

# **Red Brand Canners (RBC)**

## **Production Planning, Revenue Analysis & Optimization**

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**(DNSC 6307 Optimization I)**

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# 1. Model for Canned Tomato Products Production

To create the most optimal model we first define the objective function. Given that we want to increase the sum of profit contribution from each product we outline the objective function as follows:

## **Maximize:**

Contribution from Whole tomatoes + Contribution from Juice + Contribution from Paste

## **Contribution from Each Product:**

Profit Contribution from selling each 1000 lbs unit of products:

- a. Whole Tomatoes: \$247.67
- b. Tomato Juice: \$198
- c. Tomato Paste: \$222

Total Profit Contribution from each Product:

Profit Contribution From selling each 1000 lbs unit of product \* amount produced in 1000 lbs

- a. Whole Tomatoes Contribution:  $247.67 \times \text{amount whole produced in 1000lbs}$
- b. Tomato Juice Contribution:  $198 \times \text{amount juice produced in 1000lbs}$
- c. Tomato Paste Contribution:  $222 \times \text{amount paste produced in 1000lbs}$

## **Constraints:**

To maximize the objective above we need to satisfy the following constraints:

### **a. Demand Constraints:**

Do not exceed the demand for each product:

- Demand for Whole Tomatoes  $\leq 14000$
- Demand for Tomato Juice  $\leq 1000$
- Demand for Tomato Paste  $\leq 2000$

### **b. Supply Constraints:**

Do not exceed the supply constraints for each grade of tomato

- A grade used for paste + A grade used for Juice + A grade used for Whole  $\leq 600$
- B grade used for paste + B grade used for Juice + B grade used for Whole  $\leq 2400$

### **c. Quality Constraints:**

Ensure the quality of the produced product by making sure the mix of A and B produces the right net quality required by the product:

- Whole:  $9 \times (\text{A-grade used for whole}) + 5 \times (\text{B grade used for whole}) / \text{total used for whole} \geq 8$
- Juice:  $9 \times (\text{A grade used for juice}) + 5 \times (\text{B grade used for juice}) / \text{total used for juice} \geq 6$

In order to maximize the revenue given the constraints on quality, supply and demand it is recommended that they allocate the supply as follows in order to achieve the overall optimal profit contribution. Our optimization model (Appendix 1A,1B) recommends the distribution outlined in exhibit 1A for tomatoes used for the production of each product to maximize the total profit contribution.

Exhibit 1A: Breakdown of Canned Tomato Products Production					
	Amount (in 1000 lbs)				
Product	Grade A Tomatoes Used	Grade B Tomatoes Used	Total Product Produced	Total Demand for Product	Unmet Demand
Whole Tomatoes	525	175	700	14400	13700
Tomato Juice	75	225	300	1000	700
Tomato Paste	0	2000	2000	2000	0
Total	600	2400	3000		

Using the production plan above we can determine the product contribution for each product and the total as follows in exhibit 1B.

Exhibit 1B: Profit Contribution			
Product	Profit Contribution per 1000 lb	Amount Produced (in 1000 lbs)	Total Profit Contribution from Product
Whole Tomatoes	246.67	700	172669
Tomato Juice	198	300	59400
Tomato Paste	222	2000	444000
Total Profit Contribution			676067

Following the allocation and production plan our model has recommended the best possible value of the objective (i-e the overall profit contribution) is \$676,067.

## 2. Updating Model with Additional supply of A tomatoes

### 2.1 Decision to buy the additional A grade tomatoes at 25.5 cents/lb

With the additional availability of 80,000 lbs of grade A tomatoes we can add it to our model as a separate new supply with separate constraints and decide the allocation to each of the products as shown in appendix 2B. see if it increases the profit contribution while also adjusting the cost of the tomatoes we buy.

Exhibit 2A: Updated Supply			
	Amount of Fruit (1000 pounds)		
Supply	Grade A	Grade B	Total
Mixed Grade Tomatoes	600	2400	3000
Additional A Grade	80	0	80
Total	680	2400	3080

With the additional grade A tomatoes we updated our model (Appendix 2A), the recommended allocation (Appendix 2B) is outlined below in exhibit 2B:

Exhibit 2B: Breakdown of Production (With Additional Grade A Supply Available)						
	Amount (in 1000 lbs)					
Product	Previous Supply Grade A Tomatoes Used	Previous Supply Grade B Tomatoes Used	New Supply Grade A Tomatoes Used	Total Product Produced	Demand for Product	Unmet Demand
Whole Tomatoes	535	205	80	820	14400	13580

Tomato Juice	65	195	0	260	1000	740
Tomato Paste	0	2000	0	2000	2000	0
Total	600	2400	80	3080		

The model recommends that we should buy the additional grade A tomatoes costing 25.5 cents per pound as it improves our maximum profit contribution by an amount of \$1,280 as outlined below in exhibit 2C:

Exhibit 2C: New Profit Contribution (With additional grade A supply)			
Product	Profit Contribution (per 1000 lb)	Amount Produced (in 1000 lbs)	Net Profit Contribution from Product
Whole Tomatoes	\$246.67	820	\$202,267
Tomato Juice	\$198	260	\$51,480
Tomato Paste	\$222	2000	\$444,000
Cost Incurred From Additional Grade A Tomato Supply			(\$20,400)
Total Profit Contribution (With additional Grade A tomatoes)			\$677,347
Total Profit Contribution (Without additional Grade A tomatoes)			\$676,067
Difference in Total Profit Contribution			\$1,280

This shows that if we increase the availability of grade A tomatoes by 80,000 lbs and adjust for the incurred cost of the fruit, we still get a positive difference to the overall profit contribution after adjusting for the cost incurred so it is recommended that we buy the additional supply.

## 2.2 Allocation of the new grade A tomatoes

The best allocation for the new supply of grade A tomatoes to achieve maximum profit contribution as recommended by the new model (Appendix 2B) is that we should utilize all of the 80,000 lbs of new grade A tomatoes to produce whole tomatoes as shown in exhibit 2B.

