

hw1

September 26, 2024

1 HW Assignment 1, CMOR3 464 INDE 543 - MANUFACTURING PROCESSES AND SYSTEMS

2 Line Setup LP Problem Formulation

2.1 Problem Description

We are tasked with optimizing a production line to meet daily demand under given constraints. The problem includes calculating production with different numbers of workstations, overtime hours, and efficiency levels to determine the most cost-effective setup.

2.1.1 Data Provided:

Costs:

- Unit Material Costs: \$30
- Hourly Wage: \$60
- Overtime Hourly Wage: \$90

Current Production:

- Current Production: 315 units/day
- Daily Demand: 420 units/day

Other Costs:

- One Time Training Costs: \$10,000

2.1.2 Working Schedule:

- **Workday Start:** 8:00 AM
- **Workday End:** 4:00 PM
- **Overtime Max:** 7:00 PM
- **Normal Work Hours:** 8 hours
- **Maximum Overtime Hours:** 11 hours (including 1 hour break)
- **Break Time:** 1 hour

2.1.3 Line Setup Information:

# of Workstations	Overtime (hrs/day)	Capacity (u/day)	Efficiency
6	0		
6	3		
7	0		
7	3		
8	0		
8	3		
9	0		
9	3		

2.1.4 Activities Data:

Activities	Time (seconds)	Immediate Predecessors	Workers Allotted
1	30	n/a	1
2	50	n/a	2
3	40	1	1
4	50	1	3
5	20	2	2
6	10	3, 4	4
7	10	4, 5	5
8	20	2	3
9	10	6	4
10	40	9	5
11	20	7	5
12	30	7	7
13	50	9	4
14	50	10	6
15	10	11	6
16	40	8, 12	7

2.1.5 Working Hours:

Metric	Value
Beginning Work Time	8:00 AM
End Work Time	4:00 PM
Overtime Max	7:00 PM
Normal Hours	8 hours
Maximum Overtime Hours	11 hours
Break Time	1 hour
Normal Productive Time	7 hours
Maximal Overtime Productive Time	10 hours

2.2 Objective:

The objective is to write Linear Programs (LPs) that optimize the number of workstations. We can use Type 2 Model from the textbook:

2.3 Type 2 Assembly Line Balancing Model

2.3.1 Variables:

- t_i : Time required for activity A_i , where $i = 1, 2, \dots, N$
 - $N = 16$ for this example
- S : The number of workstations, where $S = 1, 2, \dots, n$
 - n is variable, but it'll be $n \in \{6, 7, \dots, 9\}$
- C : Cycle time of the line (maximum time allowed per workstation)
- x_{is} : A binary variable indicating whether activity A_i is assigned to station S

2.3.2 Objective:

Minimize cycle time C , i.e., the time required to complete all tasks assigned to a workstation.

$$\text{Min } C$$

2.3.3 Constraints:

1. Cycle Time Constraint:

The total time assigned to any workstation must not exceed the cycle time C :

$$\sum_i x_{is} \cdot t_i \leq C, \quad \forall s \in S$$

2. Assignment Constraint:

Each activity must be assigned to exactly one workstation:

$$\sum_s x_{is} = 1, \quad \forall i \in N$$

3. Precedence Constraint:

Precedence relationships between activities must be respected. If activity A_i must precede activity A_j , then the assignment must reflect that:

$$\sum_{s=1}^S x_{is} \cdot s \leq \sum_{s=1}^S x_{js} \cdot s, \quad \forall (i, j) \text{ where } A_i \text{ precedes } A_j$$

2.3.4 Problem Summary:

- The goal is to balance the assembly line by minimizing the cycle time while respecting the constraints of workstation assignment and task precedence.
- Each workstation must process tasks in a way that satisfies the overall demand, ensuring efficiency and minimizing idle time.

