

# Stochastic Transportation Problem with Recourse

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## Stochastic Transportation Problem (without Simple Recourse):

Consider the following problem:

- A *single* product is stored in various quantities in a network of  $n$  nodes.
- *Before* demand for the product occurs, the product may be moved from one node to another at a known cost  $C^1$ .
- *After* demand becomes known, there is still an opportunity to move the product between nodes, but at a greater cost.
- Product at a node in excess of demand has a *salvage value*, and product sold earns a revenue.

### Notation

- $S_i$  = initial supply available at node  $i$ ,  $i = 1, \dots, n$
- $C_{ij}^1$  = (*first-stage*) transportation cost from node  $i$  to node  $j$  *before* demand is known
- $C_{ij}^2$  = (*second-stage*) transportation cost from node  $i$  to node  $j$  *after* demand is known
- $V_i$  = salvage value of product at node  $i$ ,  $i = 1, \dots, n$
- $U_i$  = penalty per unit shortage at node  $i$ ,  $i = 1, \dots, n$
- $D_i$  = demand at node  $i$ ,  $i = 1, \dots, n$ ,  
which is *random* with discrete distribution:  
$$p_i^k \equiv P\{D_i = d_i^k\}, k = 1, \dots, K_i$$

We wish to formulate the optimization problem:

to determine the shipment plan for the product  
*before* the demand becomes known  
in order to minimize the sum of  
the first-stage shipment costs  
and the *expected* cost of the second-stage  
(i.e., shipment costs  
& shortage penalties,  
minus salvage values).

Define the decision variables:

**Stage 1:**

$X_{ij}$  = shipment in first stage from node  $i$  to node  $j$   
(if  $i = j$ , the amount retained at node  $i$ )

**Stage 2:**

$Y_{ij}$  = shipment in second stage from node  $i$  to node  $j$   
 $Z_i^+$  = quantity in excess of demand at node  $i$  after all shipments  
 $Z_i^-$  = shortage at node  $i$  after all shipments

**Stochastic LP model:**

$$\text{Minimize } \sum_{i=1}^n \sum_{j=1}^n C_{ij}^1 X_{ij} + E_D \{Q(X, D)\}$$

$$\text{subject to } \sum_{j=1}^n X_{ij} = S_i \quad \forall i$$

$$X_{ij} \geq 0 \quad \forall i \& j$$

where  $Q(X, D)$  is the minimum cost of the second stage:

$$Q(X, D) = \text{Min } \sum_{i=1}^n \sum_{j=1}^n C_{ij}^2 Y_{ij} + \sum_{i=1}^n (U_i Z_i^- - V_i Z_i^+)$$

$$\text{s.t. } \sum_{j=1}^n Y_{ij} - \sum_{k=1}^n Y_{ki} + Z_i^- + Z_i^+ = D_i - \sum_{k=1}^n X_{ki} \quad \forall i$$

$$Y_{ij} \geq 0, Z_i^+ \geq 0, Z_i^- \geq 0 \quad \forall i \& j$$

Example data:

$n = 3$  nodes

	Node 1	Node 2	Node 3
Supply $S_i$	2	4	14
Shortage penalty $U_i$	8	9	8
Salvage value $V_i$	1	4	5

**Shipping costs:**

Before demand occurs:

	Node 1	Node 2	Node 3
Node 1	0	2	3
Node 2	2	0	3
Node 3	3	2	0

After demand occurs:

	Node 1	Node 2	Node 3
Node 1	0	6	10
Node 2	6	0	15
Node 3	12	15	0

**Demand distributions:**

**Node 1:**

demand	4	6	8
P{demand}	0.4	0.4	0.2

**Node 2:**

demand	6	8
P{demand}	0.4	0.6

**Node 3:**

demand	6	8
P{demand}	0.5	0.5

In the notation used for the general 2-stage stochastic LP with recourse, identify the arrays

- $c=$
- $A=$
- $b=$
- $T=$
- $q=$
- $W=$
- $h=$

Which of these arrays are random?

