

Production Planning Optimization

This notebook demonstrates a simple optimization model using Linear Programming (PuLP) to maximize profit for two products (A and B) under limited resources.

Problem Setup

- Profit per unit:
 - Product A: \$40
 - Product B: \$30
- Resources required:
 - Labor: A=2 hrs, B=1 hr
 - Materials: A=1, B=1
- Resource availability:
 - Labor: 100 hrs
 - Materials: 80 units

Objective

Maximize total profit while staying within labor and material constraints.

```
In [2]: !pip install pulp
import pulp
```

Collecting pulp

Downloading pulp-3.2.1-py3-none-any.whl.metadata (6.9 kB)

Downloading pulp-3.2.1-py3-none-any.whl (16.4 MB)

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```

Installing collected packages: pulp

Successfully installed pulp-3.2.1

```
In [4]: # Create LP problem (maximize profit)
model = pulp.LpProblem("Production_Planning", pulp.LpMaximize)
```

```

# Decision variables
x = pulp.LpVariable('Product_A', lowBound=0, cat='Continuous')
y = pulp.LpVariable('Product_B', lowBound=0, cat='Continuous')

# Objective function
model += 40*x + 30*y, "Total_Profit"

# Constraints
model += 2*x + y <= 100, "Labor_Constraint"
model += x + y <= 80, "Material_Constraint"

# Solve
model.solve()

# Display results
print("Status:", pulp.LpStatus[model.status])
print("Produce Product A:", x.varValue)
print("Produce Product B:", y.varValue)
print("Maximum Profit:", pulp.value(model.objective))

```

Status: Optimal
 Produce Product A: 20.0
 Produce Product B: 60.0
 Maximum Profit: 2600.0

Insights

- The model recommends the optimal production quantities for Products A and B.
 - These values maximize total profit while respecting labor and material constraints.
 - If more labor or raw materials become available, the maximum profit could increase further.
- Visualization Code

```

In [8]: import matplotlib.pyplot as plt
import numpy as np

# Create range for plotting
x_vals = np.linspace(0, 60, 200)

# Constraints:
y1 = 100 - 2*x_vals      # from 2x + y <= 100
y2 = 80 - x_vals         # from x + y <= 80

# Feasible region (intersection of constraints)
plt.figure(figsize=(8,6))
plt.plot(x_vals, y1, label=r'2x + y ≤ 100 (Labor)')
plt.plot(x_vals, y2, label=r'x + y ≤ 80 (Material)')
plt.axhline(0, color='black')
plt.axvline(0, color='black')

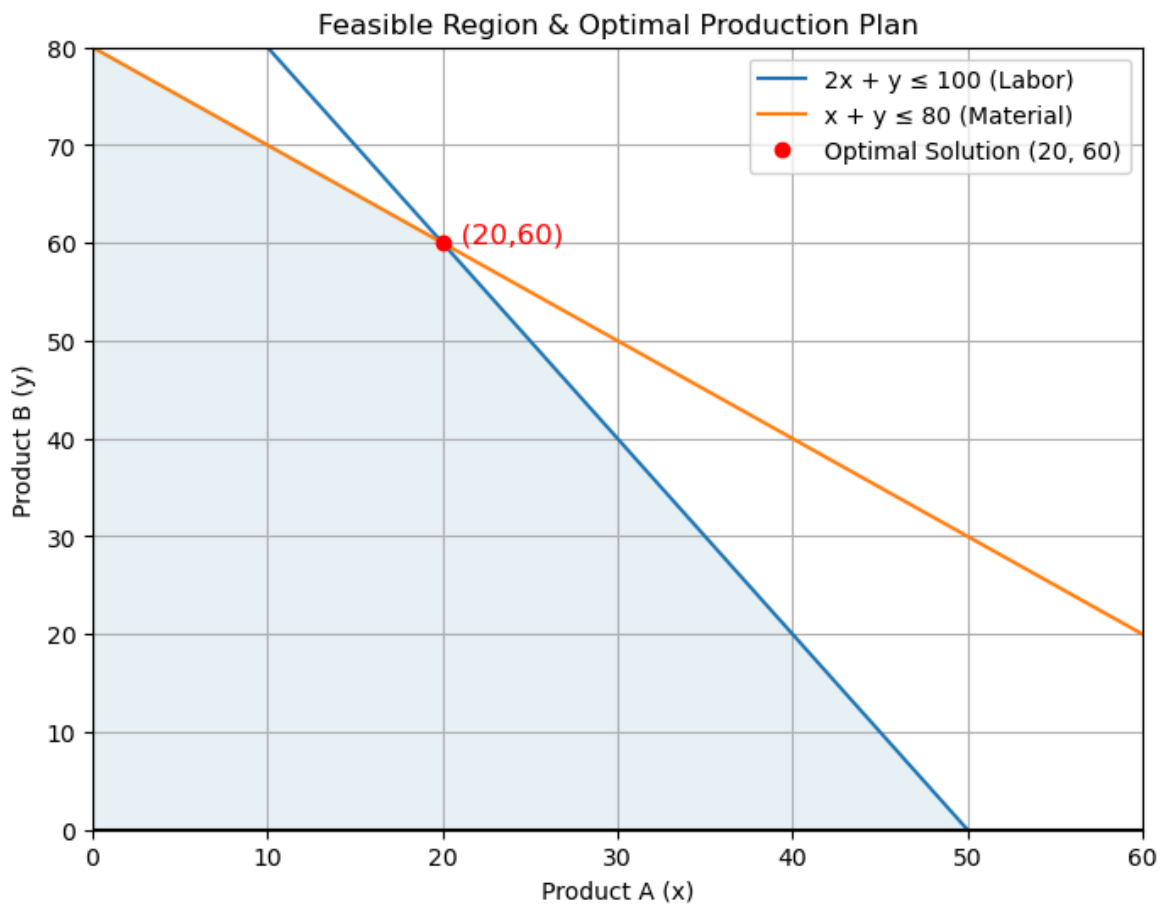
# Fill feasible region
y_min = np.minimum(y1, y2)
y_min = np.maximum(y_min, 0) # keep above 0
plt.fill_between(x_vals, y_min, 0, where=(y_min>0), alpha=0.3, color='lightblue')

# Plot optimal solution

