


```
In [4]: try:
# instantiate model object
m = gp.Model("Farmer_problem2")

#####
##### Parameters set up #####
#####
c = 3 # number of crop lots
S = 3 # number of possible scenarios
L = len(feedCrops) # amount of feed crops
A = 500
display(data_df)

# Probabilities for he below, average, and above average cases
scenario_probs = [
    1/3, # below average yield probability---
    1/3, # average yield probability
    1/3, # above average yield probability
]

crop_yieldRate_scenario = [1, 1, 1, 1, 2] # below average, average, above average
crop_price_scenario = [1, 1, 1, .8, 1] # below average, average, above average

#####
##### Variables set up #####
#####
Xc = m.addVars(C, vtype=GRB.CONTINUOUS, name="X", lb=0) # acreage per crop for planting
Bac = m.addVars(B, C, vtype=GRB.CONTINUOUS, name="B", lb=0) # expected yield for each crop & scenario
Yac = m.addVars(S, C, vtype=GRB.CONTINUOUS, name="Y", lb=0) # amount to purchase for feed crops
Cac = m.addVars(S, F, vtype=GRB.CONTINUOUS, name="C", lb=0) # purchase costs for feed crops & scenario
Wac = m.addVars(S, C, vtype=GRB.CONTINUOUS, name="W", lb=0) # amount to sell per crop

G = m.addVars(S, vtype=GRB.CONTINUOUS, name="G", lb=0) # planting costs
H = m.addVars(S, vtype=GRB.CONTINUOUS, name="H", lb=0) # profits for sales in each scenario for each crop
Ra = m.addVars(S, vtype=GRB.CONTINUOUS, name="R", lb=0) # purchase costs for each scenario for feed crops

crops = data_df.columns.to_list()

#####
##### Objective set up #####
#####
# set up expected yields for each scenario
m.addConstrs(Bac[c], c) == data_df.loc["Yield", crops[c]]*crop_yieldRate_scenario[s]*Xc[c]
for s in range(S) for c in range(C)

# set up planting costs (G)
exp = 0
for c in range(C):
    print(data_df.loc["Planting Cost", crops[c]])
    exps = Xc[c]*data_df.loc["Planting Cost", crops[c]]
    m.addConstr(G[0]==exps)

# maximum acreage constraint
tot_acres = 0
for c in range(C):
    tot_acres += Xc[c]
m.addConstr(tot_acres <= A)

# ##### Feed Crop Constraints
# set up purchasing cost for different scenarios
for s in range(S) for c in range(C) for i in range(S) for c in range(F)

# set up minimum feed crop constraints
m.addConstrs(Wac[s, c] - Wac[s, c] >= data_df.loc["min", crops[c]]
for s in range(S) for c in range(F)

# set up purchasing costs expressions for each scenario
for s in range(S):
    exp = 0
    for c in range(F):
        exp += Wac[s, c]*data_df.loc["Price", crops[c]]*crop_price_scenario[s]
    m.addConstr(R[s]==exp)

# set up sales profits expressions
for s in range(S):
    exp = 0
    for c in range(C):
        exp += Wac[s, c]*data_df.loc["Selling Cost1", crops[c]]*crop_price_scenario[s]
    exp += Wac[s, 2]*data_df.loc["Selling Cost1", crops[2]] \
        + Wac[s, 3]*data_df.loc["Selling Cost2", crops[2]]
    m.addConstr(H[s]==exp)

# ##### Sugar Beet Constraints
# set up sugar beet amounts for different levels
m.addConstrs(Bac[s, 2] - Wac[s, 2] - Wac[s, 3]) >= 0 for s in range(S)
m.addConstr(Bac[s, 2] <= quota for s in range(S)) # set up lower level of quota sales constraint

# set up expressions for profits minus purchase costs for each scenario
below_average_money_flow = (H[0] - Ra[0])*(1/3)
average_money_flow = (H[1] - Ra[1])*(1/3)
above_average_money_flow = (H[2] - Ra[2])*(1/3)

##### Objective #####
m.setObjective(below_average_money_flow + average_money_flow + above_average_money_flow \
- G[0], GRB.MAXIMIZE)

##### SOLVE:OPTIMIZE #####
m.optimize()

##### Display Results #####
displayDecisionVars(m, end_sentence="6")

print("\n-----Does it make sense?-----")
print('Obj: ',(1.2E1).format(m.ObjVal))

# catch some math errors
except up.GurobiError as e:
    print("Error code " + str(e.erno) + " : " + str(e))

except AttributeError:
    print("Encountered an attribute error")
```