Stochastic Transportation Problem, non-Simple Recourse

First-stage data:

$A, B=$   
 $1 \ 1 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 = 2$   
 $0 \ 0 \ 0 \ 1 \ 1 \ 1 \ 0 \ 0 \ 0 = 4$   
 $0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 1 \ 1 = 14$

i	variable	cost
1	X11	0
2	X12	2
3	X13	3
4	X21	2
5	X22	0
6	X23	2
7	X31	3
8	X32	2
9	X33	0

Objective: Minimize

Second-stage data

$K=$  # scenarios = 12

The following data vary by scenario:  $h$

Costs:

i	Variable	q	
1	Y12	6	
2	Y13	10	
3	Y21	6	
4	Y23	15	
5	Y31	12	
6	Y32	15	
7	EX1	-4	excess
8	EX2	-4	excess
9	EX3	-2	excess
10	SH1	15	shortage
11	SH2	20	shortage
12	SH3	30	shortage

Technology matrix  $T$

(coefficients of  $X$  in 2nd stage) =

$1 \ 0 \ 0 \ 1 \ 0 \ 0 \ 1 \ 0 \ 0$   
 $0 \ 1 \ 0 \ 0 \ 1 \ 0 \ 0 \ 1 \ 0$   
 $0 \ 0 \ 1 \ 0 \ 0 \ 1 \ 0 \ 0 \ 1$

Technology matrix  $W$

(coefficients of  $Y$  in 2nd stage) =

$-1 \ -1 \ 1 \ 0 \ 1 \ 0 \ -1 \ 0 \ 0 \ 1 \ 0 \ 0$   
 $1 \ 0 \ -1 \ -1 \ 0 \ 1 \ 0 \ -1 \ 0 \ 0 \ 1 \ 0$   
 $0 \ 1 \ 0 \ 1 \ -1 \ -1 \ 0 \ 0 \ -1 \ 0 \ 0 \ 1$

Right-hand-sides in second stage =

k	p[k]	1	2	3	k	p[k]	1	2	3
1	0.08	4	6	6	7	0.12	6	8	6
2	0.08	4	6	8	8	0.12	6	8	8
3	0.12	4	8	6	9	0.04	8	6	6
4	0.12	4	8	8	10	0.04	8	6	8
5	0.08	6	6	6	11	0.06	8	8	6
6	0.08	6	6	8	12	0.06	8	8	8

Benders Decomposition Algorithm

multi-cut version

Iteration #1

Trial X for primal subproblems  
(found by minimizing first-stage cost alone)

is

i	Variable	Value
1	X11	2
5	X22	4
9	X33	14

That is, our initial "guess" is to make no first-stage shipments between nodes.

Using the trial first-stage solution, we solve the second-stage problem for each scenario:

Scenario #1 with probability 0.08	i	variable	value	
	5	Y31	2	
Optimal objective: 46	6	Y32	4	
	i	variable	value	
	5	Y31	2	
	6	Y32	2	
	9	EX3	4	
	-----			
Scenario #2 w/ probability 0.08	i	variable	value	
Optimal objective: 50	5	Y31	2	
	6	Y32	2	
	9	EX3	2	
	-----			
Scenario #3 w/ probability 0.12	i	variable	value	
Optimal objective: 80	5	Y31	2	
	6	Y32	4	
	9	EX3	2	
	-----			
Scenario #4 w/ probability 0.12	i	variable	value	
Optimal objective: 84	5	Y31	2	
	6	Y32	2	
	9	EX3	2	
	-----			
Scenario #5 w/ probability 0.08	i	variable	value	
Optimal objective: 74	5	Y31	4	
	6	Y32	2	
	9	EX3	2	
	-----			
Scenario #6 w/ probability 0.08	i	variable	value	
Optimal objective: 78	5	Y31	4	
	6	Y32	2	
	9	EX3	2	
	-----			

Scenario #7 w/ probability 0.12	Scenario #10 w/ probability 0.04
Optimal objective: 108	Optimal objective: 108
i variable value	i variable value
5 Y31 4	5 Y31 4
6 Y32 4	6 Y32 2
	10 SH1 2
	-----
Scenario #8 w/ probability 0.12	Scenario #11 w/ probability 0.06
Optimal objective: 114	Optimal objective: 138
i variable value	i variable value
5 Y31 2	5 Y31 4
6 Y32 4	6 Y32 4
10 SH1 2	10 SH1 2
	-----
Scenario #9 w/ probability 0.04	Scenario #12 w/ probability 0.06
Optimal objective: 102	Optimal objective: 144
i variable value	i variable value
5 Y31 6	5 Y31 2
6 Y32 2	6 Y32 4
	10 SH1 4