```

```
plt.plot(20, 60, 'ro', label='Optimal Solution (20, 60)')
plt.text(20, 60, ' (20,60)', color='red', fontsize=12)

# Labels and Legend
plt.xlim(0, 60)
plt.ylim(0, 80)
plt.xlabel('Product A (x)')
plt.ylabel('Product B (y)')
plt.title('Feasible Region & Optimal Production Plan')
plt.legend()
plt.grid(True)
plt.show()
```



- Visualization with Corner Points

```
In [11]: import matplotlib.pyplot as plt
import numpy as np

# Constraint functions
x_vals = np.linspace(0, 60, 200)
y1 = 100 - 2*x_vals      # Labor constraint
y2 = 80 - x_vals         # Material constraint

# Feasible region boundaries
y_min = np.minimum(y1, y2)
y_min = np.maximum(y_min, 0)

# Calculate corner points
points = [
    (0, 0),              # Origin
    (0, min(100, 80)),   # y-axis intercept
```

```

    (50, 0),          # From  $2x + y = 100 \rightarrow (50,0)$ 
    (80, 0),          # From  $x + y = 80 \rightarrow (80,0)$  (not feasible due to Labor)
    (20, 60)          # Intersection of both constraints
]

# Filter feasible points
feasible_points = []
for x, y in points:
    if 2*x + y <= 100 and x + y <= 80 and x >= 0 and y >= 0:
        feasible_points.append((x, y))

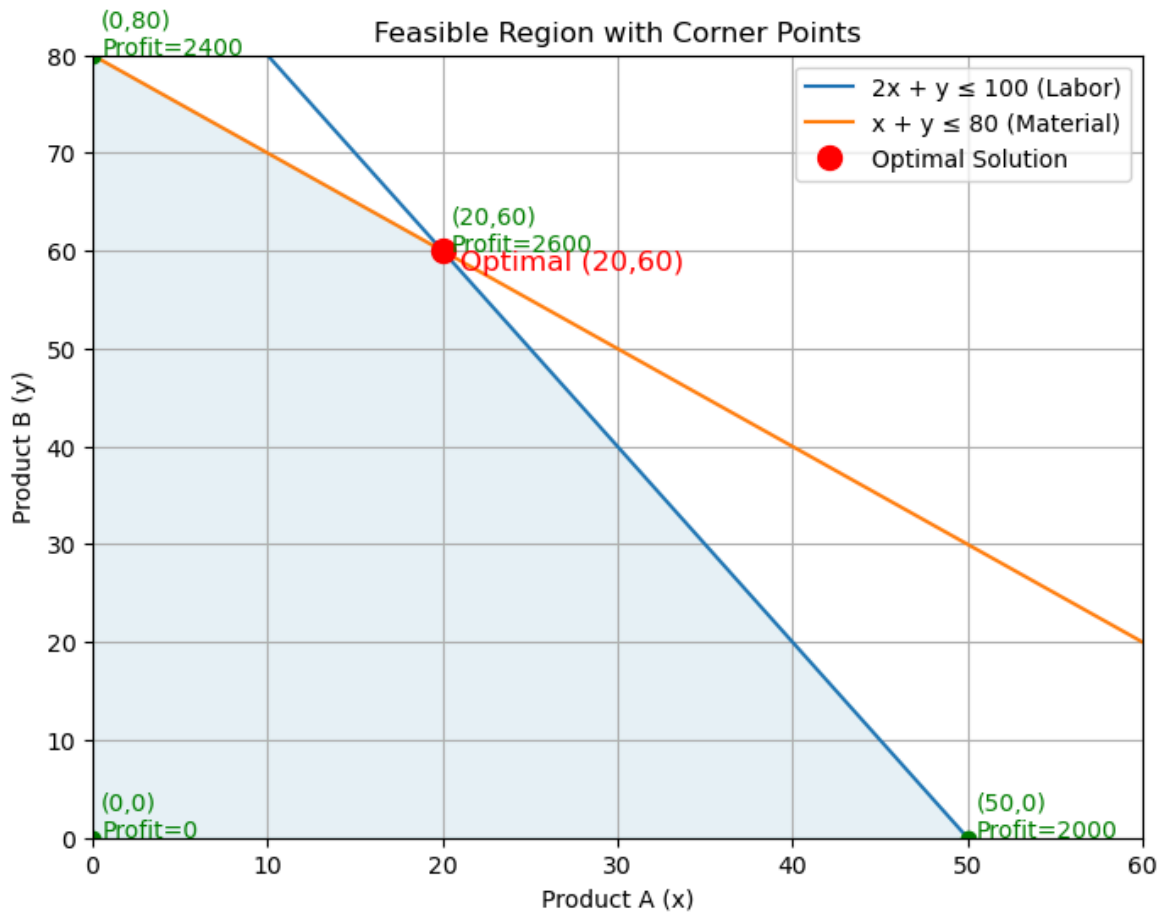
# Plot feasible region
plt.figure(figsize=(8,6))
plt.plot(x_vals, y1, label=r' $2x + y \leq 100$  (Labor)')
plt.plot(x_vals, y2, label=r' $x + y \leq 80$  (Material)')
plt.fill_between(x_vals, y_min, 0, where=(y_min>0), alpha=0.3, color='lightblue')

# Plot feasible points
for (x, y) in feasible_points:
    profit = 40*x + 30*y
    plt.plot(x, y, 'go')
    plt.text(x+0.5, y, f'({x:.0f},{y:.0f})\nProfit={profit:.0f}', fontsize=10, color='black')

# Highlight optimal solution
opt_x, opt_y = 20, 60
plt.plot(opt_x, opt_y, 'ro', markersize=10, label='Optimal Solution')
