# Cali Coffee Company Production Optimisation

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# 1 Client information

### 1.1 Available products

The Cali Coffee Company (CCC) are an international supplier of ground coffee mixes. In their production process they have 4 component beans:

- 1. Abundo
- 2. Colmado
- 3. Maximo
- 4. Saboro

CCC sells 3 blends:

- 1. Hotel (luxury) blend
- 2. Restaurant blend
- 3. Market (retail) blend

CCC has 4 bean suppliers:

- 1. Colombian Abundo
- 2. Peruvian Calmado
- 3. Brazilian Maximo
- 4. Chilean Saboro

#### 1.2 Coffee Blend information

Each blend is comprised of a mix of different components. The relative proportion of these components is indicated for each blend as a percentage (%).

component	Hotel	Restaurant	Market	$\mathrm{Cost/lbs}(\$\mathrm{usd})$	availability (lbs)
Abundo	20%	35%	10%	0.6	40000
Colmado	40%	15%	35%	0.8	25000
Maximo	15%	20%	40%	0.55	20000
Saboro	25%	30%	15%	0.7	45000
Wholesale price/lb (\$usd)	2.25	2.5	2.4		

#### 1.3 Additional Production Constraints

- The processing plant can handle no more than 100000 pounds per week.
- There are ongoing contracts for production: Minimum 10000, 25000, 30000 pounds of Hotel, Restaurant and Market blend must be produced to satisfy these contracts.

# 2 Results and Analysis

#### 2.1 Optimal Allocation

Based on the current constraints and prices, the optimal production schedule is as follows:

This production schedule generates \$129350.0 (usd).

### 2.2 Binding Constraints:

Analysis has identified 3 limiting factors which are preventing rises in profits:

- 1. 20000lbs Maximo bean availability.
- 2. Existing contractual obligation to produce 25000lbs of Hotel blend.
- 3. Existing contractual obligation to produce 30000lbs of Market blend.

#### 2.3 Sensitivity Analysis

#### 2.3.1 Maximo bean availability

For every extra pound of Maximo beans made available we can expect to generate an additional \$10 in profit. Similarly, if the availability of Maximo beans decreases, we expect to sustain losses of \$10 per pound decrease.

#### 2.3.2 Contractual obligation to supply Market blend

For every pound reduction in the market supply contract we can expect to generate an additional \$2 in profit. Similarly, if our contractual obligations increase, we expect to sustain losses of \$2 per pound increase.

# 2.3.3 Other

Changes to any other constraint will result in no difference in profits.

# 3 Linear Programming Formalism

## 3.1 Intro / Objective

Objective is to maximise profit by selecting production amounts for Hotel, Restaurant and Market blends: Profit = Revenue - Costs

# 3.2 Production schedule

Let our production schedule be denoted 
$$\mathbf{x}$$
.  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \text{Hotel (lbs)} \\ \text{restaurant (lbs)} \\ \text{market (lbs)} \end{bmatrix}$ 

#### 3.3 Coffee Blend Matrix

To get the vector of required beans for a given blend we define the following

Coffee Blend Matrix A: 
$$A = \begin{bmatrix} 0.2 & 0.35 & 0.1 \\ 0.4 & 0.15 & 0.35 \\ 0.15 & 0.2 & 0.4 \\ 0.25 & 0.3 & 0.15 \end{bmatrix}$$

#### 3.4 Bean Cost Vector

We also define a Bean Cost Vector c, each element in this vector is the

cost/lbs (\$usd) for a given component. 
$$\mathbf{c} = \begin{bmatrix} 0.6 \\ 0.8 \\ 0.55 \\ 0.7 \end{bmatrix}$$

#### 3.5 Costs

Provided a production schedule, coffee blend matrix and bean cost vector, the total cost of this amount of production is given by:  $\mathbf{Costs} = \mathbf{c}^{\top} A \mathbf{x}$ 

#### 3.6 Coffee Price Vector

We define a price vector, **p**, each element is the price/lb CCC generates for

selling a particular coffee blend: 
$$\mathbf{p} = \begin{bmatrix} 2.25 \\ 2.5 \\ 2.4 \end{bmatrix}$$

#### 3.7 Revenue

Provided a production schedule and Coffee Price Vector, the total revenue of this amount of production is given by: Revenue  $= \mathbf{p}^{\top} \mathbf{x}$ 

# 3.8 Objective

$$\max_{\mathbf{x}}(\mathbf{p}^{\top}\mathbf{x} - \mathbf{c}^{\top}A\mathbf{x})$$

# 3.9 Constraints

## 3.9.1 Ongoing coffee contracts

$$\mathbf{x} \succeq \begin{bmatrix} 10000 \\ 25000 \\ 30000 \end{bmatrix}$$

# 3.9.2 Supplier/bean availability

$$A\mathbf{x} \preceq \begin{bmatrix} 40000 \\ 25000 \\ 20000 \\ 45000 \end{bmatrix}$$

# 3.9.3 Processing Plant Capacity

$$x_1 + x_2 + x_3 \le 100000$$

## 3.9.4 Non-negative Production

$$\mathbf{x}\succeq \mathbf{0}$$

#### 4 Code

```
from scipy import optimize
import scipy
import numpy as np
A = np.matrix([[0.2, 0.35, 0.1],
               [0.4, 0.15, 0.35],
               [0.15, 0.2, 0.4],
               [0.25, 0.3, 0.15]
c = np.array([0.6, 0.8, 0.55, 0.7])
p = np.array([2.25, 2.5, 2.4])
objective = -(p.T - c.T @ A)
A_ub = np.matrix([[0.2, 0.35, 0.1], # Abundo usage
                  [0.4, 0.15, 0.35], # Colmado usage
                  [0.15, 0.2, 0.4], # Maximo usage
                  [0.25, 0.3, 0.15], # Saboro usage
                  [1, 1, 1], # Total production
                  [-1, 0, 0], # Hotel
                  [0, -1, 0], # Restaurant
                  [0, 0, -1]]) # Market
b_ub = np.array([40000, # Abundo availability
                 25000, # Colmado availability
                 20000, # Maximo availability
                 45000, # Saboro availability
                 100000, # Maximum production
                 -10000, # Min Hotel
                 -25000, # Min Restaurant
                 -30000]) # Min Market
primal_result = optimize.linprog(objective, A_ub, b_ub)
print('RESULTS OF PRIMAL LP')
print(f'Optimal assignment: {list(map(round, primal_result.x))}')
print(f'Maximum value: {round(primal_result.fun)}')
print(f'Slack values: {list(map(round, primal_result.slack))}')
```

```
print()

dual_objective = b_ub.T

dual_A_ub = -A_ub.T

dual_b_ub = objective.T

dual_result = optimize.linprog(dual_objective, dual_A_ub, dual_b_ub)

print('RESULTS OF DUAL LP')

print(f'Optimal assignment: {list(map(round, dual_result.x))}')

print(f'Maximum value: {round(dual_result.fun)}')

print(f'Slack values: {list(map(round, dual_result.slack))}')
```