Primal subproblems summary  
Second stage costs:

k	cost	p[k]
1	46	0.08
2	50	0.08
3	80	0.12
4	84	0.12
5	74	0.08
6	78	0.08
7	108	0.12
8	114	0.12
9	102	0.04
10	108	0.04
11	138	0.06
12	144	0.06

First stage cost:	0.00
Expected second stage cost:	91.48
Total:	91.48

This is an **upper** bound on the optimal solution!

For each scenario, we also obtain the simplex multipliers (shadow prices, dual variables, etc.) for the three subproblem constraints:

Lagrangian multipliers

i	1	2	3
1	14	17	2
2	14	17	2
3	14	17	2
4	15	18	3
5	14	17	2
6	15	18	3
7	15	18	3
8	15	18	3
9	15	18	3
10	15	18	3
11	15	18	3
12	15	18	3
Sum	176	212	32

These are used to generate a “cut” or linear function which is an approximation (under-estimate) of  $Q_k(X)$ , the second-stage minimum cost when scenario #k occurs.

The “master problem” then minimizes the sum of the first-stage costs plus the expected value of the (approximation) of second-stage costs:

Solution of Master Problem

i	variable	value
2	X12	2
5	X22	4
8	X32	14

Now node 1 ships all of its supply to node 2!

First-stage cost: 32

k	Q[k]	p[k]
1	-170	0.08
2	-166	0.08
3	-136	0.12
4	-132	0.12
5	-142	0.08
6	-138	0.08
7	-108	0.12
8	-102	0.12
9	-114	0.04
10	-108	0.04
11	-78	0.06
12	-72	0.06

These are the underestimates of the second-stage costs for each scenario, provided by the single cut for each.

Total (estimated) expected value: -92.52

Trial X for primal subproblems is

i	Variable	Value
2	X12	2
5	X22	4
8	X32	14

**Iteration #2**

Using this new “trial” first-stage solution, each scenario is evaluated by solving its subproblem:

Scenario #1 w/ probability 0.08

Optimal objective: 98

i	variable	value
3	Y21	4
4	Y23	6
8	EX2	4

Scenario #3 w/ probability 0.12

Optimal objective: 106

i	variable	value
3	Y21	4
4	Y23	6
8	EX2	2

Scenario #2 w/ probability 0.08

Optimal objective: 136

i	variable	value
3	Y21	4
4	Y23	8
8	EX2	2

Scenario #4 w/ probability 0.12

Optimal objective: 144

i	variable	value
3	Y21	4
4	Y23	8

Scenario #5 w/ probability 0.08

Optimal objective: 118

i	variable	value
3	Y21	6
4	Y23	6
8	EX2	2

Scenario #9 w/ probability 0.04

Optimal objective: 138

i	variable	value
3	Y21	8
4	Y23	6

Scenario #6 w/ probability 0.08

Optimal objective: 156

i	variable	value
3	Y21	6
4	Y23	8

Scenario #10 w/ probability 0.04

Optimal objective: 186

i	variable	value
3	Y21	6
4	Y23	8
10	SH1	2

Scenario #7 w/ probability 0.12

Optimal objective: 126

i	variable	value
3	Y21	6
4	Y23	6

Scenario #11 w/ probability 0.06

Optimal objective: 156

i	variable	value
3	Y21	6
4	Y23	6
10	SH1	2

Scenario #8 w/ probability 0.12

Optimal objective: 174

i	variable	value
3	Y21	4
4	Y23	8
10	SH1	2

Scenario #12 w/ probability 0.06

Optimal objective: 204

i	variable	value
3	Y21	4
4	Y23	8
10	SH1	4

Primal subproblems summary

Second stage costs:

k	cost	p[k]
1	98	0.08
2	136	0.08
3	106	0.12
4	144	0.12
5	118	0.08
6	156	0.08
7	126	0.12
8	174	0.12
9	138	0.04
10	186	0.04
11	156	0.06
12	204	0.06