## 2.3 Another Optimal allocation

Since both the existing supply of grade A tomatoes and the new supply of A tomatoes produce the same results (or have the same quality) there is another optimal allocation solution that results in the same profit contribution as solution in 2.3. The optimal solution (Appendix 2C) outlined in Exhibit 2D and 2E shows that if we maintain the total amount of whole and juice produced as before by changing the new A supply with exactly the same amount of old A supply between whole and juice we get the same profit contributions.

Exhibit 2D: Alternative Breakdown of Production (With Additional Grade A Supply Available)						
	Amount (in 1000 lbs)					
Product	Previous Supply Grade A Tomatoes Used	Previous Supply Grade B Tomatoes Used	New Supply Grade A Tomatoes Used	Total Product Produced	Demand for Product	Unmet Demand
Whole Tomatoes	600	205	15	820	14400	13580
Tomato Juice	0	195	65	260	1000	740
Tomato Paste	0	2000	0	2000	2000	0
Total	600	2400	80	3080		

  

Exhibit 2E: Alternative Optimal Solution Profit Contribution (With additional grade A supply)			
Product	Profit Contribution (per 1000 lb)	Amount Produced (in 1000 lbs)	Net Profit Contribution from Product

Whole Tomatoes	\$246.67	820	\$202,267
Tomato Juice	\$198	260	\$51,480
Tomato Paste	\$222	2000	\$444,000
Cost Incurred From Additional Grade A Tomato Supply			(\$20,400)
Total Profit Contribution (With additional Grade A tomatoes)			\$677,347
Total Profit Contribution (Without additional Grade A tomatoes)			\$676,067
Difference in Total Profit Contribution			\$1,280

## 2.4 Maximum Price for Additional Tomatoes

In sensitivity analysis for this model (Appendix 2B) we can see that the shadow price for the 80,000 lbs of new grade A supply we buy 16, meaning in the dual problem of minimizing our price to purchase this supply has room to be increased by \$16 per 1000 lb to maintain the same profit contribution as when we didn't have the new supply available. Meaning that we can pay up to a maximum of  $\$255 + \$16 = \$271$  per 1000 lb for the additional supply. If we change the cost of the additional supply of grade tomatoes in the model we can get the same profit as before the new supply (shown in Appendix 2D). If we pay the max of \$271 per 1000 lb for the new grade A tomatoes, the maximum profit contribution with optimal allocation we get is \$676,067.

## 3. Enhancing Demand with Advertising

### 3.1 Which Product to Advertise for?

From the Sensitivity Analysis in Appendix 2B we can see that the demand for juice and whole tomatoes has a slack of 7333.33 and 13,600 respectively, meaning that we are not meeting the demand for these products with the existing optimal solution while on the other hand the demand tomato paste has a slack of zero, meaning that we are completely meeting the demand. So keeping this in mind we can suggest that if we increase the demand for paste through advertising which could possibly result in a new optimal allocation that might improve the total profit contribution. To test the model under advertising criteria (Appendix 3A), we test improved demand cases for all three products in the next section.



### 3.2 Which Product to Advertise For?

Holding the option to buy the additional supply of tomatoes gives a Profit Contribution of \$677,347 (Appendix 2B). With the Marketing Department of Red Brand Canners feeling it could increase demand for any of their three tomato products by 5,000 cases through advertising for this scenario, we test 3 different models of Sensitivity Analysis to help us reach a conclusion. Firstly, we added the advertising demand for each product (Whole Tomatoes, Tomato Juice, Tomato Paste) which is a value deprived from 5000 multiplied by each product's pounds per case (18, 20, 25) divided by 1000. When adding in the advertising demand for our first model on Whole Tomatoes, the Profit Contribution came out to be the same. Similarly, when adding in the advertising demand for our second model on Tomato Juice, we found that the Profit Contribution came out to be the same as well. When we added in the advertising demand for our third model on Tomato Paste, we saw that the Profit Contribution increased from \$677,347 to \$683,388 with the difference being a profit increase of \$6,041.

### 3.3 Maximum Amount to Pay for Advertising

We are now able to conclude that Red Brand Canners should be willing to pay about \$6,041 dollars for the advertising campaign. We are also able to conclude that Tomato Paste is the only product that Red Brand Canners's advertising campaign should be directed toward as increasing the demand for Whole Tomatoes and Tomato Juice do not have any effect on the Profit Contribution at all.

Exhibit 3A: Difference in Profit after Advertising			
	Advertising for Whole Tomatoes	Advertising for Tomato Juice	Advertising for Tomato Paste
Total Profit Contribution Before Advertising	\$677,347	\$677,347	\$677,347
Total Profit Contribution After Advertising	\$677,347	\$677,347	\$683,388
Difference	0	0	\$6,041

## 4. Additional supply of B tomatoes

### 4.1 Is the Additional Stock Worth Buying?

Gordon is offered additional Grade B tomatoes for 18 cents per pound but he refuses to buy the tomatoes claiming that the 18 cents per pound is based on the fact that the incoming crop normally contains a mix of grade A and grade B tomatoes. If we continue with the same model as in 3 which includes the option of buying additional Grade A tomatoes, in addition to the increase in demand because of advertising. It is not a sensible decision to purchase additional B tomatoes because within the existing model we have a lower price for the same resource.

### 4.2 Does the price 18.8 cents / lb improve our profit contribution?

From the model in appendix 3A, we increase the demand for tomato paste with advertising and from appendix 3C we can see that the shadow price for b\_supply is 173 meaning that for every unit increase for b\_supply will increase the objective function or the maximum profit contribution by 173 units. But the cost of this resource is 180 per unit meaning it costs more than it contributes to the profit so we end up losing money and therefore should not buy the additional B tomatoes.

## 5. Closing down production lines?

Starting up the production line for each of the products entails a substantial set-up cost. The estimated set-up costs are \$50,000 per line. This means that if RBC decides not to start up a particular line, they would save \$50,000, but they would not be able to produce that particular product.