3 Initialize the model

import gurobipy as gp from gurobipy import GRB

[224]: *# Dictionary to store time for each job*

```
job_times = {  
    1: 30,  
    2: 50,  
    3: 40,  
    4: 50,  
    5: 20,  
    6: 10,  
    7: 10,  
    8: 20,  
    9: 10,  
    10: 40,  
    11: 20,  
    12: 30,  
    13: 50,  
    14: 50,  
    15: 10,  
    16: 40  
}
```

[225]: `import numpy as np`

Define the precedence dictionary based on your table (activity -> list of ↪ immediate predecessors)

```
precedence_dict = {  
    1: [],  
    2: [],  
    3: [1],  
    4: [1],  
    5: [2],  
    6: [3, 4],  
    7: [4, 5],  
    8: [2],  
    9: [6],  
    10: [9],  
    11: [7],  
    12: [7],  
    13: [9],  
    14: [10],  
    15: [11],  
    16: [8, 12]  
}
```

Initialize an empty precedence matrix with zeros (16x16)

```

precedence_matrix = np.zeros((num_jobs, num_jobs), dtype=int)

# Fill the precedence matrix based on the precedence_dict
for activity, predecessors in precedence_dict.items():
    for predecessor in predecessors:
        precedence_matrix[activity - 1][predecessor - 1] = 1

# Print the precedence matrix to verify
print("Precedence Matrix:")
print(precedence_matrix)

```

Precedence Matrix:

```

[[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
 [1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
 [1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
 [0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
 [0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0]
 [0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0]
 [0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0]
 [0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0]
 [0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0]
 [0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0]
 [0 0 0 0 0 0 0 1 0 0 0 0 1 0 0 0]]

```

[226]:

```

[231]: # Define the number of jobs
num_jobs = 16

# List of different numbers of stations to try
num_stations_list = [6, 7, 8, 9]

# Dictionary to store results for each number of stations
results = {}

for num_stations in num_stations_list:
    print(f"Solving for {num_stations} stations...")

    # Create a new Gurobi model
    model = gp.Model("assembly_line_balancing")

    C = model.addVar(name="C")
    model.update()

```

```

model.setObjective(C, GRB.MINIMIZE)

# Create a matrix of Gurobi binary variables  $X_{is}$  for  $i$  in  $[1, 16]$  and  $s$ 
↪in  $[1, \text{num\_stations}]$ 
X = [[model.addVar(vtype=GRB.BINARY, name="X_{}_{}".format(i+1, s+1))
      for s in range(num_stations)]
      for i in range(num_jobs)]

# Add cycle time constraints to the model
for s in range(num_stations):
    model.addConstr(
        gp.quicksum(X[i][s] * job_times[i+1] for i in range(num_jobs)) <= C,
        name=f"cycle_time_station_{s+1}"
    )

# Add assignment constraints to the model
for i in range(num_jobs):
    model.addConstr(
        gp.quicksum(X[i][s] for s in range(num_stations)) == 1,
        name=f"assignment_job_{i+1}"
    )

# Add precedence constraints to the model
for i in range(num_jobs):
    for j in range(num_jobs):
        if precedence_matrix[i][j] == 1:
            for k in range(num_stations):
                lhs = X[i][k]
                rhs = gp.quicksum(X[j][z] for z in range(k+1))
                model.addConstr(lhs <= rhs,
↪name=f"precedence_{i+1}_{j+1}_station_{k+1}")

# Update the model to integrate the new constraints
model.update()

# Optimize the model
model.optimize()

# Check if the model has an optimal solution
if model.status == GRB.OPTIMAL:
    # Create a dictionary to store the assignment of activities to stations
    station_assignment = {s+1: [] for s in range(num_stations)}

    # Iterate over the variables to get the assignment
    for i in range(num_jobs):
        for s in range(num_stations):

```

```

        if X[i][s].x > 0.5: # If the variable is 1 (assigned)
            station_assignment[s+1].append(i+1)

# Store the results
results[num_stations] = {
    'cycle_time': C.x,
    'station_assignment': station_assignment
}

# Print the assignment
print(f"Optimal cycle time for {num_stations} stations: {C.x}")
for station, activities in station_assignment.items():
    print(f"Station {station}: Activities {activities}, Total Time_
↪{sum(job_times[activity] for activity in activities)} seconds")
else:
    print(f"No optimal solution found for {num_stations} stations.")

# Print the results summary
print("\nSummary of results:")
for num_stations, result in results.items():
    print(f"{num_stations} stations: Cycle time = {result['cycle_time']}")

```

Solving for 6 stations...