First stage cost: 32.00  
 Expected second stage cost: 141.20  
 Total: 173.20

*This is another upper bound on the optimal expected cost (but not as good as the earlier upper bound!)*

*Again, we obtain the dual variables of the three constraints from each subproblem solution:*

Lagrangian multipliers

i	1	2	3
1	10	4	19
2	10	4	19
3	10	4	19
4	15	9	24
5	10	4	19
6	15	9	24
7	15	9	24
8	15	9	24
9	15	9	24
10	15	9	24
11	15	9	24
12	15	9	24
Sum	160	88	268

These are the *third* set of such dual variables for each scenario, so each  $Q_k$  will now be approximated by the maximum of three linear “cuts” or supports.

*The master problem is now solved again:*

Solution of Master Problem

i	variable	value
1	X11	2.0
5	X22	4.0
8	X32	5.4
9	X33	8.6

First-stage cost: 10.8

k	[k]	p[k]
1	-35	0.08
2	-5	0.08
3	-1	0.12
4	3	0.12
5	-7	0.08
6	15	0.08
7	27	0.12
8	33	0.12
9	21	0.04
10	45	0.04
11	57	0.06
12	63	0.06

Total (estimated) expected value: 25.52

*This is an **under-estimate** of the minimum expected cost!*

*...etc.*

Trial X for primal subproblems is

i	Variable	Value
1	X11	2
5	X22	4
7	X31	2
8	X32	4
9	X33	8

**Iteration #5**

*The master problem in iteration #4 yields the “trial” solution: from node 3, ship 2 units to node 1 and 4 to node 2.*

Scenario #1 w/ probability 0.08

Optimal objective: -12

i	variable	value
8	EX2	2
9	EX3	2

Scenario #2 w/ probability 0.08

Optimal objective: -8

i	variable	value
8	EX2	2

Scenario #3 w/ probability 0.12

Optimal objective: -4

i	variable	value
9	EX3	2

-----  
 Scenario #4 w/ probability 0.12  
 Optimal objective: 0  
 i variable value

-----  
 Scenario #5 w/ probability 0.08  
 Optimal objective: 8  
 i variable value  
 3 Y21 2  
 9 EX3 2

-----  
 Scenario #6 w/ probability 0.08  
 Optimal objective: 12  
 i variable value  
 3 Y21 2

-----  
 Scenario #7 w/ probability 0.12  
 Optimal objective: 24  
 i variable value  
 5 Y31 2

-----  
 Scenario #8 w/ probability 0.12  
 Optimal objective: 30  
 i variable value  
 10 SH1 2

-----  
 Scenario #9 w/ probability 0.04  
 Optimal objective: 36  
 i variable value  
 3 Y21 2  
 5 Y31 2

-----  
 Scenario #10 w/ probability 0.04  
 Optimal objective: 42  
 i variable value  
 3 Y21 2  
 10 SH1 2

-----  
 Scenario #11 w/ probability 0.06  
 Optimal objective: 54  
 i variable value  
 5 Y31 2  
 10 SH1 2

-----  
 Scenario #12 w/ probability 0.06  
 Optimal objective: 60  
 i variable value  
 10 SH1 4

Cuts for scenario(s) 3 4 6 7 9 10 11 12 were generated previously!

--Primal subproblems summary  
 Second stage costs:

k	cost	p[k]
1	-12	0.08
2	-8	0.08
3	-4	0.12
4	0	0.12
5	8	0.08
6	12	0.08
7	24	0.12
8	30	0.12
9	36	0.04
10	42	0.04
11	54	0.06
12	60	0.06

First stage cost: 14.00  
 Expected second stage cost: 15.96  
 Total: 29.96 *a new upper bound!*

Lagrangian multipliers

i	1	2	3
1	4	4	2
2	4	4	14
3	4	10	2
4	14	20	24
5	10	4	2
6	15	9	24
7	15	18	3
8	15	20	25
9	15	9	3
10	15	9	24
11	15	18	3
12	15	20	25
Sum	141	145	151

Solution of Master Problem

Optimal value= 29.96

i	variable	value
1	X11	2
5	X22	4
7	X31	2
8	X32	4
9	X33	8

Total (estimated) expected value: 29.96

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Converged at iteration #5!