### 5.1 Removing Paste Line

If we decide not to start up the paste line, we will see a decrease in our total profit, the explanation for the same is as follows. Initially, if we set up all the production lines for whole, juice and paste we receive a profit contribution of \$676,067 and after deducting the setup cost of 150,000 from this gives us a total profit of \$526,067. After excluding only the paste line, the profit contribution that we saw from the optimization and sensitivity analysis (Appendix 4A) after running this code was \$313,111 and if we subtract the setup costs of \$100,000 for whole and juice from this we get \$213,111 which is the total profit of this combination. As we can clearly

see, \$213,111 is way lower than the actual total profit of \$526,067 and so if we decide not to start up the Paste line, it will lower our total profit.

Exhibit 5A: Profit Contribution if all lines started (Adjusted for Startup Costs)			
Product	Profit Contribution per 1000 lb	Amount Produced (in 1000 lbs)	Total Profit Contribution from Product
Whole Tomatoes	246.67	700	172669
Tomato Juice	198	300	59400
Tomato Paste	222	2000	444000
Profit Contribution			676067
Setup Cost			(150000)
Total Profit Contribution			626067

## 5.2 Which production lines should RBC to start up

We have checked and tested for various scenarios (Appendix 4A,4B, 4C) to find out which lines would fetch us more profit when setup. From our results, we can say that starting the production lines for juice and paste by excluding the whole tomato production line. The profit contribution for this combination was 642,000 and after removing the set-up costs for juice and paste we will have 542,000 as the total profit which is greater than the original total profit 526,067. By removing juice and starting whole, paste lines we got 541,333 which is still lower than 542,000 which we got for juice and paste lines.

Exhibit 5B: Profit Contribution (After Removing Whole Tomato Line)			
Product	Profit Contribution per 1000 lb	Amount Produced (in 1000 lbs)	Total Profit Contribution from Product
Tomato Juice	198	1000	198000
Tomato Paste	222	2000	444000

Profit Contribution	642000
Set-up Cost	(100000)
Total Profit Contribution	542000

Exhibit 5C: Profit Contribution (After Removing Paste Line)

Product	Profit Contribution per 1000 lb	Amount Produced (in 1000 lbs)	Total Profit Contribution from Product
Tomato Whole	246.67	466.667	115111
Tomato Juice	198	1000	198,000
Profit Contribution			313111
Set-up Cost			(100000)
Total Profit Contribution			213111

Exhibit 5D: Profit Contribution (After Removing Juice Line)

Product	Profit Contribution per 1000 lb	Amount Produced (in 1000 lbs)	Total Profit Contribution from Product
Tomato Whole	246.67	800	197333
Tomato Paste	222	2000	444000
Profit Contribution			641333
Set-up Cost			(100000)
Total Profit Contribution			541333

## 6. One year later

An acceptable price would be 20 cents per pound for the next year, they could order up to 13 million pounds of tomatoes. Demand is pretty much the same as last year, i.e. they would be able to sell as many whole canned tomatoes as they can make, and demand for juice and paste is again limited to 50,000 and 80,000 cases, respectively. The only uncertainty is the quality of the crop. Looking at historical data, in the past there were basically three types of years.

- In “wet” years, like last year, the crop was quite poor, and contained only about 20% grade A.
- In “sunny” years, the percentage of A-tomatoes was around 60%.
- In “normal” years, the crop contained approximately 50% grade A tomatoes.

25% of the years were “sunny”, 50% “normal” years and 25% “poor”.

### 6.1 Purchase and Allocation Decisions for Sunny Year

As per the optimal decision variables in Appendix 5B, we should order the whole 13,000,000 lbs of tomatoes given it is a sunny year and allocate the resources as outlined in exhibit 6A.

Exhibit 6A: Breakdown of Production (For Sunny Year)				
	Amount (in 1000 lbs)			
Product	Grade A Tomatoes Used	Grade B Tomatoes Used	Total Product Produced	Profit Contribution
Whole Tomatoes	7525	2525	10,050	\$2,489,084
Tomato Juice	225	675	900	\$178,200
Tomato Paste	0	2000	2000	\$444,000
Total Profit Contribution				\$3,133,533
Cost Incurred for Purchasing				(\$2,600,000)
Net Profit				\$513,533

## 6.2 Purchase and Allocation Decisions for Normal Year

For a normal year, the recommended order quantity (Appendix 5C) is 8,000,000 lbs of tomatoes and the allocation is outlined in Exhibit 6B:

Exhibit 6B: Breakdown of Production (For Normal Year)				
	Amount (in 1000 lbs)			
Product	Grade A Tomatoes Used	Grade B Tomatoes Used	Total Product Produced	Profit Contribution
Whole Tomatoes	3750	1250	5000	\$1,238,350
Tomato Juice	250	750	1000	\$198,000
Tomato Paste	0	2000	2000	\$444,000
Total Profit Contribution				\$1,875,833
Cost Incurred for Purchasing				(\$1,600,000)
Net Profit				\$275,333

## 6.3 Purchase and Allocation Decisions for Poor Year

For a normal year, the recommended order quantity (Appendix 5D) is 2,727,270 lbs of tomatoes and the allocation is outlined in Exhibit 6A:

Exhibit 6C: Breakdown of Production (For Poor Year)	
	Amount (in 1000 lbs)

Product	Grade A Tomatoes Used	Grade B Tomatoes Used	Total Product Produced	Profit Contribution
Whole Tomatoes	545.455	181.818	727.273	\$180,124
Tomato Juice	0	0	0	0
Tomato Paste	0	2000	2000	\$444,000
Total Profit Contribution				\$623,393
Cost Incurred for Purchasing				(\$545,454)
Net Profit				\$77,939

## 6.4 Sunny Season Order Scenario Analysis

In our model from appendix 5A, we change the constraint for order quantity to be binding == 13000 and change the weather to get the optimal profit (Appendix 5E, 5F).

Exhibit 6D compares the difference in each scenario:

Exhibit 6D: Breakdown of Production (Order Quantity = 13,000,000 lbs)			
	Net Revenue		
Year	Good Year	Normal Year	Poor Year
Revenue Generated	\$3,133,533	\$2,697,556	\$1,414,890
Cost	(\$2,600,000)	(\$2,600,000)	(\$2,600,000)
Net Profit	\$513,533	\$97,556	(\$1,185,110)

Looking at the scenario analysis above, if we order according to a sunny year and get a normal year we still make some profit while if it were a poor year RBC would end up losing money.

