# Discrete Optimization Project Examination Timetabling Problem

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## Introduction

#### **Introduction** - Data

```
exams: E = \{e1, e2, ..., en\}
```

students:  $S = \{s1, s2, ..., sm\}$ . Each student is enrolled in a non-empty subset of exams.

time slots: Ts =  $\{t1 \le t2 \le ... \le tT\}$ 

ne1,e2 = number of students enrolled in both e1 and e2; if ne1,e2 > 0 then e1 and e2 are conflicting

#### **Introduction** - Constraints

each exam is scheduled exactly once during the examination period;

two exams e1, e2  $\in$  E are called conflicting if they have at least one student enrolled in both, i.e., if ne1,e2 > 0.

two conflicting exams are not scheduled in the same time-slot;

the total penalty resulting from the created timetable is minimized

given two exams e1, e2  $\in$  E scheduled at distance i of time-slots, with 1  $\le$  i  $\le$  5, the relative penalty is  $2^{(5-i)} \cdot (\text{ne1,e2}) / |S|$ .

## **Basic Model**

#### **Basic Model** - Variables

 $X_{ij} = \text{student } s_i \text{ is enrolled in exam } e_j \text{ , } X_{ij} \in \{0, 1\}, \quad \exists \ s_i \in S, \quad e_j \in E$ 

```
x = {}
for exam_id in exams:
    for time_slot in range(1, num_time_slots + 1):
        x[exam_id, time_slot] = model.addVar(vtype=GRB.BINARY) # 1 if exam e is scheduled in timeslot t, 0 otherwise
```

#### **Basic Model** - Variables

```
Y_{ij} = \text{exam } e_i \text{ takes place at timeslot } t_j, Y_{ij} \in \{0, 1\}, \quad \exists \ e_i \in E, \quad t_j \in T
```

```
y = {}
penalty_distance = 5  # Maximum time slot difference for penalty calculation

for distance in range(1, penalty_distance + 1):
    y[exam1, exam2, distance] = model.addVar(vtype=GRB.BINARY)
```

#### **Basic Model** - Constraints

```
\sum_{i=1}^{n} \sum_{j=1}^{n} Y_{ij} = 1: each exam is scheduled exactly once during the examination period (= sum of all the timeslots).
```

```
# Basic Constraint: Ensure that each exam is scheduled exactly once
for exam_id in exams:
    model.addConstr(
        gp.quicksum(x[exam_id, t] for t in range(1, num_time_slots + 1)) == 1
    )
```

#### **Basic Model** - Constraints

```
\sum\limits_{i=m,p}Y_{ij}=1 : conflicting exams e_m and e_p can't take place in the same time-slots.
```

#### **Basic Model** - Constraints

By adding this additional constraint, the solutions found have lower penalty

```
# Basic Constraint: Ensure conflicting exam pairs are not scheduled together
for exam1, exam2 in conflicting_pairs:
   for time slot in range(1, num time slots + 1):
       model.addConstr(
           x[exam1, time slot] + x[exam2, time slot] <= 1
       v = \{\}
       penalty distance = 5  # Maximum time slot difference for penalty calculation
       for distance in range(1, penalty distance + 1):
           y[exam1, exam2, distance] = model.addVar(vtype=GRB.BINARY)
           # Add constraints to enforce the relationship between y and x variables
           for time_slot in range(1, num_time_slots - distance + 1):
               model.addConstr(
                   y[exam1, exam2, distance] >= x[exam1, time slot] + x[exam2, time slot + distance] - 1
```

### **Basic Model** - Objective Function

If 2 exams  $e_m$  and  $e_p$  are conflicting, we have:

$$Y_{mi} = 1$$
,  $\exists m \in E$ , for the first exam

$$Y_{p,j+k} = 1 \;\; \exists \; p \in E \; and \; p \neq m, \; \exists \; 1 \leq k < 5 \;$$
 , for the next exam and the

penalty is: 
$$Z_{m, p, j, j+k} = 2^{5-k} \cdot \frac{n_{e_{m,p}}}{|S|}$$

#### **Objective function**

$$\operatorname{Min} \sum_{i=0}^{n} \sum_{i=0}^{n} Z_{m,p,j,j+k}$$

#### **Basic Model** - Objective Function

```
# Calculate the objective function to optimize
objective expr = 0
for exam1 in exams:
    for exam2 in exams:
        if exam1 != exam2:
            shared students = calculate common enrollment(exam1,exam2)
            if shared students > 0:
                for t1 in range(1, num_time_slots + 1):
                    for i in range(1, 6):
                        t2 = (t1 + i) if (t1 + i) \le num time slots else (t1 + i - num time slots)
                        penalty = ((2 ** (5 - i)) * (shared students)) / len(students)
                        #print("penalty: " + str(penalty))
                        objective expr += penalty
# Set the objective to minimize the total penalty
model.setObjective(objective expr, GRB.MINIMIZE)
```

## Advanced Model + Advanced Constraints

#### **Advanced Model** - New Equity Measure

**Maximum distance**: This measure is the maximum number of time-slots between any two conflicting exams. A timetable with a high maximum distance is more equitable for students, as it means that they will have more time between exams.

#### **Advanced Model** - Implementation

```
# Add constraints for the maximum distance
for i in range(1, len(exams)):
    exam1 = exams[i]
    for j in range(i+1, len(exams)):
       exam2 = exams[j]
        shared students = len(exam to students.get(exam1, set()) & exam to students.get(exam2, set()))
        if shared students > 0:
            for t1 in range(1, num_time_slots + 1):
                for t2 in range(1, num_time_slots + 1):
                    model.addConstr(
                        z >= abs(t1 - t2) * x[exam1, t1] * x[exam2, t2]
```

#### **Advanced Model** - Objective Function

```
# Set the objective to minimize the total penalty while maximizing the maximum distance
penalty weight = 1 # Adjust this weight factor based on the importance of minimizing penalty
distance weight = 1 # Adjust this weight factor based on the importance of maximizing distance
objective expr = 0
for i in range(1, len(exams)):
    exam1 = exams[i]
    for j in range(i+1, len(exams)):
        if exam1 != exam2:
            shared students = len(exam to students.get(exam1, set()) & exam to students.get(exam2, set()))
            if shared students > 0:
                for t1 in range(1, num time slots + 1):
                    for t2 in range(1, num time slots + 1):
                        objective expr += penalty weight * (shared students / len(students)) * x[exam1, t1] * x[exam2, t2]
                        objective expr += distance weight * abs(t1 - t2) * x[exam1, t1] * x[exam2, t2] # Use + instead of -
# Set the objective in the model
model.setObjective(objective expr, GRB.MINIMIZE)
```

#### **Advanced Constraints** - Introduction

At most 3 consecutive time slots can have conflicting exams.

If two consecutive time slots contain conflicting exams, then no conflicting exam can be scheduled in the next 3 time slots.

Include a bonus profit each time no conflicting exams are scheduled for 6 consecutive time slots.

Change the constraints that impose that no conflicting exams can be scheduled in the same time slot. Instead, impose that at most 3 conflicting pairs can be scheduled in the same time slot.

At most 3 consecutive time slots can have conflicting exams.

```
# Advanced Constraint: At most 3 consecutive time slots with conflicts
for time_slot in range(1, num_time_slots - 2): # Update the range here
    model.addConstr(
        y[time_slot] + y[time_slot + 1] + y[time_slot + 2] <= 3
    )</pre>
```

If two consecutive time slots contain conflicting exams, then no conflicting exam can be scheduled in the next 3 time slots.

→ NOT IMPLEMENTED/DOESN'T WORK

Include a bonus profit each time no conflicting exams are scheduled for 6 consecutive time slots.

```
objective_expr = 0
# Advanced Constraint: Include a bonus profit each time no conflicting exams are scheduled for 6 consecutive time slots
bonus_reward = 100  # Adjust this value as needed
for time_slot in range(1, num_time_slots - 5):  # Update the range here
    model.addConstr(
        gp.quicksum(y[t] for t in range(time_slot, time_slot + 6)) <= 5  # At most one conflict in 6 consecutive time slots
    )
    objective_expr += bonus_reward * (1 - gp.quicksum(y[t] for t in range(time_slot, time_slot + 6)))</pre>
```

Change the constraints that impose that no conflicting exams can be scheduled in the same time slot. Instead, impose that at most 3 conflicting pairs can be scheduled in the same time slot.

```
# Advanced Constraint: At most 3 consecutive time slots with conflicts
for time_slot in range(1, num_time_slots - 2): # Update the range here
    model.addConstr(
        y[time_slot] + y[time_slot + 1] + y[time_slot + 2] <= 3
    )</pre>
```

## Results

#### **Results**

| instances   | basic model          | advanced model       |
|-------------|----------------------|----------------------|
| instance 01 | 187.702              | 186.216              |
| instance 02 | 53.095               | 52.049               |
| instance 03 | 59.575               | 53.20                |
| instance 04 | 12.988               | 12.745               |
| instance 05 | 27.211               | 33.738               |
| instance 06 | No feasible solution | No feasible solution |
| instance 07 | 18.902               | 18.902               |
| instance 08 | 45.175               | 45.345               |
| instance 09 | 22.009               | 21.599               |
| instance 10 | No feasible solution | No feasible solution |
| instance 11 | No feasible solution | No feasible solution |

No feasible solutions for all the instances when adding the additional constraints.