X was generated by previous master problem!

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Optimality Cuts (Multi-cut version)

For each scenario k, a set of linear functions of X has been generated, each of which is an underestimate of the second-stage cost as a function of X.

Hence the maximum provides an underestimating piecewise-linear function of X, each of the form  $\text{Max} \{ (X)(\text{Lambda}[k]) + \text{Alpha}[k] \}$ .

Scenario 1

Cut	Lambda									Alpha	
1	-14	-17	-2	-14	-17	-2	-14	-17	-2	170	
2	-10	-4	-19	-10	-4	-19	-10	-4	-19	178	
3		-10	-4	-2	-10	-4	-2	-10	-4	76	
4		-4	-10	-2	-4	-10	-2	-4	-10	88	
5			-4	-4	-2	-4	-4	-2	-4	-2	52

Scenario 2

Cut	Lambda									Alpha	
1	-14	-17	-2	-14	-17	-2	-14	-17	-2	174	
2	-10	-4	-19	-10	-4	-19	-10	-4	-19	216	
3		-10	-4	-2	-10	-4	-2	-10	-4	80	
4	-4	-10	-14	-4	-10	-14	-4	-10	-14	188	
5		-4	-4	-14	-4	-4	-14	-4	-4	-14	152

Scenario 3

1		-15	-18	-3	-15	-18	-3	-15	-18	-3	222
2		-15	-9	-24	-15	-9	-24	-15	-9	-24	336
3			-15	-9	-3	-15	-9	-3	-15	-9	168
4	-14	-20	-24	-14	-20	-24	-14	-20	-24	396	
5		-15	-9	-24	-15	-9	-24	-15	-9	-24	336

Scenario 7

Cut	Lambda									Alpha
1	-15	-18	-3	-15	-18	-3	-15	-18	-3	252
2	-15	-9	-24	-15	-9	-24	-15	-9	-24	306
3		-15	-9	-3	-15	-9	-3	-15	-9	180
4	-14	-20	-5	-14	-20	-5	-14	-20	-5	274
5	-15	-18	-3	-15	-18	-3	-15	-18	-3	252

Scenario 8

Cut	Lambda									Alpha
1	-15	-18	-3	-15	-18	-3	-15	-18	-3	258
2	-15	-9	-24	-15	-9	-24	-15	-9	-24	354
3		-15	-9	-3	-15	-9	-3	-15	-9	186
4	-14	-20	-24	-14	-20	-24	-14	-20	-24	436
5	-15	-20	-25	-15	-20	-25	-15	-20	-25	450

Scenario 9

Cut	Lambda									Alpha
1	-15	-18	-3	-15	-18	-3	-15	-18	-3	246

Cut	Lambda									Alpha	
1	-14	-17	-2	-14	-17	-2	-14	-17	-2	204	
2	-10	-4	-19	-10	-4	-19	-10	-4	-19	186	
3		-14	-8	-2	-14	-8	-2	-14	-8	132	
4	-4	-10	-2	-4	-10	-2	-4	-10	-2	108	
5		-4	-10	-2	-4	-10	-2	-4	-10	-2	108

Scenario 4

Cut	Lambda									Alpha
1	-15	-18	-3	-15	-18	-3	-15	-18	-3	228
2	-15	-9	-24	-15	-9	-24	-15	-9	-24	324
3		-15	-9	-3	-15	-9	-3	-15	-9	156
4	-14	-20	-24	-14	-20	-24	-14	-20	-24	408
5	-14	-20	-24	-14	-20	-24	-14	-20	-24	408

Scenario 5

Cut	Lambda									Alpha	
1	-14	-17	-2	-14	-17	-2	-14	-17	-2	198	
2	-10	-4	-19	-10	-4	-19	-10	-4	-19	198	
3		-14	-8	-2	-14	-8	-2	-14	-8	144	
4	-4	-10	-2	-4	-10	-2	-4	-10	-2	96	
5		-10	-4	-2	-10	-4	-2	-10	-4	-2	96