## 6.5 Expected Value Analysis

The probability for each type of weather is as follows:

- Sunny = 0.25
- Normal = 0.5
- Poor = 0.25

The average profit overall based on the probabilities would be:

Exhibit 6E: Expected Value Analysis (Order Quantity = Optimized for Season)			
	Net Revenue		
Year	Good Year	Normal Year	Poor Year
Revenue Generated	\$3,133,533	\$1,875,833	\$623,393
Cost	(\$2,600,000)	(\$1,600,000)	(\$545,454)
Net Profit	\$513,533	\$275,333	\$77,939
Probability	0.25	0.5	0.25
Expected Value	$= 0.25 * \$513,533 + 0.5 * \$275,333 + 0.25 * \$77,939$ $= \$285,535$		

## 6.6 Final Purchase Recommendation:

Given the different analyses above, the best course of action recommended for purchase is to buy according to the normal year production so the chances of profit would be the largest



# Appendices

## 1A: Model 1

```
# Model m =
gp.Model("RBC")

# Create decision variables for tomatoes usage
aw = m.addVar(name="aw") aj =
m.addVar(name="aj") ap = m.addVar(name="ap")
bw = m.addVar(name="bw") bj =
m.addVar(name="bj") bp = m.addVar(name="bp")

# The objective is to maximize the profit contribution
obj = (4.44/18*1000)*(aw+bw)+198*(aj+bj)+222*(ap+bp)
m.setObjective(obj, GRB.MAXIMIZE)

# Demand constraints con1 =
m.addConstr(aw+bw<=14400, name='w_dem') con2 =
m.addConstr(aj+bj<=1000, name='j_dem') con3 =
m.addConstr(ap+bp<=2000, name='p_dem')

# Supply constraints con4 =
m.addConstr(aw+aj+ap<=600, name='a_sup') con5 =
m.addConstr(bw+bj+bp<=2400, name='b_sup')

# Quality constraints con6 =
m.addConstr(9*aw+5*bw>=8*(aw+bw), name='w_qual') con7
= m.addConstr(9*aj+5*bj>=6*(aj+bj), name='j_qual')

# Solve
m.optimize()
```

## 1B: Model 1 Optimal Decision Variables and Sensitivity Report

Optimal objective 6.760666667e+05

Profit contribution: 676067

Decision

variables: aw =

525 aj = 75 ap = 0

bw = 175 bj = 225

bp = 2000

Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range
0 aw	525.0	0.000000	246.666667	709.777778	181.777778
1 aj	75.0	0.000000	198.000000	262.888889	-265.111111
2 ap	0.0	-97.333333	222.000000	319.333333	-inf
3 bw	175.0	0.000000	246.666667	1636.000000	181.777778
4 bj	225.0	0.000000	198.000000	240.962963	43.629630
5 bp	2000.0	0.000000	222.000000	inf	173.666667

Name	Shadow Price	RHS Coeff	Slack	Upper Range	Lower Range
0 w_dem	0.000000	14400	13700.0	inf	700.000000
1 j_dem	0.000000	1000	700.0	inf	300.000000
2 p_dem	48.333333	2000	0.0	2200.000000	1533.333333
3 a_sup	271.000000	600	0.0	1200.000000	133.333333
4 b_sup	173.666667	2400	0.0	2866.666667	2200.000000
5 w_qual	-24.333333	0	0.0	466.666667	-600.000000
6 j_qual	-24.333333	0	0.0	1400.000000	-200.000000

## 2A: Model 2

```
# Model m_2 =
gp.Model("RBC_2")

# Create decision variables for tomatoes usage
aw = m_2.addVar(name="aw") aj =
m_2.addVar(name="aj") ap =
m_2.addVar(name="ap") bw =
m_2.addVar(name="bw") bj =
m_2.addVar(name="bj")
```

```

bp = m_2.addVar(name="bp")
AAw = m_2.addVar(name="AAw")
AAp = m_2.addVar(name="AAp")
AAj = m_2.addVar(name="AAj")

adDemandW = 0
adDemandJ = 0
adDemandP = 0

costAA = 255

obj_2 =
(4.44/18*1000)*(AAw+aw+bw)+198*(AAj+aj+bj)+222*(AAp+ap+bp)-costAA*(AAw+A
Aj+AAp) m_2.setObjective(obj_2, GRB.MAXIMIZE)

# Demand constraints con11 =
m_2.addConstr(aw+bw+AAw<=14400+adDemandW, name='w_dem') con22
= m_2.addConstr(aj+bj+AAj<=1000+adDemandJ, name='j_dem') con33
= m_2.addConstr(ap+bp+AAp<=2000+adDemandP, name='p_dem')

# Supply constraints con44 =
m_2.addConstr(aw+aj+ap<=600, name='a_sup') con55 =
m_2.addConstr(bw+bj+bp<=2400, name='b_sup') con66 =
m_2.addConstr(AAw+AAj+AAp<=80, name='AA_sup')

# Quality constraints con77 =
m_2.addConstr(9*aw+9*AAw+5*bw>=8*(aw+bw+AAw), name='w_qual') con88 =
m_2.addConstr(9*aj+9*AAj+5*bj>=6*(aj+bj+AAj), name='j_qual')

# Solve
m_2.optimize()

```

## 2B: Model 2 Optimal Decisions Variables and Sensitivity Report

Optimal objective 6.773466667e+05

Profit contribution: 677347

Decision variables:

aw = 535

```

aj = 65 ap
= 0 bw =
205 bj =
195 bp =
2000 AAw =
80
AAp = 0
AAj = 0

```

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range	0	aw
		535.0	0.000000e+00	246.666667	246.666667	181.777778		
1	aj	65.0	0.000000e+00	198.000000	262.888889	198.000000		
2	ap	0.0	-9.733333e+01	222.000000	319.333333	-inf		
3	bw	205.0	0.000000e+00	246.666667	1636.000000	204.000000		
4	bj	195.0	0.000000e+00	198.000000	240.666667	43.629630		
5	bp	2000.0	0.000000e+00	222.000000	inf	173.666667		
6	AAw	80.0	0.000000e+00	-8.333333	inf	-8.333333		
7	AAj	0.0	1.065814e-14	-57.000000	-57.000000	-inf		
8	AAp	0.0	-9.733333e+01	-33.000000	64.333333	-inf		

