VI International Workshop on Locational Analysis and Related Problems

New products supply chains: the effect of short lifecycles on the supply chain network design. GEL ICA

GRUPO ESPAÑOL DE LOCALIZACIÓN



Work in-progress by

Mozart B.C. Menezes (Kedge Business School)

Diego Ruiz-Hernandez (CUNEF) Kai Luo (Kedge Business School) Oihab All-Cherif (Kedge Business School)







CREATED BY BEM & EUROMED MANAGEMENT

The elevator speech

- Node-demand is stochastic
- Facility location
 - Transportation_Cost
- Capacity decisions
 - Newsboy-type problem
 - Cost of Undercapacity: C_u= (Unit_Revenue Unit_Cost) Margin_Outsourcing
 - Cost Overcapacity: C_o=Capacity_Cost
 - Critical Ratio: CR=C_u/(C_u+C_o)
- Difficulty
 - Unit_Cost(Capacity_Cost, Transportation_Cost)



Agenda

- Motivation
- The problem
- The cost of capacity
- The formulation
- The single facility case
- The multi-facility case
- Bounds on the objective function



Overview

- H. Frank (1966): Optimum Location on a Graph with Probabilistic Demands
- Facility Location as an strategic level problem: capacity
- Consider a problem where capacity has to be built up or reserved/bought from third-party
- Planning period is well ahead of product-introduction period
- Demand is unknown and described by scenarios and their probabilities
- Result: Facility Location w/ Capacity Decision (newsvendor type)



The Problem (*a*)

- Consider a vector of demand scenarios: $\partial = (D_1, D_2, ..., D_y)$ and a network G(N,A). |N|=n. D_s is the demand vector for all node-demand in scenario s.
- Each scenario ∂_i , happens with probability $f(\partial_i)$
- Facilities are labeled 1, 2, ..., are located according to a location set S={S₁, S₂,..., S_p}, where p≤n. Vector C is the capacity vector corresponding to each facility.
- For every pair node-facility (*i*, *j*), there is a corresponding unit margin for serving node *i* from facility at *j*.
- *a* denotes an allocation matrix where element a_{ij}^d defines the quantity sent from facility at j to node at i when demand is d.



The Problem (b)

• A feasible allocation *a* is such that $\sum_{i} a_{ij}^{d} \leq C_{j}; \forall j \in \{1, ..., p\}, d \in \partial and$ $\sum_{i} a_{ij}^{d} \leq d_{i}; \forall i \in N, d \in \partial$

- **Cost parameters**
 - $-m_0$ is unit margin when outsourcing, m_{ii} is the margin from serving node *i* from facility at *j*.
- Quantity outsourced $L(C) = \max\{\sum_{i} D_i \sum_{i} C_j, 0\}$
 - Cost of outsourcing is $m_0 L(C)$
- Capacity cost: $\varphi(C)$







The Problem (c)



• The problem

$$\max_{S,C,a} V(S,C,a) = E_d[Z(S,C,a \mid d)]$$



One facility case: sensitivity analysis on capacity







1-Facility Location Solution is Robust

- When independently generating node weights randomly from distributions with same mean value when building scenarios...
 - If the network is large enough then transportation cost dominates and induce solutions among the "usual suspects"; i.e. in the vicinity of each other.





Observations suggest...

- Solutions are robust for changes in critical ratios
- For a fixed set of facility location profit is concave in capacity vector
- Objective is(?) submodular on location set F and C^{*}(F)
- Proposed Greedy Heuristic
 - a) For a fix location set F
 - b) |F|=f, find node $k_f = \arg_k \max Z(C,a | F U k)$
 - c) Stop when profit is not improved; otherwise go to (a) above.



Still hard to compute optimality?

Bounds to the objective function value

- Consider dynamic setting where total capacity is chosen a priori...
- ... but distribution of the capacity is done after scenario is realized
- The bound is easy to find for a fixed set S.
- Recall: Objective function seems to inherity Submodularity property from other location problems (to be proven).





THANKS!