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	Wheat	Corn	Sugar Beets
Yield	25	3	20
Planting Cost	1500	230	260
Selling Cost1	1700	150	36
Selling Cost2	1700	150	10
Price	2380	210	0
min	2000	240	0

```
150.0
230
260
Gurobi Optimizer version 9.5.0 build v9.5.0rc5 (win64)
Thread count: 6 physical cores, 12 logical processors, using up to 12 threads
Optimize a model with 35 rows, 43 columns and 85 nonzeros
Model fingerprint: 0x2db09eb0
Coefficient statistics:
  Matrix range [1e+00, 3e+02]
  Objective range [3e+01, 1e+00]
  Bounds range [1e+00, 1e+00]
  RHS range [2e+02, 6e+03]
Presolve removed 28 rows and 28 columns
Presolve time: 0.00s
Presolved: 7 rows, 15 columns, 21 nonzeros

Iteration    Objective          Primal Inf.    Dual Inf.        Time
   0         4.621333e+33      6.000000e+30  4.621333e+03      0s
  10        1.0685067e+05      0.000000e+00  0.000000e+00      0s

Solved in 10 iterations and 0.01 seconds (0.00 work units)
Optimal objective 1.06850667e+05
X(0) 170.00000
X(1) 80.00000
X(2) 250.00000
E(0,0) 340.00000
E(0,1) 192.00000
E(0,2) 4000.00000
E(1,0) 425.00000
E(1,1) 240.00000
E(1,2) 5000.00000
E(2,0) 510.00000
E(2,1) 288.00000
E(2,2) 6000.00000
Y(0,0) 0.00000
Y(0,1) 48.00000
Y(1,0) 0.00000
Y(1,1) 0.00000
Y(2,0) 0.00000
Y(2,1) 0.00000
Y(2,2) 261.40000
C(0,0) 231.00000
C(1,0) 238.00000
C(1,1) 210.00000
C(2,0) 214.20000
C(2,1) 189.00000
W(0,0) 140.00000
W(0,1) 0.00000
W(0,2) 4000.00000
W(1,0) 0.00000
W(1,1) 225.00000
W(1,2) 0.00000
W(2,0) 5000.00000
W(2,1) 0.00000
W(2,2) 310.00000
R(0,0) 46.00000
R(0,1) 6000.00000
R(0,2) 0.00000
R(1,0) 108900.00000
R(1,1) 170180.00000
R(1,2) 218250.00000
R(2,0) 269910.00000
R(2,1) 11088.00000
R(2,2) 0.00000
R(2,3) 0.00000
-----
Obj: 106850.67
```

Solution Discussion: Problem 2

From the results with Gurobi, the optimal lot sizes(X) for wheat, corn, and sugar beets are 170, 80, and 250 acres, respectively. This lot-sizing leads to a profit of \$106850.67.

The below-average scenario yields 340 tons of wheat, 192 tons of corn, and 4000 tons of sugar beets. Because there are less than 6KT of sugar beets, it will be sold at the higher sales price for this scenario. Due to the corn crop generating less than 240 tons of corn for feed, the farmer purchases (Y(0,1)) 48 tons of corn. Thus, this scenario leads to an expected purchasing cost (R(0)) of \$11088.00. From sales of excess (W(0,1)) 140 tons of corn, there is a total profit of (H(0)) of \$170180.00.

The average scenario for this lot-sizing leads to yields of 425, 240, 5000 tons of wheat, corn, and sugar beets, respectively. Therefore, there are 225 tons of wheat and 5K tons of sugar beets to sell for this scenario. There are no excess tons of corn, but there is just enough to meet the feed crop requirements, and so no crops need be purchased by the farmer in the average case. This case leads to no purchases necessary (Y(1,1)) thus no purchasing costs (R(1)) and only sales profits(H(1)) of \$218250.00

And finally, in the above-average scenario, there will be 510, 288, and 6K tons of wheat, corn, and sugar beets, respectively. These yields lead to an excess of 310 tons of wheat, 48 tons of corn, and 6K tons of sugar beets to sell. This lot allotment generates a sales profit (H(2)) of \$269910.00. There is enough of each feed crop to meet the minimum requirements; therefore, there is no need to purchase any additional crops, leading to zero purchasing costs (R(2)).