plt.text(opt_x+1, opt_y-2, f'Optimal ({opt_x},{opt_y})', color='red', fontsize=12)

# Labels and Legend
plt.xlim(0, 60)
plt.ylim(0, 80)
plt.xlabel('Product A (x)')
plt.ylabel('Product B (y)')
plt.title('Feasible Region with Corner Points')
plt.legend()
plt.grid(True)
plt.show()

```



- Table of Corner Points & Profits

```
In [14]: import pandas as pd

# Define feasible corner points again
corner_points = [
    (0, 0),
    (0, 80),      # from material constraint
    (50, 0),      # from labor constraint
    (20, 60)      # intersection of both constraints
]

# Filter feasible ones
feasible = []
for x, y in corner_points:
    if 2*x + y <= 100 and x + y <= 80 and x >= 0 and y >= 0:
        profit = 40*x + 30*y
        feasible.append([x, y, profit])

# Create DataFrame
df = pd.DataFrame(feasible, columns=["Product A (x)", "Product B (y)", "Profit"])

# Highlight the maximum profit row
df.style.highlight_max(subset=["Profit"], color="lightgreen")
```




Out[14]:

	Product A (x)	Product B (y)	Profit
0	0	0	0
1	0	80	2400
2	50	0	2000
3	20	60	2600



Conclusion

The Production Planning Optimization model successfully identified the best allocation of resources to maximize profit.

-  **Optimal Solution:** Produce **20 units of Product A** and **60 units of Product B**
-  **Maximum Profit:** **\$2600**
-  **Resource Utilization:**
 - Labor: Fully used (100 hours)
 - Raw Materials: Fully used (80 units)

Both constraints were binding, meaning resources were fully utilized without any wastage.

The feasible region and the profit table confirm that no other production plan yields a higher profit.



Key Insight:

To further increase profit, the company would need to expand its resource capacity — either by adding more labor hours or acquiring additional raw materials.

In []: