Academic Timetable Optimization Using Constraint Satisfaction Problem

Sistemi Intelligenti

Problems to solve

- Too much problems with the timetable because I have subjects from differents degrees and they usually overlap.
- Fixed schedules
- Need to attend as many classes as possible

Algorithm to obtain the best combination of classes?

Goals

- Maximize attendance without overlaps.
- Minimize travel between buildings.
- Compact schedules and reduce downtime.
- Evaluate all possible combinations.

What is a CSP?

Constraint Satisfaction Problem

Main features

- **★** Some variables can take different values.
- ★ There are constraints that define which combinations of values are valid.
- ★ The goal is to find a combination of values for the variables that satisfies all the constraints.

Variables and domains

```
subjects = {
   "Mobile App Dev": ["Mon 8-10", "Fri 14-16"],
   "Sistemi Intelligenti": ["Tue 10-12", "Thu 14-16"],
   "Sistemi Elettronici": ["Fri 8-10", "Thu 16-18"],
   "Elettronica Biomedica": ["Wed 10-12", "Wed 16-18"],
   "Applicazioni di Fisica": ["Mon 14-16", "Tue 14-16"],
   "Circuiti": ["Wed 8-10", "Thu 10-12"],
   "Data Science": ["Tue 8-10", "Thu 8-10"],
   "AI Fundamentals": ["Mon 10-12", "Wed 14-16"],
   "Embedded Systems": ["Tue 16-18", "Fri 10-12"],
   "Computer Networks": ["Mon 16-18", "Thu 16-18"],
   "Biomedical Signals": ["Tue 10-12", "Wed 16-18"],
   "Digital Control": ["Wed 14-16", "Fri 8-10"],
   "Physics Lab": ["Thu 8-10", "Fri 16-18"],
   "Digital Design": ["Tue 14-16", "Thu 10-12"],
    "Robotics": ["Wed 10-12", "Fri 14-16"]
```

Each subject is a variable and the available time slot is the domain

Restrictions between variables

Restrictions between variables

Graph of the campus



```
campus_graph = {
    "A": [("B", 7), ("C", 15), ("D", 12)],
    "B": [("A", 7), ("C", 5), ("D", 10)],
    "C": [("A", 15), ("B", 5), ("D", 6)],
    "D": [("A", 12), ("B", 10), ("C", 6)]
```

graph where the edges are the travel time and the nodes are the buildings

CSP modeling

```
problem = Problem()
for subject, slots in subjects.items():
   options = []
   if len(slots) > 1:
       options += [combo for combo in itertools.combinations(slots, 2) if no overlap(*combo)]
   options += [(slot,) for slot in slots]
   problem.addVariable(subject, options)
subject list = list(subjects.keys())
for i in range(len(subject_list)):
   for j in range(i + 1, len(subject_list)):
       s1, s2 = subject_list[i], subject_list[j]
       def constraint(a, b, s1=s1, s2=s2):
           for slot1 in a:
               for slot2 in b:
                   if not no overlap(slot1, slot2) or not no travel conflict(slot1, slot2, location by slot, campus graph):
                       return False
           return True
       problem.addConstraint(constraint, (s1, s2))
solutions = problem.getSolutions()
```

Is there only one valid solution?

CSPs can generate many valid combinations.
Each solution respects all constraints.
Not all are equally efficient or practical.

We use a heuristic to pick the best ones.

```
def evaluate schedule(solution):
   flat solution = {subj: slot for subj, slots in solution.items() for slot in (slots if isinstance(slots, tuple) else [slots])}
   daily slots = defaultdict(list)
   for subj, slot in flat solution.items():
       day, start, end = parse time slot(slot)
       daily slots[day].append((start, end, location by slot[slot]))
   total dead time = 0
   total moves = 0
   total classes attended = sum(len(slots) if isinstance(slots, tuple) else 1 for slots in solution.values())
   for day, blocks in daily_slots.items():
       blocks.sort()
       for i in range(len(blocks) - 1):
           , end1, loc1 = blocks[i]
           start2, _, loc2 = blocks[i + 1]
           dead time = start2 - end1
           total dead time += max(0, dead time)
           if loc1 != loc2:
               total moves += 1
   days_with_class = len(daily_slots)
   has friday free = "Fri" not in daily slots
```

Heuristic evaluator

Ranking

```
ranked = sorted(
    [evaluate_schedule(sol) for sol in solutions],
    key=lambda x: (-x['classes_attended'], x['dead_time'], x['moves'], x['days'], not x['bonus_friday_off'])
)
```

First Solution

```
Schedule #1:
Schedule Summary:
Mon:
 14-16 -> Applicazioni di Fisica
  16-18 -> Computer Networks
Tue:
  8-10 -> Data Science
  10-12 -> Biomedical Signals
Wed:
  8-10 -> Circuiti
  10-12 -> Robotics
  14-16 -> AI Fundamentals
  16-18 -> Elettronica Biomedica
Thu:
  8-10 -> Physics Lab
  10-12 -> Digital Design
  14-16 -> Sistemi Intelligenti
  16-18 -> Sistemi Elettronici
Fri:
  8-10 -> Digital Control
  10-12 -> Embedded Systems
  14-16 -> Mobile App Dev
```

Second Solution

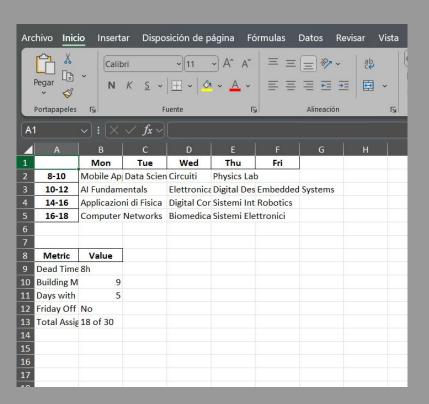
Schedule Summary: Mon: 8-10 -> Mobile App Dev 10-12 -> AI Fundamentals 14-16 -> Applicazioni di Fisi Tue: 8-10 -> Data Science Wed: 8-10 -> Circuiti 10-12 -> Elettronica Biomedic 14-16 -> Digital Control 16-18 -> Biomedical Signals Thu: 8-10 -> Physics Lab 10-12 -> Digital Design 14-16 -> Sistemi Intelligenti 16-18 -> Computer Networks Fri: 8-10 -> Sistemi Elettronici 10-12 -> Embedded Systems

14-16 -> Robotics

Third Solution

```
Schedule Summary:
Mon:
  8-10 -> Mobile App Dev
  10-12 -> AI Fundamentals
  14-16 -> Applicazioni di Fisica
 16-18 -> Computer Networks
Tue:
  8-10 -> Data Science
Wed:
  8-10 -> Circuiti
 10-12 -> Elettronica Biomedica
 14-16 -> Digital Control
 16-18 -> Biomedical Signals
Thu:
  8-10 -> Physics Lab
 10-12 -> Digital Design
  14-16 -> Sistemi Intelligenti
 16-18 -> Sistemi Elettronici
Fri:
 10-12 -> Embedded Systems
 14-16 -> Robotics
```

Excel



Conclusions

- ★ Schedules optimized using constraint programming.
- ★ Criteria: idle time, campus moves, active days, Friday off.
- ★ Results exported to Excel and shown in console.
- ★ Improves academic planning and student experience.

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