Problem 3:

Model Formulation | Objective | Gurobi Implementation | Solution

1. Consider a division of the problem in which the farmer possesses four fields of sizes 185, 145, 105, and 65, respectively. The total of 500 acres is unchanged. However, the farmer wishes to only plant one type of crop on each field. Formulate and solve a stochastic program and solve it with a solver

Model Formulation Problem 3

Indices, Parameters, and Sets 3:

- C set of crops $\{1, 2, 3, 4\}$ $c \in C$
- F set of possible planting fields, $\{1, 2, 3, 4\}$ $f \in F$
- $D_{f,c}$ binary variable, 1 if field f will have crop c planted, 0 otherwise
- $X_{f,c}$ acres allocated to crop c , in field f , $c \in \{1, 2, 3\}$
- Y_c tons purchased for crop $c \in \{1, 2\}$
- E_c Expected yield (tons) for crop c in tons
- W_c tons of crop c sold
- C_c purchase cost for crop $c \in \{1, 2\}$
- L_c planting cost (\$/acre) for crop $c \in C$
- P_c sale price for crop c
- μ_c minimum amount of feed crop $c \in \{1, 2\}$
- A_f maximum amount of acreage available for field f
- ρ_c average expected ton/acre yield rate for crop c
- G Total Planting costs
- R Total Purchasing costs for feed crops
- H Total sales profits from all excess crops
- M large number

Equations and Constraints

- Nonzero acreage constraint: $X_{f,c} \geq 0, \forall f, c$
- Nonzero acreage constraint: $X_{f,c} \leq M \cdot D_{f,c}, \forall f, c$
- Binary Planting Decision variables: $\sum_{c=1}^3 (D_{f,c}) \leq 1, \forall f$
- Maximum Acreage Available: $\sum_{f=1}^4 \sum_{c=1}^3 (X_{f,c}) \leq 500$
- Expected Yield for a given crop: $E_c = \sum_{f=1}^4 (\rho_c \cdot X_{f,c}), c \in \{1, 2, 3\}$
- Purchasing Costs for crop c : $C_c = E_c \cdot P_c$
- Total purchasing costs: $R = \sum_{c=1}^2 (C_c), c \in \{1, 2\}$
- Total Sales Profits costs: $H = \sum_{c=1}^4 W_c \cdot P_c$
- Total Planting Costs Expression: $G = \sum_{f=1}^4 \sum_{c=1}^3 (X_{f,c} \cdot L_c)$
- Minimum feed crop constraints: $Y_c + E_c - W_c \geq \mu_c, \forall c$ where $c \in \{\text{wheat}(1), \text{corn}(2)\}$
- Sugar Beets Sales constraint: $E_3 - \sum_{c=1}^3 W_c \geq 0, \forall s$
- Sugar Beet below quota value constraint: $W_3 \leq 6000, \forall s$

Objective 3:

Maximize: $H - R - G$

Gurobi Implementation and Solution Problem 3

```
In [5]: try:
# instantiate model object
m = gp.Model("Farmer_problem3")

#####
##### Parameters set up #####
#####
c = 3 # number of crop lots
S = 3 # number of possible scenarios
L = len(feedCrops) # amount of feed crops
A = 500
field_sizes = [185, 145, 105, 65]
display(data_df)

crop_yieldRate_scenario = [1, 1, 1, 2] # below average, average, above average
crop_price_scenario = [1, 1, 1, .9] # below average, average, above average

#####
##### Variables set up #####
#####
Xfc = m.addVars(L, C, vtype=GRB.CONTINUOUS, name="X", lb=0) # acreage per crop for planting
Dfc = m.addVars(L, C, vtype=GRB.BINARY, name="D", lb=0) # Decision Variable field to plant what crop
Bc = m.addVars(C, vtype=GRB.CONTINUOUS, name="B", lb=0) # expected yield for each crop & scenario
Yc = m.addVars(F, vtype=GRB.CONTINUOUS, name="Y", lb=0) # amount to purchase for feed crops
Wc = m.addVars(C, vtype=GRB.CONTINUOUS, name="W", lb=0) # amount to sell per crop
Gc = m.addVars(C, vtype=GRB.CONTINUOUS, name="G", lb=0) # planting costs
Hc = m.addVars(S, vtype=GRB.CONTINUOUS, name="H", lb=0) # profits for sales in each scenario for each crop
Rc = m.addVars(S, vtype=GRB.CONTINUOUS, name="R", lb=0) # purchase costs for each scenario for feed crops
M = 1e5 # large number is just the total lot size times a half

# get list of the crop names
crops = data_df.columns.to_list()

#####
##### Objective set up #####
#####
# set up expected yields for each crop over all fields
for c in range(C):
    tot_crop_c = 0
    for f in range(L):
        tot_crop_c += Xfc[f, c]
    # set this crops total Estimated yield to be the total acreage it was given time
    # the expected yield rate
    m.addConstr(Bc[c] == data_df.loc["Yield", crops[c]]*tot_crop_c)

# set up planting costs (G)
exps = 0
for c in range(L):
    for c in range(C):
        print(data_df.loc["Planting Cost", crops[c]])
        exps = Xfc[f, c]*data_df.loc["Planting Cost", crops[c]]
    m.addConstr(G[0]==exps)

# add big M constraint for decision variable usage
for each of the possible crop selections
for c in range(C):
    for f in range(L):
        m.addConstr(Xfc[f, c] <= M*Dfc[f, c])

# add singular lot to crop assignment constraint
for l in range(L):
    lot_size = 0
    for c in range(C):
        lot_size += Xfc[l, c]
    # for each lot sum all decision variables and make sure they do not go above 0
    for c in range(C):
        lot_size += Xfc[l, c]
    m.addConstr(lot_size <= field_sizes[l])

# maximum acreage constraint
tot_acres = 0
for l in range(L):
    for c in range(C):
        tot_acres += Xfc[l, c]
m.addConstr(tot_acres <= A)

# lot size constraints for each lot
for l in range(L):
    lot_size = 0
    # initial total lot size to zero
    for c in range(C):
        lot_size += Xfc[l, c]
    # add the constraint that the total acres allotted
    # can not exceed what is available
    m.addConstr(lot_size <= field_sizes[l])

# ##### Feed Crop Constraints
# set up purchasing cost for different scenarios

# set up minimum feed crop constraints
m.addConstrs(Wac[s, c] - Wac[s, c] >= data_df.loc["min", crops[c]]
for s in range(S) for c in range(F)

# set up purchasing costs expressions for each scenario
exp = 0
for c in range(F):
    exp += Wac[s]*data_df.loc["Price", crops[c]]
m.addConstr(R[0]==exp)

# set up the total profits expression for the feed crops
for s in range(S):
    exp = 0
    for c in range(C):
        exp += Wac[s]*data_df.loc["Selling Cost1", crops[c]]
    exp += Wac[s]*data_df.loc["Selling Cost2", crops[2]] \
        + Wac[s]*data_df.loc["Selling Cost2", crops[2]]
m.addConstr(H[0]==exp)

# ##### Sugar Beet Constraints
m.addConstr(Bc[2] - Wc[2] - Wc[3]) >= 0 # set up sugar beet amount constraint
m.addConstr(Bc[2] <= quota) # set up lower level of quota sales constraint

# get each scenario's purchasing costs and sales profits

#####
##### Constraint set up #####
#####
m.setObjective(H[0] - R[0] - G[0], GRB.MAXIMIZE)

##### SOLVE:OPTIMIZE #####
m.optimize()

##### Display Results #####
displayDecisionVars(m, end_sentence="10")

print("\n-----Does it make sense?-----")
print('Obj: ',(1.2E1).format(m.ObjVal))

# catch some math errors
except up.GurobiError as e:
    print("Error code " + str(e.erno) + " : " + str(e))

except AttributeError:
    print("Encountered an attribute error")
```

