation in solving the master problem more than compensates for the additional subproblem!



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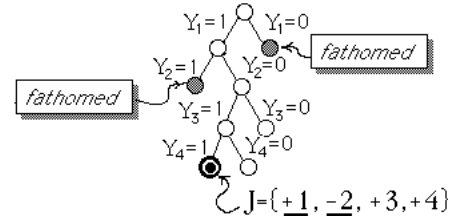
The approximation $v_5(Y)$ which we have computed is useful in answering "what-if" questions, e.g.,

"Although it is optimal to open plants at locations #1, 3, and 4, what if we were to open a plant at location 2 instead of location 3, i.e., is there a large penalty for choosing location 2 instead of 3?"



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The status of the search tree is currently:



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Master Problem

*** No solution with $v(Y)$ less than incumbent! ***
 (Current incumbent: 1091, with plants #1 3 4 open)
 CPU time: 0.75 sec.

The search tree has been completely enumerated!

One more subproblem was required than in the algorithm which the Master Problem was optimized at each iteration!



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Current List of Supports of $v(Y)$

Current approximation of $v(Y)$ is $\text{Maximum} \{ \alpha[i]Y + b[i] \}$ where α & b are:

α_1	α_2	α_3	α_4	β
300	400	250	200	201
-129700	-129600	-99750	-89800	290000
300	1310	340	425	-626
625	1115	280	200	-612
1210	400	790	830	-1739

Current incumbent: 1091



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Evaluation of (approximation of) $v(Y)$

Open plants: 1 2 4

support	value
1	1101
2	-59100
3	1409
4	1328
5	701

Maximum value of the five supports at $Y=(1,1,0,1)$ is 1409, so we know that the cost would be increased by at least $1409-1091=318$

*** Maximum value, namely 1409 is approximation (underestimate) of $v(Y)$ (Note: incumbent is 1091)

trial solution



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Subproblem Solution

Plants opened: # 1 2 4

Minimum transport cost = 553
Fixed cost of plants = 900
Total = 1453

> 1409 (approximation)

CPU time = 16.25 sec.

Generated support is $\alpha Y + b$, where

$\alpha = 586 \ 1076 \ 250 \ 344$

$\& \ b = -553$

That is, $v(Y) \geq \alpha Y + b$

This is support # 6

In actuality, the cost is
increased by $1453 - 1091 = 361$

