### 40.016 The Analytics Edge

Prescriptive analytics with Julia (Part 2)

Stefano Galelli

**SUTD** 

Term 6, 2020

### Outline

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

### **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

### **Challenges** in allocating students to projects:

- **Q** Fairness  $\rightarrow$  Students have equal chances in obtaining their preferred capstone
- Multidisciplinary projects → Every project must be multidisciplinary and involve students from at least two pillars
- ullet Flexibility o 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.

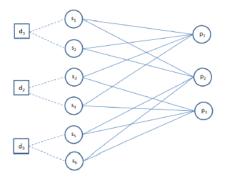


Figure 1: Source: Magnanti and Natarajan (2018).

### Model inputs:

- Network representation
- Lower and upper bounds on the number of students of each discipline (needed for each project)
- Student preferences for projects
- ightarrow 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

**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**

 $\mathsf{UB}(p,d) = \mathsf{Upper}$  bound on the number of students of type d needed for project p

 $\mathsf{LB}(p,d) = \mathsf{Lower}$  bound on the number of students of type d needed for project p

util(s, p) = Utility of project p for student s in the network

#### **Formulation**

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**

 ${\mathcal S}$  : set of student nodes

 $\mathcal{P}$  : set of project nodes

 $\mathcal{D}$  : set of possible types (disciplines) of the students

 $ightarrow {\sf G}({\cal S} \cup {\cal P}, arepsilon)$  : bipartite graph

 $\varepsilon \subseteq \mathcal{S} \times \mathcal{P}$ : set of undirected edges of the graph

#### **Formulation**

$$\begin{aligned} \max \sum_{(s,p) \in \varepsilon} \operatorname{util}(s,p) x_{sp} \\ \text{s.t.} \sum_{p \in \mathcal{P}: (s,p) \in \varepsilon} x_{sp} &= 1 & \forall s \in \mathcal{S} \\ x_{sp} &\leq y_p & \forall (s,p) \in \varepsilon \subseteq \mathcal{S} \times \mathcal{P} \\ \sum_{s \in \mathcal{S}: (s,p) \in \varepsilon, d(s) = d} x_{sp} &\geq \operatorname{LB}(p,d) y_p & \forall d \in \mathcal{D}(p), \forall p \in \mathcal{P} \\ \sum_{s \in \mathcal{S}: (s,p) \in \varepsilon, d(s) = d} x_{sp} &\leq \operatorname{UB}(p,d) y_p & \forall d \in \mathcal{D}(p), \forall p \in \mathcal{P} \\ x_{sp} &\in \{0,1\} & \forall (s,p) \in \varepsilon \subseteq \mathcal{S} \times \mathcal{P} \\ y_p &\in \{0,1\} & \forall p \in \mathcal{P} \end{aligned}$$

#### 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

## 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()

### **JuMP**

### Workflow for creating a model:

- Load JuMP and a solver / optimizer (e.g., GLPK)
- ② Create a model with the Model() function
- 4 Add variables (@variable())
- Add objective (@objective())
- Add constraints, if any (@constrain())

### **JuMP**

### Workflow for solving a problem:

- Use the optimize!() function
- ② Check the status of the solution (termination\_status())

# Implementation in Julia

Please refer to the file capstone.ipynb

### References

### Julia (with links)

- Julia
- IJ
- IJulia
- JuMP

### References

### **Capstone Allocation**

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