Gurobi Optimizer version 11.0.1 build v11.0.1rc0 (mac64[arm] - Darwin 23.6.0 23G93)

CPU model: Apple M2 Max

Thread count: 12 physical cores, 12 logical processors, using up to 12 threads

Optimize a model with 124 rows, 97 columns and 657 nonzeros

Model fingerprint: 0xe1af4366

Variable types: 1 continuous, 96 integer (96 binary)

Coefficient statistics:

Matrix range [1e+00, 5e+01]

Objective range [1e+00, 1e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Found heuristic solution: objective 410.0000000

Presolve removed 17 rows and 0 columns

Presolve time: 0.00s

Presolved: 107 rows, 97 columns, 470 nonzeros

Variable types: 0 continuous, 97 integer (96 binary)

Root relaxation: objective 8.000000e+01, 64 iterations, 0.00 seconds (0.00 work units)

Nodes	Current Node	Objective Bounds	Work
Expl Unexpl	Obj Depth IntInf	Incumbent BestBd Gap	It/Node Time

	0	0	80.000000	0	21	410.000000	80.000000	80.5%	-	0s
H	0	0				120.00000000	80.000000	33.3%	-	0s
H	0	0				100.00000000	80.000000	20.0%	-	0s
H	0	0				90.00000000	80.000000	11.1%	-	0s
H	0	0				80.00000000	80.000000	0.00%	-	0s
	0	0	80.000000	0	27	80.000000	80.000000	0.00%	-	0s

Cutting planes:

Gomory: 5
 Cover: 5
 Implied bound: 1
 Clique: 6
 MIR: 4
 StrongCG: 2
 Zero half: 3
 RLT: 6
 Relax-and-lift: 1
 BQP: 2

Explored 1 nodes (165 simplex iterations) in 0.02 seconds (0.00 work units)
 Thread count was 12 (of 12 available processors)

Solution count 5: 80 90 100 ... 410

Optimal solution found (tolerance 1.00e-04)
 Best objective 8.0000000000000e+01, best bound 8.0000000000000e+01, gap 0.00000%
 Optimal cycle time for 6 stations: 80.0
 Station 1: Activities [1, 4], Total Time 80 seconds
 Station 2: Activities [2, 5, 7], Total Time 80 seconds
 Station 3: Activities [3, 6, 8, 9], Total Time 80 seconds
 Station 4: Activities [12, 13], Total Time 80 seconds
 Station 5: Activities [10, 16], Total Time 80 seconds
 Station 6: Activities [11, 14, 15], Total Time 80 seconds
 Solving for 7 stations...
 Gurobi Optimizer version 11.0.1 build v11.0.1rc0 (mac64[arm] - Darwin 23.6.0 23G93)

CPU model: Apple M2 Max
 Thread count: 12 physical cores, 12 logical processors, using up to 12 threads

Optimize a model with 142 rows, 113 columns and 826 nonzeros
 Model fingerprint: 0x8d604f7e
 Variable types: 1 continuous, 112 integer (112 binary)
 Coefficient statistics:

Matrix range [1e+00, 5e+01]
 Objective range [1e+00, 1e+00]
 Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]
 Found heuristic solution: objective 220.0000000
 Presolve removed 24 rows and 4 columns
 Presolve time: 0.00s
 Presolved: 118 rows, 109 columns, 567 nonzeros
 Variable types: 0 continuous, 109 integer (108 binary)

 Root relaxation: objective 6.857143e+01, 74 iterations, 0.00 seconds (0.00 work units)

Nodes		Current Node			Objective Bounds			Work		
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node	Time	
	0	0	68.57143	0	14	220.00000	68.57143	68.8%	-	0s
H	0	0				80.0000000	68.57143	14.3%	-	0s
H	0	0				70.0000000	70.00000	0.00%	-	0s
	0	0	70.00000	0	26	70.00000	70.00000	0.00%	-	0s

Cutting planes:

Gomory: 1
 Cover: 6
 Clique: 2
 MIR: 4
 GUB cover: 1
 Zero half: 1
 RLT: 4
 BQP: 1

Explored 1 nodes (376 simplex iterations) in 0.02 seconds (0.01 work units)
 Thread count was 12 (of 12 available processors)

Solution count 3: 70 80 220

Optimal solution found (tolerance 1.00e-04)
 Best objective 7.000000000000e+01, best bound 7.000000000000e+01, gap 0.0000%
 Optimal cycle time for 7 stations: 70.0
 Station 1: Activities [2, 5], Total Time 70 seconds
 Station 2: Activities [1, 3], Total Time 70 seconds
 Station 3: Activities [4, 6, 9], Total Time 70 seconds
 Station 4: Activities [7, 13], Total Time 60 seconds
 Station 5: Activities [10, 12], Total Time 70 seconds
 Station 6: Activities [11, 14], Total Time 70 seconds
 Station 7: Activities [8, 15, 16], Total Time 70 seconds
 Solving for 8 stations...
 Gurobi Optimizer version 11.0.1 build v11.0.1rc0 (mac64[arm] - Darwin 23.6.0 23G93)

CPU model: Apple M2 Max

Thread count: 12 physical cores, 12 logical processors, using up to 12 threads

Optimize a model with 160 rows, 129 columns and 1012 nonzeros

Model fingerprint: 0xa29d5b87

Variable types: 1 continuous, 128 integer (128 binary)

Coefficient statistics:

Matrix range [1e+00, 5e+01]

Objective range [1e+00, 1e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Found heuristic solution: objective 300.0000000

Presolve removed 21 rows and 2 columns

Presolve time: 0.00s

Presolved: 139 rows, 127 columns, 725 nonzeros

Variable types: 0 continuous, 127 integer (126 binary)

Root relaxation: objective 6.000000e+01, 85 iterations, 0.00 seconds (0.00 work units)

Nodes		Current Node			Objective Bounds			Work	
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node	Time
	0	0	60.00000	0	38	300.00000	60.00000	80.0%	- 0s
H	0	0				110.0000000	60.00000	45.5%	- 0s
H	0	0				90.0000000	60.00000	33.3%	- 0s
H	0	0				80.0000000	60.00000	25.0%	- 0s
H	0	0				70.0000000	60.00000	14.3%	- 0s
	0	0	70.00000	0	39	70.00000	70.00000	0.00%	- 0s

Cutting planes:

Gomory: 12

Cover: 7

Clique: 11

MIR: 5

StrongCG: 1

GUB cover: 1

Zero half: 11

RLT: 8

Relax-and-lift: 1

BQP: 2

Explored 1 nodes (217 simplex iterations) in 0.02 seconds (0.01 work units)

Thread count was 12 (of 12 available processors)

Solution count 5: 70 80 90 ... 300

Optimal solution found (tolerance 1.00e-04)

Best objective 7.000000000000e+01, best bound 7.000000000000e+01, gap 0.0000%

Optimal cycle time for 8 stations: 70.0
 Station 1: Activities [1, 3], Total Time 70 seconds
 Station 2: Activities [4, 6, 9], Total Time 70 seconds
 Station 3: Activities [10], Total Time 40 seconds
 Station 4: Activities [14], Total Time 50 seconds
 Station 5: Activities [2, 5], Total Time 70 seconds
 Station 6: Activities [13], Total Time 50 seconds
 Station 7: Activities [7, 8, 11, 15], Total Time 60 seconds
 Station 8: Activities [12, 16], Total Time 70 seconds
 Solving for 9 stations...
 Gurobi Optimizer version 11.0.1 build v11.0.1rc0 (mac64[arm] - Darwin 23.6.0 23G93)