	Name	Shadow Price	RHS	Coeff	Slack	Upper Range	Lower Range
0	w_dem	0.000000	14400	13580.0	inf	820.000000	
1	j_dem	0.000000	1000	740.0	inf	260.000000	
2	p_dem	48.333333	2000	0.0	2173.333333	1506.666667	
3	a_sup	271.000000	600	0.0	1120.000000	124.444444	
4	b_sup	173.666667	2400	0.0	2893.333333	2226.666667	
5	AA_sup	16.000000	80	0.0	600.000000	0.000000	
6	w_qual	-24.333333	0	0.0	546.666667	546.666667	
7	j_qual	-24.333333	0	0.0	1426.666667	1426.666667	

## 2C: Alternative Optimal Solution for Model 2

Optimal objective 6.773466667e+05

Profit contribution: 677347

Decision variables:

aw = 600

aj = 0

ap = 0

bw = 205

bj = 195

bp = 2000

AAw = 15

AAp = 0

AAj = 65

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range
0	aw	600.0	0.000000	246.666667	inf	246.666667
1	aj	0.0	-0.000000	198.000000	198.000000	-inf
2	ap	0.0	-97.333333	222.000000	319.333333	-inf
3	bw	205.0	0.000000	246.666667	1636.000000	204.000000
4	bj	195.0	0.000000	198.000000	240.666667	43.629630
5	bp	2000.0	0.000000	222.000000	inf	173.666667
6	AAw	15.0	0.000000	-8.333333	-8.333333	-22.555556
7	AAj	65.0	0.000000	-57.000000	7.888889	-57.000000
8	AAp	0.0	-97.333333	-	64.333333	-inf
				33.000000		
	Name	Shadow Price	RHS Coeff	Slack	Upper Range	Lower Range
0	w_dem	0.000000	14400	13580.0	inf	820.000000
1	j_dem	0.000000	1000	740.0	inf	260.000000
2	p_dem	48.333333	2000	0.0	2173.333333	1960.000000
3	a_sup	271.000000	600	0.0	1120.000000	480.000000
4	b_sup	173.666667	2400	0.0	2440.000000	2226.666667
5	AA_sup	16.000000	80	0.0	600.000000	66.666667
6	w_qual	-24.333333	0	0.0	120.000000	120.000000
7	j_qual	-24.333333	0	0.0	40.000000	40.000000

## 2D: Maximum Price Paid for New Supply

Optimal objective 6.760666667e+05

Profit contribution: 676067

Decision variables:

aw = 600

aj = 0

ap = 0

bw = 200

bj = 200

bp = 2000

AAw = 0

AAp = 0

AAj = 66.6667

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range
0	aw	600.000000	0.000000e+00	246.666667	inf	246.666667
1	aj	0.000000	-5.684342e-14	198.000000	198.000000	-inf
2	ap	0.000000	-9.733333e+01	222.000000	319.333333	-inf
3	bw	200.000000	0.000000e+00	246.666667	246.666667	246.666667
4	bj	200.000000	0.000000e+00	198.000000	198.000000	198.000000
5	bp	2000.000000	0.000000e+00	222.000000	inf	173.666667
6	AAw	0.000000	7.105427e-14	-24.333333	-24.333333	-inf
7	AAj	66.666667	0.000000e+00	-73.000000	-0.000000	-73.000000
8	AAp	0.000000	-9.733333e+01	-49.000000	48.333333	-inf

  

	Name	Shadow Price	RHS Coeff	Slack	Upper Range	Lower Range
0	w_dem	0.000000	14400	13600.000000	inf	800.000000
1	j_dem	0.000000	1000	733.333333	inf	266.666667
2	p_dem	48.333333	2000	0.000000	2200.0	1960.000000
3	a_sup	271.000000	600	0.000000	1200.0	480.000000
4	b_sup	173.666667	2400	0.000000	2440.0	2200.000000
5	AA_sup	0.000000	80	13.333333	inf	66.666667
6	w_qual	-24.333333	0	0.000000	120.0	120.000000
7	j_qual	-24.333333	0	0.000000	40.0	40.000000

### 3A: Model with Advertising

```
# Model m_2 =
gp.Model("RBC_2")

# Create decision variables for tomatoes usage
aw = m_2.addVar(name="aw") aj =
m_2.addVar(name="aj") ap =
m_2.addVar(name="ap") bw =
m_2.addVar(name="bw") bj =
m_2.addVar(name="bj") bp =
m_2.addVar(name="bp")
AAw = m_2.addVar(name="AAw")
AAp = m_2.addVar(name="AAp")
AAj = m_2.addVar(name="AAj")

adDemandW = 0
adDemandJ = 0
adDemandP = 0

costAA = 255
#adDemandW = (5000*18)/1000
#adDemandJ = (5000*20)/1000
#adDemandP = (5000*25)/1000
```

```

obj_2 =
(4.44/18*1000)*(AAw+aw+bw)+198*(AAj+aj+bj)+222*(AAp+ap+bp)-costAA*(AAw+A
Aj+AAp) m_2.setObjective(obj_2, GRB.MAXIMIZE)

# Demand constraints con11 =
m_2.addConstr(aw+bw+AAw<=14400+adDemandW, name='w_dem') con22
= m_2.addConstr(aj+bj+AAj<=1000+adDemandJ, name='j_dem') con33
= m_2.addConstr(ap+bp+AAp<=2000+adDemandP, name='p_dem')

# Supply constraints con44 =
m_2.addConstr(aw+aj+ap<=600, name='a_sup') con55 =
m_2.addConstr(bw+bj+bp<=2400, name='b_sup') con66 =
m_2.addConstr(AAw+AAj+AAp<=80, name='AA_sup')

# Quality constraints con77 =
m_2.addConstr(9*aw+9*AAw+5*bw>=8*(aw+bw+AAw), name='w_qual') con88 =
m_2.addConstr(9*aj+9*AAj+5*bj>=6*(aj+bj+AAj), name='j_qual')

# Solve
m_2.optimize()

