

40.016: The Analytics Edge

Week 12 Lecture 2 (Optional)

PRESCRIPTIVE ANALYTICS WITH JULIA (PART 2)

Term 5, 2022



SINGAPORE UNIVERSITY OF
TECHNOLOGY AND DESIGN

Outline

- 1 Capstone Allocation at SUTD
- 2 Implementation in Julia, with IJulia, Jupyter, and JuMP

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Capstone Allocation

Key features:

- Course started in 2015 and offered once a year
- Requires students from at least two disciplines (pillars)
- Projects sourced primarily from companies operating in Singapore
- Capstone office works with faculty and companies to scope the projects

Capstone Allocation (cont'd)

Challenges in allocating students to projects:

- **Efficiency** → Eliciting student preferences for projects and using them for allocation is critical for efficiency, rather than just allocating projects randomly
- **Fairness** → Students have equal chances in obtaining their preferred capstone
- **Multidisciplinary projects** → Every project must be multidisciplinary and involve students from at least two pillars
- **Flexibility** → It must be easy to incorporate any additional new constraints that might arise during the allocation process

Network representation

Representation of a Capstone project allocation instance with 6 students, 3 projects and 3 disciplines.

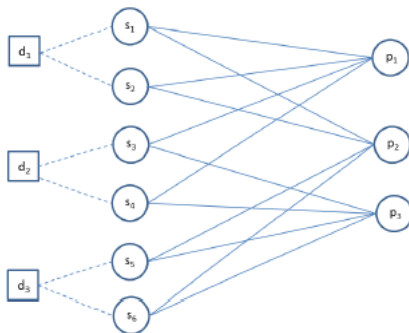


Figure: Source: Magnanti and Natarajan (2018).

Model inputs

- Network representation
- Lower and upper bounds on the number of students of each discipline (needed for each project)
- Student preferences for projects
 - The student preferences are converted to utility values on the arcs in the network where the value is set to K (e.g., $= 10$) for each student's topmost preferred project, $K - 1$ for their second most preferred project, down to 1 for the project ranked the lowest in their list

Capstone Allocation

Decision variables: Two sets of binary variables corresponding to

- Which project is assigned to a student
- Which project is launched

Objective function: Total utility, or efficiency, given by the sum of the utilities of the projects assigned to students (maximized)

Constraints:

- Each student must be allocated to a single project
- A student is assigned to a project only if a project is launched
- The number of students of the different disciplines in a project that is allocated lies between the prescribed lower and upper bounds

Formulation

- $UB(p, d)$ = Upper bound on the number of students of discipline d needed for project p
- $LB(p, d)$ = Lower bound on the number of students of discipline d needed for project p
- $util(s, p)$ = Utility of project p for student s in the network

As mentioned above, we have two sets of decision variables:

$$x_{sp} = \begin{cases} 1 & \text{if student } s \text{ is allocated to project } p \\ 0 & \text{otherwise} \end{cases}$$

$$y_p = \begin{cases} 1 & \text{if project } p \text{ is offered} \\ 0 & \text{otherwise} \end{cases}$$

Formulation (cont'd)

- \mathcal{S} : set of student nodes
- \mathcal{P} : set of project nodes
- \mathcal{D} : set of possible types (disciplines) of the students
- $\rightarrow G(\mathcal{S} \cup \mathcal{P}, \mathcal{E})$: bipartite graph
- $\mathcal{E} \subseteq \mathcal{S} \times \mathcal{P}$: set of undirected edges of the graph

Formulation (cont'd)

$$\max \sum_{(s,p) \in \mathcal{E}} \text{util}(s,p) x_{sp}$$

$$\text{s.t.} \quad \sum_{p \in \mathcal{P}: (s,p) \in \mathcal{E}} x_{sp} = 1 \quad \forall s \in \mathcal{S}$$

$$x_{sp} \leq y_p \quad \forall (s,p) \in \mathcal{E} \subseteq \mathcal{S} \times \mathcal{P}$$

$$\sum_{s \in \mathcal{S}: (s,p) \in \mathcal{E}, d(s)=d} x_{sp} \geq \text{LB}(p,d) y_p \quad \forall d \in \mathcal{D}(p), \forall p \in \mathcal{P}$$

$$\sum_{s \in \mathcal{S}: (s,p) \in \mathcal{E}, d(s)=d} x_{sp} \leq \text{UB}(p,d) y_p \quad \forall d \in \mathcal{D}(p), \forall p \in \mathcal{P}$$

$$x_{sp} \in \{0, 1\} \quad \forall (s,p) \in \mathcal{E} \subseteq \mathcal{S} \times \mathcal{P}$$

$$y_p \in \{0, 1\} \quad \forall p \in \mathcal{P}$$

Data

- `UpperBound.csv` [4,61]
- `LowerBound.csv` [4,61]
- `Rank.csv` (utility) [170,61]
- `Pillar.csv` (pillar of each student) [170,4]

So, we have 170 students, 61 potential projects, and 4 Pillars

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IJulia and Jupyter

How to launch Jupyter for Julia?

Option 1: Type `jupyter notebook` in the terminal

Option 2: Type the following commands in Julia (`julia>` prompt)

```
using IJulia  
notebook()
```

Workflow for creating a model

- Load JuMP and a solver / optimizer (e.g., GLPK)

```
import Pkg  
Pkg.add("JuMP")  
Pkg.add("GLPK")
```

see file `Packages.ipynb`.

- Create a model with the `Model()` function
- Add variables (`@variable()`)
- Add objective (`@objective()`)
- Add constraints, if any (`@constraint()`)

Workflow for solving a problem

- Use the `optimize!()` function
- Check the status of the solution (`termination_status()`)
- Check the obtained objective function (`objective_value()`)
- Check the obtained solution (`value.()`)

Implementation in Julia

Please refer to the file `capstone.ipynb`

References: Julia (with links)

- Julia <https://julialang.org/>
- IJ <https://github.com/JuliaLang/IJulia.jl>
- IJulia <https://julialang.github.io/IJulia.jl/stable/manual/installation/>
- JuMP <https://jump.dev/JuMP.jl/stable/>

References: Capstone Allocation

- Magnanti, T.L., Natarajan, K. (2018). Allocating students to multidisciplinary capstone projects using discrete optimization. *Interfaces*, 48(3), 204-216.