

Advanced linear programming

Stochastic programming
Benders decomposition

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Two stage stochastic linear programming example: farmer's problems

	Wheat	Corn	Sugar Beets
Yield (T/acre)	2.5	3	20
Planting cost (\$/acre)	150	230	260
Selling price (\$/T)	170	150	36 under 6000 T 10 above 6000 T
Purchase price (\$/T)	238	210	–
Minimum requirement (T)	200	240	–
Total available land: 500 acres			



Optimal solution

Culture	Wheat	Corn	Sugar Beets
Surface (acres)	120	80	300
Yield (T)	300	240	6000
Sales (T)	100	—	6000
Purchase (T)	—	—	—
Overall profit: \$118,600			



Scenarios

- Harvest yield very sensitive to weather:
 - Yield 20% more in good year, 20% reduced in bad year
- Suppose type of year would be known beforehand:
 - Good year: 167.667
 - Average year: 118.600
 - Bad year: 59.950 (buy corn)
 - Average: 115.406

- This is not realistic!!!!



Two-stage model

- First stage: land assignment
- Second stage: sales and purchase
- Maximize expected profit

- Solution value 108.390

- Expected value of perfect information:
 - $115.406 - 108.390 = 7.016$



Two-stage stochastic linear programming problem

- Scenario's: $1, 2, \dots, K$
- α_ω : probability scenario s occurs
- First stage decision variables: x
- Second stage decision variables: y_ω ($\omega=1, 2, \dots, K$)

- Minimize expected cost:

$$cx + \alpha_1 fy_1 + \alpha_2 fy_2 + \dots + \alpha_K fy_K$$



Two stage stochastic linear programming problem

$$\min cx + \alpha_1 fy_1 + \alpha_2 fy_2 + \dots + \alpha_K fy_K$$

s.t

$$\begin{array}{rcccc} Ax & & & = b \\ B_1x + & Dy_1 & & = d_1 \\ B_2x + & & Dy_2 & = d_2 \\ \vdots & & \ddots & \vdots \\ B_kx + & & & Dy_K = d_K \end{array}$$

$$x, y_1, \dots, y_K \geq 0$$



Benders decomposition sub problem

■ For fixed x :

$$P_{\omega} : z_{\omega}(x) = \min f y_{\omega}$$

$$\text{subject to } B_{\omega}x + D y_{\omega} = d_{\omega}$$

$$y_{\omega} \geq 0$$

■ Equivalent to:

$$P_{\omega} : z_{\omega}(x) = \min f y_{\omega}$$

$$\text{subject to } D y_{\omega} = d_{\omega} - B_{\omega}x$$

$$y_{\omega} \geq 0$$



Benders decomposition

■ Dual of P_ω :

$$\begin{aligned} \max \quad & p_\omega (d_\omega - B_\omega x) \\ \text{s.t.} \quad & p_\omega D \leq f \end{aligned}$$

$$P = \{p \mid pD \leq f\}$$

p^i ($i \in I$) extreme points

w^j ($j \in J$) extreme rays

■ We have

$$w^j (d_\omega - B_\omega x) \leq 0 \quad \forall j \quad (\text{required for primal feasibility})$$

$$z_\omega = \max_{i \in I} p^i (d_\omega - B_\omega x)$$

$$\text{Hence } p^i (d_\omega - B_\omega x) \leq z_\omega \quad \forall i$$



Benders decomposition: master problem

$$\min cx + \sum_{\omega=1}^K \alpha_{\omega} z_{\omega}$$

s.t.

$$Ax = b$$

$$p^i (d_{\omega} - B_{\omega}x) \leq z_{\omega} \quad \forall i, \omega$$

$$w^j (d_{\omega} - B_{\omega}x) \leq 0 \quad \forall j, \omega$$

$$x \geq 0$$

Number of constraints increased
Number of variables has decreased
Cutting plane algorithm!



Benders decomposition: initial relaxed master problem

$$\min cx + \sum_{\omega=1}^K \alpha_{\omega} z_{\omega}$$

s.t.

$$Ax = b$$

$$x \geq 0$$



Benders decomposition: Solution algorithm

1. Solve relaxed master problem: x^*, z_1^*, \dots, z_K^*
2. For all ω do
 1. Solve sub problem by dual simplex

$$\min f y_\omega \quad s.t. \quad D y_\omega = (d_\omega - B_\omega x^*), \quad y_\omega \geq 0.$$

2. If optimum $> z_\omega^*$ add

$p^i (d_\omega - B_\omega x) \leq z_\omega$, where p^i optimal extreme point in dual

3. If sub problem infeasible, then dual has infinite solution add

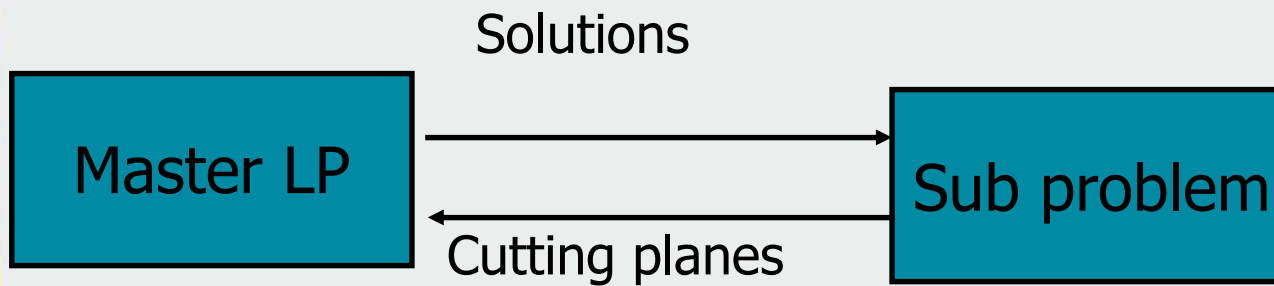
$$w^j (d_\omega - B_\omega x) \leq 0, \text{ where } w^j \text{ extreme ray in dual}$$

which gives infinite solution value

4. If feasible and optimum $\leq z_\omega^*$ do nothing
3. If feasible and optimum $\leq z_\omega^*$ for all ω , problem solved to optimality, otherwise solve relaxed problem again (go to step 1)



Benders decomposition: use cutting plane algorithm



Example: facility location, just one scenario

Depot i	1	2	3
Fixed $c F_i$	2	3	3
customers	C_{ij}		
1	2	4	5
2	3	3	4
3	4	1	2
4	5	2	1
5	7	6	3

v_j and w_{ij} first round

Cust j	v_j	w_{1j}	w_{2j}	w_{3j}
1	2	0	0	0
2	3	0	0	0
3	4	0	3	2
4	5	0	3	4
5	7	0	1	4