Scenario 6

Cut	Lambda									Alpha
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2	-15	-9	-24	-15	-9	-24	-15	-9	-24	318
3	-15	-9	-3	-15	-9	-3	-15	-9	-3	192
4	-14	-20	-5	-14	-20	-5	-14	-20	-5	262
5	-15	-9	-3	-15	-9	-3	-15	-9	-3	192

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Scenario 10

<u>Cut</u>	<u>Lambda</u>									<u>Alpha</u>
1	-15	-18	-3	-15	-18	-3	-15	-18	-3	252
2	-15	-9	-24	-15	-9	-24	-15	-9	-24	366
3	-15	-9	-3	-15	-9	-3	-15	-9	-3	198
4	-15	-20	-25	-15	-20	-25	-15	-20	-25	440
5	-15	-9	-24	-15	-9	-24	-15	-9	-24	366

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Scenario 11

<u>Cut</u>	<u>Lambda</u>									<u>Alpha</u>
1	-15	-18	-3	-15	-18	-3	-15	-18	-3	282
2	-15	-9	-24	-15	-9	-24	-15	-9	-24	336
3	-15	-9	-3	-15	-9	-3	-15	-9	-3	210
4	-14	-20	-5	-14	-20	-5	-14	-20	-5	302
5	-15	-18	-3	-15	-18	-3	-15	-18	-3	282

-----  
Scenario 12

<u>Cut</u>	<u>Lambda</u>									<u>Alpha</u>
1	-15	-18	-3	-15	-18	-3	-15	-18	-3	288
2	-15	-9	-24	-15	-9	-24	-15	-9	-24	384

3	-15	-9	-3	-15	-9	-3	-15	-9	-3	216
4	-15	-20	-25	-15	-20	-25	-15	-20	-25	480
5	-15	-20	-25	-15	-20	-25	-15	-20	-25	480

Best Solution of Benders

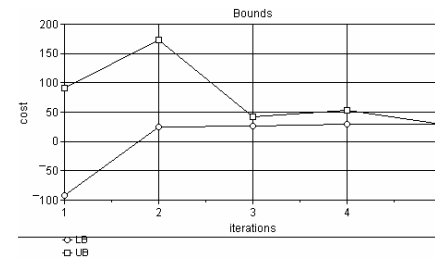
-----  
Total cost: 29.96, found at iteration #5  
Best lower bound: 29.96

Gap= -3.5527137E-15, or -1.185819E-14%

Non-zero Stage One Variables:

<u>i</u>	<u>variable</u>	<u>value</u>
1	X11	2
5	X22	4
7	X31	2
8	X32	4
9	X33	8

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Certainty-Equivalent Tableau

Solving the LP problem assuming the demands are certain to be their expected values:

b	z	1	2	3	4	5	6	7	8	9	1	2	3	4	5	6	7	8	9	10	11	12
0	1	0	2	3	2	0	2	3	2	0	6	10	6	15	12	15	-4	-4	-2	15	20	30
2	0	1	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	1	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
14	0	0	0	0	0	0	0	1	1	1	0	0	0	0	0	0	0	0	0	0	0	0
5.6	0	1	0	0	1	0	0	1	0	0	-1	-1	1	0	1	0	-1	0	0	1	0	0
7.2	0	0	1	0	0	1	0	0	1	0	1	0	-1	-1	0	1	0	-1	0	0	1	0
7	0	0	0	1	0	0	1	0	0	1	0	1	-1	-1	0	0	-1	0	0	1	0	1

Optimal Solution

Found by solving **certainty equivalent problem**, i.e., replacing all random parameters by their expected values.

Total objective function: 16.8

Stage One: nonzero variables:

i	variable	value
1	X11	2.0
5	X22	4.0
7	X31	3.6
8	X32	3.4
9	X33	7.0

Second Stage: nonzero variables

i	variable	value
8	EX2	0.2

The optimal cost (\$16.80) of this LP is not the *actual* expected cost of this first-stage shipment plan!

To evaluate the expected cost, we must solve each scenario's subproblem, using this shipment plan as "trial" solution.