	Wheat	Corn	Sugar Beets
Yield	25	3	20
Planting Cost	1500	230	260
Selling Cost1	1700	150	36
Selling Cost2	1700	150	10
Price	2380	210	0
min	2000	240	0

```
150.0
230
260
Gurobi Optimizer version 9.5.0 build v9.5.0rc5 (win64)
Thread count: 6 physical cores, 12 logical processors, using up to 12 threads
Optimize a model with 31 rows, 36 columns and 106 nonzeros
Model fingerprint: 0x32a537
Variable types: 24 continuous, 12 integer (12 binary)
Coefficient statistics:
  Matrix range [1e+00, 1e+02]
  Objective range [1e+00, 1e+00]
  Bounds range [1e+00, 1e+00]
  RHS range [1e+00, 6e+03]
Found heuristic solution: -30400.00000
Presolve removed 12 rows and 14 columns
Presolve time: 0.00s
Presolved: 19 rows, 22 columns, 54 nonzeros
Variable types: 14 continuous, 8 integer (8 binary)

Root relaxation: objective 1.186000e+05, 8 iterations, 0.00 seconds (0.00 work units)

Nodes | Current Node | Objective Bounds | Work
Expl Unexpl | Obj Depth IntInf | Incumbent BestObj Gap | It/Node Time
   0   |   0 | 118600.000   | 0   | 2 -30400.0000 118600.000  49.9%   | -   | 0s
   1   |   0 |   65200.0000 118600.000  81.9%   | -   | 0s
   2   |   0 | 72000.000000 118600.000  64.7%   | -   | 0s
   3   |   0 | 99225.000000 118600.000  19.5%   | -   | 0s
   4   |   0 | 106425.0000 118600.000  11.4%   | -   | 0s
   5   |   0 | 3106425.000 118600.000  11.4%   | -   | 0s
   6   |   0 | 107975.0000 118600.000  9.84%   | -   | 0s
   7   |   0 | 114875.0000 118600.000  3.24%   | -   | 0s
   8   |   0 | 2114875.000 114875.000  0.00%   | -   | 0s

Cutting planes:
MIR: 1
Flow cover: 3

Explored 1 nodes (24 simplex iterations) in 0.02 seconds (0.00 work units)
Thread count was 12 (of 12 available processors)

Solution count 5: 114875 107975 106425 -30400

Optimal solution found (tolerance 1.0e-04)
Best objective 1.148750000000e+05, best bound 1.148750000000e+05, gap 0.0000%
X(0,0) 0.00000
X(0,1) 0.00000
X(0,2) 185.00000
X(1,0) 145.00000
X(1,1) 0.00000
X(1,2) 0.00000
X(2,0) 0.00000
X(2,1) 0.00000
X(2,2) 105.00000
X(3,0) 0.00000
X(3,1) 65.00000
X(3,2) 0.00000
X(3,3) 0.00000
D(0,0) 0.00000
D(0,1) 0.00000
D(0,2) 0.00000
D(1,0) 0.00000
D(1,1) 0.00000
D(1,2) 0.00000
D(2,0) 0.00000
D(2,1) 0.00000
D(2,2) 1.00000
D(3,0) 0.00000
D(3,1) 0.00000
D(3,2) 0.00000
E(0) 162.50000
E(1) 195.00000
E(2) 5800.00000
E(3) 5800.00000
Y(1) 45.00000
Y(2) 162.50000
W(0) 145.00000
W(1) 65.00000
W(2) 5800.00000
W(3) 0.00000
W(4) 112100.00000
H(0) 236425.00000
R(0) 9450.00000
-----
Obj: 114875.00
```

Solution 3 Exploration & Discussion

Solution

From the results with Gurobi, the optimal lot sizes(X) and field allocations are plant 145 acres of wheat in lot 2, plant 65 acres of corn in lot 4, plant 105 acres of sugar beets in lot 1 and 105 in lot 3. This lot-sizing and field allocation scheme leads to an optimal profit value of \$114875.

The cell above shows the solution in a condensed form and points out several things about the solution:

- The given solution will require the purchase of 65 tons of corn due to the yield being below the minimum requirement. This leads to an extra purchasing cost of 9450.
- Due to there being below the quota of sugar beets yielded they can all be sold at the higher price point.

acreage Decisions:

- Lot 1:
 - X[0,2] 185.00000 --> Sugar Beets
- Lot 2:
 - X[1,0] 145.00000 --> Wheat
- Lot 3:
 - X[2,1] 105.00000 --> Sugar beets
- Lot 4:
 - X[3,1] 65.00000 --> Corn
- Total acreage used: 185 + 145 + 105 + 65 = 500 = MAX Available

Planting Costs:

- G[0] = 112100.00000

Expected Yields:

- E[0] = 362.50000 --> no need to buy any wheat, excess of 362.5 - 200 = 162.5
- E[1] = 195.00000 --> need to buy 240-195=45 tons of corn
- E[2] = 5800.00000 --> have below the quota of sugar beets so do not need to sale any at reduced price

amounts of feed crops to purchase:

Y[0] = 0.00000 --> wheat Y[1] = 45.00000 --> corn (Matches above)

Purchasing costs:

- R[0] = 9450.00000 --> should be 45*210=9450, thus the answer checks out

Tons of Excess crops:

W[0] = 162.50 --> wheat W[1] 0.00000 --> Corn W[2] 5800.00000 --> Beets at highest price W[3] 0.00000 --> Beets at lowest price

Profits from crop sales:

- H[0] = 236425.00000
- have 162 excess wheat so should make = 162*170=27.625
- have 5800 beets to sell so should make = 5800*36=208800
- Total profits 27.625 + 208890 = 236,425

Objective Check

- Thus H - G - R = 236425 - 112100 - 9450 = 114875, which is what the given objective was thus at least the math works out.

```
In [ ]: # save the notebook as a pdf
# to_PDF(notebook_title)
```