```

### 3B: Advertising for Whole Tomatoes

Profit contribution: 677347

Decision variables:

aw = 535

aj = 65

ap = 0

bw = 205

bj = 195

bp = 2000

AAw = 80

AAp = 0

AAj = 0

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range
0	aw	535.00	.000000e+00	246.666667	246.666667	181.777778



1	aj	65.00.000000e+00	198.000000	262.888889	198.000000	
2	ap	0.0 -9.733333e+01	222.000000	319.333333	-inf	
3	bw	205.0 0.000000e+00	246.666667	1636.000000	204.000000	
4	bj	195.0 0.000000e+00	198.000000	240.666667	43.629630	
5	bp	2000.0 0.000000e+00	222.000000	inf	173.666667	
6	AAw	80.0 0.000000e+00	-8.333333	inf	-8.333333	
7	AAj	0.0 1.065814e-14	-57.000000	-57.000000	-inf	
8	AAp	0.0 -9.733333e+01	-	64.333333	-inf	
			33.000000			
	Name	Shadow Price	RHS Coeff	Slack	Upper Range	Lower Range
0	w_dem	0.000000	14490.0	13670.0	inf	820.000000
1	j_dem	0.000000	1000.0	740.0	inf	260.000000
2	p_dem	48.333333	2000.0	0.0	2173.333333	1506.666667
3	a_sup	271.000000	600.0	0.0	1120.000000	124.444444
4	b_sup	173.666667	2400.0	0.0	2893.333333	2226.666667
5	AA_sup	16.000000	80.0	0.0	600.000000	0.000000
6	w_qual	-24.333333	0.0	0.0	546.666667	546.666667
7	j_qual	-24.333333	0.0	0.0	1426.666667	1426.666667

### 3C: Advertising for Tomato Juice

Optimal objective 6.773466667e+05

Profit contribution: 677347

Decision variables:

aw = 535 aj = 65 ap =

0 bw = 205 bj = 195

bp = 2000 AAw = 80

AAp = 0

AAj = 0

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range	0	aw
	535.0	0.000000e+00	246.666667	246.666667	181.777778			
1	aj	65.0	0.000000e+00	198.000000	262.888889	198.000000		
2	ap	0.0	-9.733333e+01	222.000000	319.333333	-inf		
3	bw	205.0	0.000000e+00	246.666667	1636.000000	204.000000		
4	bj	195.0	0.000000e+00	198.000000	240.666667	43.629630		
5	bp	2000.0	0.000000e+00	222.000000	inf	173.666667		
6	AAw	80.0	0.000000e+00	-8.333333	inf	-8.333333		
7	AAj	0.0	1.065814e-14	-57.000000	-57.000000	-inf		
8	AAp	0.0	-9.733333e+01	-33.000000	64.333333	-inf		

	Name	Shadow Price	RHS Coeff	Slack	Upper Range	Lower Range
0	w_dem	0.000000	14400.0	13580.0	inf	820.000000
1	j_dem	0.000000	1100.0	840.0	inf	260.000000

2	p_dem	48.333333	2000.0	0.0	2173.333333	1440.000000
3	a_sup	271.000000	600.0	0.0	1120.000000	124.444444
4	b_sup	173.666667	2400.0	0.0	2960.000000	2226.666667
5	AA_sup	16.000000	80.0	0.0	600.000000	0.000000
6	w_qual	-24.333333	0.0	0.0	546.666667	546.666667
7	j_qual	-24.333333	0.0	0.0	1426.666667	1426.666667

### 3D: Advertising for Tomato Paste

Optimal objective 6.833883333e+05

Profit contribution: 683388

Decision variables:

aw = 581.875

aj = 18.125

ap = 0

bw =

220.625 bj

= 54.375 bp

= 2125 AAw

= 80

AAp = 0

AAj = 0

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range	0	aw
		581.875	0.000000e+00	246.666667	246.666667	181.777778		
1	aj	18.125	0.000000e+00	198.000000	262.888889	198.000000		
2	ap	0.000	-9.733333e+01	222.000000	319.333333	-inf		
3	bw	220.625	0.000000e+00	246.666667	1636.000000	204.000000		
4	bj	54.375	0.000000e+00	198.000000	240.666667	43.629630		
5	bp	2125.000	0.000000e+00	222.000000	inf	173.666667		
6	AAw	80.000	0.000000e+00	-8.333333	inf	-8.333333		
7	AAj	0.000	1.065814e-14	-57.000000	-57.000000	-inf		
8	AAp	0.000	-9.733333e+01	-33.000000	64.333333	-inf		

	Name	Shadow Price	RHS Coeff	Slack	Upper Range	Lower Range
0	w_dem	0.000000	14400.0	13517.5	inf	882.500000
1	j_dem	0.000000	1000.0	927.5	inf	72.500000
2	p_dem	48.333333	2125.0	0.0	2173.333333	1506.666667
3	a_sup	271.000000	600.0	0.0	745.000000	82.777778
4	b_sup	173.666667	2400.0	0.0	3018.333333	2351.666667
5	AA_sup	16.000000	80.0	0.0	225.000000	0.000000
6	w_qual	-24.333333	0.0	0.0	588.333333	588.333333
7	j_qual	-24.333333	0.0	0.0	1551.666667	1551.666667

## 4A: Optimal Solution with Startup Costs (Removing Paste Line)

Optimal objective 3.131111111e+05

Profit contribution: 313111

Decision variables:

aw = 350 aj = 250

bw = 116.667 bj =  
750

Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range
0 aw	350.000000	0.0	246.666667	709.777778	-8.222222e+01
1 ap	250.000000	0.0	198.000000	526.888889	-2.651111e+02
2 bw	116.666667	0.0	246.666667	1636.000000	-2.842171e-14
3 bp	750.000000	0.0	198.000000	inf	4.362963e+01

Name	Shadow Price	RHS Coeff	Slack	Upper Range	Lower Range
0 w_dem	0.000000	14000	13933.333333	2400.0	466.666667
1 j_dem	115.777778	1000	0.000000	2400.0	0.000000
2 a_sup	328.888889	600	0.000000	5200.0	250.000000
3 b_sup	0.000000	2400	1533.333333	inf	866.666667
4 w_qual	-82.222222	0	0.000000	350.0	866.666667
5 j_qual	-82.222222	0	0.000000	1400.0	-1000.000000