Evaluation of trial solution

i	variable	X[i]
1	X11	2.0
5	X22	4.0
7	X31	3.6
8	X32	3.4
9	X33	7.0

Summary

Second stage objective:

k	objective	p[k]
1	-14.0	0.08
2	2.0	0.08
3	-2.4	0.12
4	13.6	0.12
5	-3.6	0.08
6	17.4	0.08
7	13.8	0.12
8	43.0	0.12
9	20.4	0.04
10	47.4	0.04
11	43.8	0.06
12	73.0	0.06

First stage objective:	17.60
Expected second stage objective:	<u>18.02</u>
Total:	35.62

If we were to use this shipment plan, our expected cost would exceed the minimum (\$26.96) by

$$\$35.62 - 26.96 = \$ 8.66 = \text{VSS (value of stochastic solution) !}$$

Suppose that we had "perfect information" about the demands before choosing the first-stage variables:

Optimization with perfect information

Solution for scenario #1			Solution for scenario #2		
Optimal cost: 2			Optimal cost: 6		
Stage One: nonzero variables:			Stage One: nonzero variables:		
<u>i</u>	<u>value</u>	<u>Name</u>	<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11	1	2.00	X11
5	4.00	X22	5	4.00	X22
7	2.00	X31	7	2.00	X31
8	6.00	X32	8	4.00	X32
9	6.00	X33	9	8.00	X33
Second-stage: nonzero variables			Second-stage: nonzero variables		
<u>i</u>	<u>value</u>	<u>Name</u>	<u>i</u>	<u>value</u>	<u>Name</u>
8	4.00	EX2	8	2.00	EX2

Solution for scenario #3

Optimal cost: 10  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	2.00	X31
8	6.00	X32
9	6.00	X33

Second-stage: nonzero variables

<u>i</u>	<u>value</u>	<u>Name</u>
8	2.00	EX2

Second-stage: nonzero variables (none)

Solution for scenario #5

Optimal cost: 12  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	4.00	X31
8	4.00	X32
9	6.00	X33

Second-stage: nonzero variables

<u>i</u>	<u>value</u>	<u>Name</u>
8	2.00	EX2

Solution for scenario #4

Optimal cost: 14  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	2.00	X31
8	4.00	X32
9	8.00	X33

Solution for scenario #6

Optimal cost: 16  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	4.00	X31
8	2.00	X32
9	8.00	X33

Second-stage: nonzero variables: (none)

Solution for scenario #8

Optimal cost: 44  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	2.00	X31
8	4.00	X32
9	8.00	X33

Second-stage: nonzero variables

Solution for scenario #7

Optimal cost: 20  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	4.00	X31
8	4.00	X32
9	6.00	X33

Second-stage: nonzero variables

<u>i</u>	<u>value</u>	<u>Name</u>
10	2.00	SH1

Solution for scenario #9

Optimal cost: 22  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	6.00	X31
8	2.00	X32
9	6.00	X33

Second-stage: nonzero variables

<u>i</u>	<u>value</u>	<u>Name</u>
10	2.00	SH1

Solution for scenario #11

Optimal cost: 50  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	4.00	X31
8	4.00	X32
9	6.00	X33

Second-stage: nonzero variables: (none)

Solution for scenario #10

Optimal cost: 46  
 Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	4.00	X31
8	2.00	X32
9	8.00	X33

Second-stage: nonzero variables

<u>i</u>	<u>value</u>	<u>Name</u>
10	2.00	SH1

Solution for scenario #12

Second-stage: nonzero variables

Optimal cost: 74

<u>i</u>	<u>value</u>	<u>Name</u>
10	4.00	SH1

Stage One: nonzero variables:

<u>i</u>	<u>value</u>	<u>Name</u>
1	2.00	X11
5	4.00	X22
7	2.00	X31
8	4.00	X32
9	8.00	X33

Expected cost with perfect information: 23.6

The **EVPI** (*expected value of perfect information*) is therefore the difference between the expected value *with* perfect information and the expected value *without* such information, i.e.,

$$\text{EVPI} = \$26.96 - \$23.60 = \$6.36$$