

Problem Definition: Transportation Optimization

A company needs to transport goods from **three warehouses** (W1, W2, W3) to **four retail stores** (S1, S2, S3, S4) while **minimizing transportation costs**. Each warehouse has a **limited supply**, and each store has a **specific demand**. The transportation cost per unit between each warehouse and store is given.

1. Transportation Cost (in \$ per unit)

The cost to transport one unit of goods from a warehouse to a store is given in the table below:

	S1	S2	S3	S4
W1	4	8	8	6
W2	6	7	3	5
W3	9	4	7	3

For example, the cost of transporting one unit from **W1 to S1** is **\$4**, while from **W3 to S2** is **\$4**.

Mathematical Formulation:

Decision Variables:

Let  $x_{ij}$  represent the **number of units transported from warehouse  $i$  to store  $j$** :

- $x_{11}$  = Units transported from **W1 → S1**
- $x_{12}$  = Units transported from **W1 → S2**
- $x_{13}$  = Units transported from **W1 → S3**
- $x_{14}$  = Units transported from **W1 → S4**
- ... (similar variables for W2 and W3)

Objective Function:

Minimize total transportation cost:

$$\min Z = 4x_{11} + 8x_{12} + 8x_{13} + 6x_{14} + 6x_{21} + 7x_{22} + 3x_{23} + 5x_{24} + 9x_{31} + 4x_{32} + 7x_{33} + 3x_{34}$$

Constraints:

1. Supply Constraints (Each warehouse has a limited supply):

- $x_{11} + x_{12} + x_{13} + x_{14} \leq 100$  (Supply from W1)
- $x_{21} + x_{22} + x_{23} + x_{24} \leq 120$  (Supply from W2)
- $x_{31} + x_{32} + x_{33} + x_{34} \leq 130$  (Supply from W3)

2. Demand Constraints (Each store must receive the required quantity):

- $x_{11} + x_{21} + x_{31} = 80$  (Demand for S1)
- $x_{12} + x_{22} + x_{32} = 70$  (Demand for S2)
- $x_{13} + x_{23} + x_{33} = 90$  (Demand for S3)
- $x_{14} + x_{24} + x_{34} = 110$  (Demand for S4)

3. Non-Negativity Constraints:

$$x_{ij} \geq 0 \quad \forall i, j$$

Brief Explanation of the Code

This Python script solves a **transportation optimization problem** using **Linear Programming (LP)** with the **PuLP** library. The goal is to **minimize transportation costs** while satisfying supply and demand constraints.

Libraries Used:

- **pulp**: A Python library for Linear Programming (LP) and Mixed-Integer Programming (MIP).

Key Functions & Their Purpose:

1. **LpProblem("Transportation\_Optimization", LpMinimize)**
  - Creates an LP problem named *"Transportation\_Optimization"* with a **minimization** objective.
2. **LpVariable(f"x\_{i,j}", lowBound=0)**
  - Defines decision variables  $x_{i,j}$ , representing **units transported** from warehouse  $i$  to store  $j$ .
  - `lowBound=0` ensures variables are **non-negative**.

### 3. Objective Function:

```
model += sum(x[i, j] for i in range(1, 4)) == demand[j], f"Demand_Constraint_S{j}"
```

Minimizes **total transportation cost** using a summation of cost per unit multiplied by transported units.

### 4. Constraints:

- **Supply Constraints** ensure each warehouse does not exceed its available supply.
- **Demand Constraints** ensure each store receives exactly its required amount.

### 5. model.solve()

- Solves the LP problem using an appropriate solver.

### 6. Printing Results:

- Displays **optimal transportation plan** and **minimum cost**.

### 7. Sensitivity Analysis (Shadow Prices & Slack):

```
# Sensitivity Analysis (Shadow Prices)
print("\nSensitivity Analysis (Shadow Prices):")
for constraint in model.constraints:
    print(f"{constraint}: Shadow Price = {model.constraints[constraint].pi}, Slack = {model.constraints[constraint].slack}")
```

• **Shadow price:** Impact of a unit increase in supply/demand.

• **Slack:** Unused capacity in a constraint.

### Results:

GLPSOL---GLPK LP/MIP Solver 5.0

Parameter(s) specified in the command line:

```
--cpxlp /var/folders/4n/6ydjnj8d16x_m4by160ndjw80000gn/T/d69e0c74262d456d9ad4c475dfcc2757-pulp.lp
```

```
-o /var/folders/4n/6ydjnj8d16x_m4by160ndjw80000gn/T/d69e0c74262d456d9ad4c475dfcc2757-pulp.sol
```

Reading problem data from '/var/folders/4n/6ydjnj8d16x\_m4by160ndjw80000gn/T/d69e0c74262d456d9ad4c475dfcc2757-pulp.lp'...

7 rows, 12 columns, 24 non-zeros

13 lines were read

GLPK Simplex Optimizer 5.0

7 rows, 12 columns, 24 non-zeros

Preprocessing...

7 rows, 12 columns, 24 non-zeros

Scaling...

A: min|a<sub>ij</sub>| = 1.000e+00 max|a<sub>ij</sub>| = 1.000e+00 ratio = 1.000e+00

Problem data seem to be well scaled

Constructing initial basis...

Size of triangular part is 7

0: obj = 1.960000000e+03 inf = 2.200e+02 (1)

4: obj = 1.510000000e+03 inf = 0.000e+00 (0)

\* 8: obj = 1.320000000e+03 inf = 0.000e+00 (0)

OPTIMAL LP SOLUTION FOUND

Time used: 0.0 secs

Memory used: 0.0 Mb (33757 bytes)

Writing basic solution to '/var/folders/4n/6ydjnj8d16x\_m4by160ndjw80000gn/T/d69e0c74262d456d9ad4c475dfcc2757-pulp.sol'...

Optimal Transportation Plan:

Warehouse 1 to Store 1: 80.0 units

Warehouse 1 to Store 2: 0.0 units

Warehouse 1 to Store 3: 0.0 units

Warehouse 1 to Store 4: 20.0 units

Warehouse 2 to Store 1: 0.0 units

Warehouse 2 to Store 2: 0.0 units

Warehouse 2 to Store 3: 90.0 units

Warehouse 2 to Store 4: 30.0 units

Warehouse 3 to Store 1: 0.0 units

Warehouse 3 to Store 2: 70.0 units

Warehouse 3 to Store 3: 0.0 units

Warehouse 3 to Store 4: 60.0 units

Minimum Transportation Cost: \$1320.0

Sensitivity Analysis (Shadow Prices):

Supply\_Constraint\_W1: Shadow Price = None, Slack = None

Supply\_Constraint\_W2: Shadow Price = None, Slack = None

Supply\_Constraint\_W3: Shadow Price = None, Slack = None

Demand\_Constraint\_S1: Shadow Price = None, Slack = None

Demand\_Constraint\_S2: Shadow Price = None, Slack = None

Demand\_Constraint\_S3: Shadow Price = None, Slack = None

Demand\_Constraint\_S4: Shadow Price = None, Slack = None