## 4B: Optimal Solution with Startup Costs (Removing Juice Line)

Optimal objective 6.413333333e+05

Profit contribution: 641333

Decision variables:

aw = 600

ap = 0

bw = 200

bp = 2000

Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range
0 aj	600.0	0.000000	246.666667	inf	-8.222222e+01
1 ap	0.0	-328.888889	246.666667	550.888889	-inf

2	bj	200.0	0.000000	198.000000	inf	-2.842171e-14
3	bp	2000.0	0.000000	198.000000	inf	-0.000000e+00

  

	Name	Shadow Price	RHS	Coeff	Slack	Upper Range	Lower Range
0	w_dem	0.000000	14400	13600.0	inf	800.0	
1	p_dem	222.000000	2000	0.0	2200.0	0.0	
2	a_sup	328.888889	600	0.0	1200.0	0.0	
3	b_sup	0.000000	2400	200.0	inf	2200.0	
4	w_qual	-82.222222	0	0.0	600.0	-600.0	

#### 4C: Optimal Solution with Startup Costs (Removing Whole Line)

Optimal objective 6.42000000 0e+05

Profit contribution: 642000

Decision variables:

aj = 250

ap = 350

bj = 750

bp = 1650

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	Lower Range
0	aj	250.0	0.0	198	198.0	-594.0
1	ap	350.0	0.0	222	222.0	222.0
2	bj	750.0	0.0	198	inf	198.0
3	bp	1650.0	0.0	222	222.0	222.0

	Name	Shadow Price	RHS	Coeff	Slack	Upper Range	Lower Range
0	j_dem	198.0	1000	0.0	1000.0	533.333333	
1	p_dem	222.0	2000	0.0	2000.0	1650.000000	
2	a_sup	0.0	600	0.0	inf	600.000000	
3	b_sup	0.0	2400	0.0	2750.0	2400.000000	
4	j_qual	-0.0	0	0.0	1400.0	-1000.000000	

## 5A: Model with Purchase Quantity Decision and Weather

```
# Model m =  
gp.Model("RBC")  
  
# Create decision variables for tomatoes usage  
aw = m.addVar(name="aw") aj =  
m.addVar(name="aj")
```

```

ap = m.addVar(name="ap") bw =
m.addVar(name="bw") bj =
m.addVar(name="bj") bp =
m.addVar(name="bp") order_quant =
m.addVar(name="order_quant")

weather_pred = 'S'

# The objective is to maximize the profit contribution
obj = (4.44/18*1000)*(aw+bw) + 198*(aj+bj) + 222*(ap+bp)
-
200*(order_quant)
m.setObjective(obj, GRB.MAXIMIZE)

# Demand constraints

con2 = m.addConstr(aj+bj<=1000, name='j_dem')
con3 = m.addConstr(ap+bp<=2000, name='p_dem')

# Quality constraints con4 =
m.addConstr(9*aw+5*bw>=8*(aw+bw), name='w_qual') con5
= m.addConstr(9*aj+5*bj>=6*(aj+bj), name='j_qual')

#Purchase Constraint con6 = m.addConstr(order_quant <= 13000,
name='purchase_quantity')

#Supply Constraint for wet year
if weather_pred == 'S':
    A_prop = 0.6
    B_prop = 0.4
    P_W = 0.25
#Supply Constraint for wet year
elif weather_pred == 'N':
    A_prop = 0.5
    B_prop = 0.5
    P_W = 0.5

```

```
#Supply Constraint for wet year
```

```
elif weather_pred == 'P':
```

```
    A_prop = 0.2
```

```
    B_prop = 0.8
```

```
    P_W = 0.25
```



```

con7 = m.addConstr(aw+aj+ap<=A_prop*order_quant) con8 =
m.addConstr(bw+bj+bp<=B_prop*order_quant)

# Solve
m.optimize()

```

## 5B: Optimal Decision for Sunny Year

Optimal objective 5.135333333e+05

Profit contribution: 513533

Decision variables:

aw = 7575 aj = 225 ap =

0 bw = 2525 bj =

675 bp = 2000

order\_quant = 13000

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range \
0	aw	7575.0	0.000000	246.666667	709.777778
1	aj	225.0	0.000000	198.000000	262.888889
2	ap	0.0	-97.333333	222.000000	319.333333
3	bw	2525.0	0.000000	246.666667	1636.000000
4	bj	675.0	0.000000	198.000000	240.962963
5	bp	2000.0	0.000000	222.000000	inf
6	order_quant	13000.0	0.000000	-200.000000	inf

Lower Range

0	185.587302
1	-229.555556
2	-inf
3	181.777778
4	55.481481
5	173.666667
6	-232.066667

	Name	Shadow Price	RHS	Coeff	Slack	Upper Range	Lower
Range							
0	j_dem	0.000000	1000.0	100.0	inf		
900.000000							
1	p_dem	48.333333	2000.0	0.0	2.600000e+03		
1933.333333							
2	w_qual	-24.333333		0.0	0.0	2.000000e+02	

-1800.000000					
3	p_qual	-24.333333	0.0	0.0	2.000000e+02
-600.000000	4				
order_quantity		32.066667	13000.0	0.0	1.333333e+04
10000.000000					
5	a_sup	271.000000	7800.0	0.0	1.800000e+03
-200.000000					
6	b_sup	173.666667	5200.0	0.0	6.666667e+01
-600.000000					

