# RAISE - Resource Allocation Intelligent Support Engine Model Description

#### Problem Statement

When facing a pandemic disease like currently COVID-19 a major challenge is to fairly allocate scarce resources within regions (e.g., hospitals, cities, provinces or nations).

The objective of the RAISE project is to provide decision support in such cases based on an optimization model.

## Input Data

Data needed for the optimization model are:

- Regions i to be covered in the problem
- Resources r to be allocated
- Demand  $Q_i^r$  of resource r in region i
- Available supply  $A_i^r$  of resource r in region i

Most challenging is the estimation of  $Q_i^r$  which can be based on different sources (e.g., expert knowledge, prediction model, population data). A method to achieve the latter is described for example in (Koyuncu and Erol, 2010). Input data used is:

- Total population  $P_i$  of region i
- $\bullet$  Risk groups g in which the population can be divided
- $\bullet$  Population  $P_{ig}$  in region i of risk group g
- Infection rate  $\alpha_{ig}$  for region i and risk group g
- $\bullet$  Demand  $d_g^r$  of resource r for each patient of risk group g

Then  $Q_i^r$  can be calculated for each i and r as:

$$Q_i^r = \sum_{g} P_{ig} * \alpha_{ig} * d_g^r \tag{1}$$

### Fairness and Preprocessing

Cost in any form (e.g., monetary cost, time, distance) plays a crucial role in allocating goods because of limited available budget. This is covered in the RAISE model by using a cost matrix with entries  $c_{ij}^r$  giving the cost for allocating one unit of resource r from region i to region j. In a our model we focus on time as cost factor.

Nevertheless, another important factor in allocating scarce resources in a pandemic situation is fairness. Different concepts of fair allocation can be used as for example described by (Katoh et al., 2013), who focus on covering fairness completely in the opjective function of the optimization model. In the RAISE project we build our idea of covering fairness on suggestions by (Arora et al., 2010) and use a preprocessing step to fairly distribute the shortage of scarce resources between regions and risk groups.

To achieve this we introduce two auxiliary variables:

- Corrected demand  $\tilde{Q}_i^r$  of resource r in region i
- Corrected supply  $\tilde{A}_i^r$  of resource r in region i (only required in case of surplus of resource r)

The total shortage  $S^r$  can be easily calculated as  $S^r = \sum_i Q_i^r - A_i^r$ . To enable the inclusion of ethical considerations, factors  $\delta^r_{ig}$  for each region r, risk group g and resource r are introduced with  $\sum_i \sum_g \delta^r_{ig} = 1 \ \forall \ r$ . Then the shortage  $S^r_i$  for each resource r and region i is calculated as:

$$S_i^r = \sum_{g} S^r * \delta_{ig}^r \tag{2}$$

with  $\sum_{r} \sum_{i} S_{i}^{r} = S^{r}$  and  $S_{i}^{r} \in \mathbb{Z}$ .

Finally, the corrected demand  $\tilde{Q}_i^r = Q_i^r - S_i^r$  is calculated for each i and r. The corrected supply is simply set to  $\tilde{A}_i^r = A_i^r$  for each i and r. This holds for  $S^r > 0$ .

If  $S^r < 0$  the corrected supply  $\tilde{A}_i^r$  is calculated accordingly and the corrected demand  $\tilde{Q}_i^r$  is set just to the given one.

#### Mathematical Model

Based on the preprocessing step we have to solve a transportation problem with decision variable  $x_{ij}^r$ , which reflects the units of resource r allocated from region i to region j.

The objective function is given by:

$$minimize \sum_{r} \sum_{i} \sum_{j} x_{ij}^{r} * c_{ij}^{r}$$

$$\tag{3}$$

subject to

$$\sum_{j} x_{ij}^{r} = \tilde{A}_{i}^{r} \ \forall \ i, \ r \tag{4}$$

$$\sum_{i} x_{ij}^{r} = \tilde{Q}_{j}^{r} \ \forall \ j, \ r \tag{5}$$

$$x_{ij}^r \ge 0 \tag{6}$$

$$x_{ij}^r \in \mathbb{Z} \tag{7}$$

# Summary

An overview of the steps required to get a fair allocation of scarce resources is displayed in Figure 1.

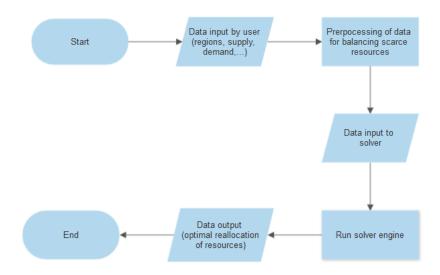


Figure 1: Overview of process steps

## References

- Arora, H., Raghu, T., and Vinze, A. (2010). Resource allocation for demand surge mitigation during disaster response. Decision Support Systems, 50(1):304-315.
- Katoh, N., Shioura, A., and Ibaraki, T. (2013). Resource Allocation Problems, pages 2897–2988. Springer New York, New York, NY.
- Koyuncu, M. and Erol, R. (2010). Optimal resource allocation model to mitigate the impact of pandemic influenza: A case study for turkey. *Journal of medical systems*, 34:61–70.