

Sets and indices

i, j, k, α : indices for nodes, components, subassemblies, final assemblies and plants

n : indices for disruption scenarios, $n = 1, 2, \dots, R$

t^n : duration of disruption in disruption scenarion n , (time to recover – TTR)

$\Phi^{(n)}$: set of nodes (facilities) disrupted in scenarion n

Nodes : set of all nodes in the supply chain

Ω_{jk} : set of all nodes (facilities) that are in the upstream of node j and of part type k

$N^+(i)$: set of subassembly or final assemblies that require component or subassembly i

$N^-(i)$:

set of subassemblies or components that require to produce subassembly or final assembly i

D : all final assembly and subassemblies in BOM (except the last tier)

U : all nodes except the final (assembly) nodes

Y : set of final products /assemblies

A_α : set of all nodes (components, subassemblies and assemblies) produced in plant α

Input parameters

GHG_i : green house gas emission index of node (plant) i for each product produced

GHG_{ij} : green house gas emission of transporting a unit of product from plant i to j .

SI_i : societal impact (employability, following legislations) index of plant i

f_j : profit margin of product j

$c_\alpha t^{(n)}$: capacity of plant α during the all disruption duration $t^{(n)}$ in scenario n

Decision variables

l_j : lost production amount of product j

y_{ij} : amount allocated from upstream node i to downstream node j during disruption duration

u_j : total production quantity of node j during disruption scenario $t^{(n)}$.

Mathematical model:

$$\begin{aligned} \text{Minimize } & \left\{ \sum_{j \in Y} f_j l_j ; \left\{ \sum_{i \in \text{Nodes}} GHG_i u_i + \sum_{i \in \text{Nodes}, j \in N^+(i)} GHG_{ij} y_{ij} \right\} \right\} \\ \text{Maximize } & \sum_{i \in \text{Nodes}} SI_i u_i \end{aligned} \quad (1)$$

Subject to :

$$u_j \leq \sum_{i \in \Omega_{jk}} (y_{ij} / r_{kj}) \quad \forall k \in N^-(j), \forall j \in D \quad (2)$$

$$\sum_{j \in N^+(i)} y_{ij} \leq u_i + s_i \quad \forall i \in U \quad (3)$$

$$u_j = 0 \quad \forall j \in \Phi^{(n)} \quad (4)$$

$$l_j + \sum_{k \in V_j} u_k \geq d_j t^{(n)} \quad \forall j \in Y \quad (5)$$

$$\sum_{k \in A_\alpha} u_k \leq c_\alpha t^{(n)} \quad \forall \alpha \in A \quad (6)$$

$$y_{ij}, u_j, l_j \geq 0 \quad (7)$$

Multi-objective function (1) minimizes the lost sales due to disruption scenario $t^{(n)}$ while trying to best resource allocation plans that minimize GHG index and maximize the societal impact lexicographically.

Constraints (2) are Bill of Material (BOM) inequalities for subassemblies and final products and ensure that for each final product and subassembly j , production is limited with the most scarce parent subassembly/component. See the below figure for representing the relationship.

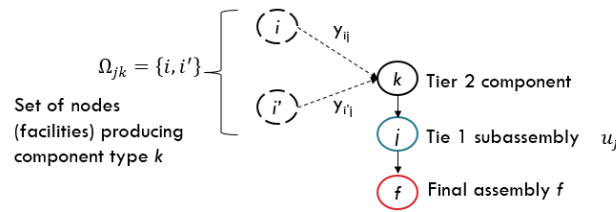


Figure 1: A representative BOM structure

Constraints (3) ensure the BOM inequality for subassemblies and components that total outflow of that subassembly/component to other subassemblies and final products is limited with the production and inventory amount.

Constraints (4) limits the production amount for each facilities in the disruption scenario n .

Constraints (5) guarantee that the lost sales (amount of final products can not be delivered) and the production amount for each final product is at least equal to the demand of that final product during the disruption duration considered.

Constraints (6) ensures that the total production amount in each facility is limited with the capacity of the facility during disruption duration.