## 5C: Optimal Decision for Normal Year

Optimal objective 2.753333333e+05					
Profit contribution: 275333					
Decision variables:					
aw = 3750 aj = 250					
ap = 0 bw = 1250					
bj = 750 bp = 2000					
order_quant = 8000					
	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range \
0	aw	3750.0	0.000000	246.666667	317.777778
1	aj	250.0	0.000000	198.000000	384.666667
2	ap	0.0	-186.666667	222.000000	408.666667
3	bw	1250.0	0.000000	246.666667	460.000000
4	bj	750.0	0.000000	198.000000	inf
5	bp	2000.0	0.000000	222.000000	inf
6	order_quant	8000.0	0.000000	-200.000000	-164.444444
Lower Range					
0		187.111111			
1		19.333333			
2		-inf			
3		153.333333			
4		138.444444			
5		106.666667			
6		-222.333333			
	Name	Shadow Price	RHS Coeff	Slack	Upper Range Lower
Range					
0	j_dem	44.666667	1000.0	0.0	3500.000000
0.000000					
1	p_dem	115.333333	2000.0	0.0	3666.666667
0.000000					
2	w_qual	-46.666667	0.0	0.0	2500.000000
-5000.000000					
		-46.666667	0.0	0.0	3000.000000

3	p_qual				
	-1000.000000	4			
	order_quantity	0.000000	13000.0	5000.0	inf
	8000.000000				
5	a_sup	293.333333	6500.0	0.0	5000.000000
	-2500.000000	106.666667	6500.0	0.0	2500.000000
6	b_sup				
	-1666.666667				

## 5D: Optimal Decision for Poor Year

Optimal objective 7.793939394e+04					
Profit contribution: 77939.4					
Decision variables:					
aw = 545.455					
aj = 0					
ap = 0					
bw = 181.818					
bj = 0					
bp = 2000					
order_quant = 2727.27					
	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range \
0	aw	545.454545	0.000000	246.666667	917.777778
1	aj	0.000000	0.000000	198.000000	222.969697
2	ap	0.000000	-84.848485	222.000000	306.848485
3	bw	181.818182	0.000000	246.666667	2260.000000
4	bj	0.000000	-8.323232	198.000000	206.323232
5	bp	2000.000000	0.000000	222.000000	inf
6	order_quant	2727.272727	0.000000	-200.000000	-193.133333
	Lower Range				
0		184.444444			

1	-inf
2	-inf
3	188.333333
4	-inf
5	183.030303
6	-228.577778

  

	Name	Shadow Price	RHS	Coeff	Slack	Upper Range
\						
0	j_dem	0.000000	1000.0	1000.000000		inf
1	p_dem	38.969697	2000.0	0.000000	9.533333e+03	
2	w_qual	-21.212121	0.0	0.000000	5.000000e+02	
3	p_qual	-23.292929	0.0	0.000000	1.500000e+03	
4	order_quantity	0.000000	13000.0	10272.727273		inf
5	a_sup	267.878788	2600.0	0.000000	2.260000e+04	
6	b_sup	183.030303	10400.0	0.000000	2.000000e+03	
Lower Range						
0	0.000000	1				
	0.000000					
2	-					
	22600.00000					
	0					
3	-0.000000					
4	2727.272727					
5	-500.000000					
6	-					
	7533.333333					

## 5E: Order for Sunny But Actual Weather Poor

Optimal objective -1.185111111e+06

Profit contribution: -1.18511e+06

Decision variables:

aw = 2350

aj = 250

ap = 0

bw = 783.333

bj = 750

```
bp = 2000
order_quant = 13000
```

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range \
0	aw	2350.000000	0.000000	246.666667	709.777778
1	aj	250.000000	0.000000	198.000000	526.888889
2	ap	0.000000	-328.888889	222.000000	550.888889
3	bw	783.333333	0.000000	246.666667	1636.000000
4	bj	750.000000	0.000000	198.000000	inf
5	bp	2000.000000	0.000000	222.000000	inf
6	order_quant	13000.000000	0.000000	-200.000000	inf

```
Lower Range
0 -8.222222e+01
1 -2.651111e+02
2 -inf
3 -2.842171e-14
4 4.362963e+01
5 -0.000000e+00
```

```
6 -inf
```

	Name	Shadow Price	RHS Coeff	Slack	Upper Range \
0	j_dem	115.777778	1000.0	0.000000	1.040000e+04
1	p_dem	222.000000	2000.0	0.000000	8.866667e+03
2	w_qual	-82.222222	0.0	0.000000	2.350000e+03
3	p_qual	-82.222222	0.0	0.000000	3.000000e+03
4	order_quantity	-134.222222	13000.0	0.000000	inf
5	a_sup	328.888889	2600.0	0.000000	2.060000e+04
6	b_sup	0.000000	10400.0	6866.666667	inf

```
Lower Range
0 0.000000
```

1	0.000000
2	-
	20600.000000
3	-1000.000000
4	3636.363636
5	-2350.000000
6	-6866.666667

## 5F: Order for Sunny But Actual Weather Normal

Optimal objective 9.755555556e

Profit contribution: 97555.6

Decision variables:

aw = 6250

aj = 250

ap = 0

bw = 2083.33

bj = 750

bp = 2000

order\_quant = 13000

	Name	Final Value	Reduced Cost	Obj Coeff	Upper Range	\
0	aw	6250.000000	0.000000	246.666667	709.777778	
1	aj	250.000000	0.000000	198.000000	526.888889	
2	ap	0.000000	-328.888889	222.000000	550.888889	
3	bw	2083.333333	0.000000	246.666667	1636.000000	
4	bj	750.000000	0.000000	198.000000	inf	
5	bp	2000.000000	0.000000	222.000000	inf	
6	order_quant	13000.000000	0.000000	-200.000000	inf	

Lower Range

0	-8.222222e+01
1	-2.651111e+02
2	-inf
3	-2.842171e-14

```

4  4.362963e+01
5  0.000000e+00
6  -inf

```

	Name	Shadow Price	RHS Coeff	Slack	Upper Range \
0	j_dem	115.777778	1000.0	0.000000	3500.000000
1	p_dem	222.000000	2000.0	0.000000	3666.666667
2	w_qual	-82.222222	0.0	0.000000	6250.000000
3	p_qual	-82.222222	0.0	0.000000	3000.000000
4	order_quantity	-35.555556	13000.0	0.000000	inf
5	a_sup	328.888889	6500.0	0.000000	5000.000000
6	b_sup	0.000000	6500.0	1666.666667	inf

```

Lower Range
0  0.000000
1  0.000000
2  -5000.000000
3  -1000.000000
4  8000.000000
5  -6250.000000
6  -1666.666667

```