CPU model: Apple M2 Max
 Thread count: 12 physical cores, 12 logical processors, using up to 12 threads

Optimize a model with 178 rows, 145 columns and 1215 nonzeros

Model fingerprint: 0xb4366da3

Variable types: 1 continuous, 144 integer (144 binary)

Coefficient statistics:

Matrix range [1e+00, 5e+01]

Objective range [1e+00, 1e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Found heuristic solution: objective 260.0000000

Presolve removed 21 rows and 2 columns

Presolve time: 0.00s

Presolved: 157 rows, 143 columns, 890 nonzeros

Variable types: 0 continuous, 143 integer (142 binary)

Root relaxation: objective 5.333333e+01, 120 iterations, 0.00 seconds (0.00 work units)

Nodes		Current Node				Objective Bounds			Work	
Expl	Unexpl	Obj	Depth	IntInf		Incumbent	BestBd	Gap	It/Node	Time
	0	0	53.33333	0	25	260.00000	53.33333	79.5%	-	0s
H	0	0				120.0000000	53.33333	55.6%	-	0s
H	0	0				100.0000000	53.33333	46.7%	-	0s
H	0	0				90.0000000	53.33333	40.7%	-	0s
H	0	0				80.0000000	53.33333	33.3%	-	0s
	0	0	60.00000	0	8	80.00000	60.00000	25.0%	-	0s
H	0	0				70.0000000	60.00000	14.3%	-	0s
H	0	0				60.0000000	60.00000	0.00%	-	0s
	0	0	60.00000	0	8	60.00000	60.00000	0.00%	-	0s

Cutting planes:

Gomory: 2

Cover: 6
Implied bound: 4
Clique: 1
MIR: 2
RLT: 4

Explored 1 nodes (366 simplex iterations) in 0.02 seconds (0.01 work units)
Thread count was 12 (of 12 available processors)

Solution count 7: 60 70 80 ... 260

Optimal solution found (tolerance 1.00e-04)
Best objective 6.000000000000e+01, best bound 6.000000000000e+01, gap 0.0000%
Optimal cycle time for 9 stations: 60.0
Station 1: Activities [2], Total Time 50 seconds
Station 2: Activities [1, 8], Total Time 50 seconds
Station 3: Activities [4], Total Time 50 seconds
Station 4: Activities [3, 6, 9], Total Time 60 seconds
Station 5: Activities [5, 7, 12], Total Time 60 seconds
Station 6: Activities [10, 11], Total Time 60 seconds
Station 7: Activities [14, 15], Total Time 60 seconds
Station 8: Activities [13], Total Time 50 seconds
Station 9: Activities [16], Total Time 40 seconds

Summary of results:
6 stations: Cycle time = 80.0
7 stations: Cycle time = 70.0
8 stations: Cycle time = 70.0
9 stations: Cycle time = 60.0

```
[233]: import csv

# Define the file path
output_file_path = '/Users/warrenweissbluth/Documents/
↳CMOR-464-INDE-543-MANUFACTURING-PROCESSES-AND-SYSTEMS/results.csv'

# Write the results to the CSV file
with open(output_file_path, 'w', newline='') as file:
    writer = csv.writer(file)

    # Write the header
    writer.writerow(["Number of Stations", "Cycle Time", "Station",
↳"Activities", "Total Time (seconds)"])

    # Write the data
    for num_stations, result in results.items():
        for station, activities in result['station_assignment'].items():
```

```
        total_time = sum(job_times[activity] for activity in activities)
        writer.writerow([num_stations, result['cycle_time'], station,
↪activities, total_time])

print(f"Results written to {output_file_path}")
```

Results written to
/Users/warrenweissbluth/Documents/CMOR-464-INDE-543-MANUFACTURING-PROCESSES-AND-
SYSTEMS/results.csv