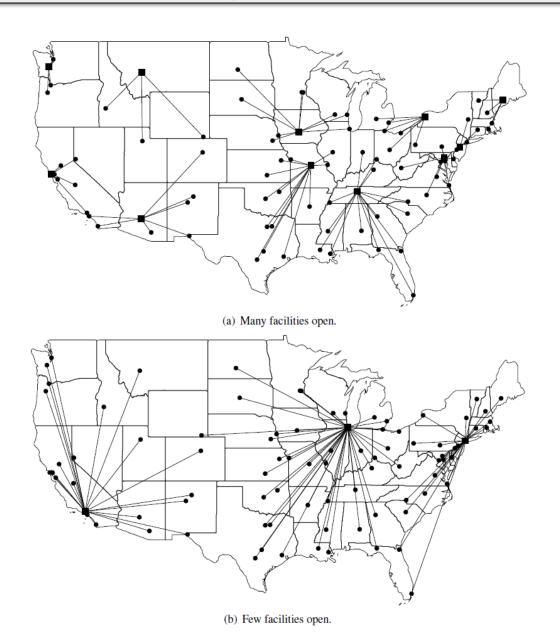
LEC012 Facility Location I

VG441 SS2020

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Configurations



Formulation

Sets

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I = \text{set of customers}

J = \text{set of potential facility locations}
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Parameters

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h_i = \text{annual demand of customer } i \in I

c_{ij} = \text{cost to transport one unit of demand from facility } j \in J \text{ to customer } i \in I

f_j = \text{fixed annual cost to open a facility at site } j \in J
```

Decision Variables

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x_j = 1 if facility j is opened, 0 otherwise y_{ij} = the fraction of customer i 's demand that is served by facility j
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Transportation costs c_{ij}

Euclidean distance

$$\sqrt{(a_1 - a_2)^2 + (b_1 - b_2)^2}$$

Manhattan or rectilinear metric

$$|a_1 - a_2| + |b_1 - b_2|$$

Great circle

$$r \arccos (\sin \alpha_1 \sin \alpha_2 + \cos \alpha_1 \cos \alpha_2 \cos(\Delta \beta))$$

GPS/MAP distance

MILP Formulation

Decision variables:

 $x_j = 1$ if facility j is opened, 0 otherwise $y_{ij} =$ the fraction of customer i 's demand that is served by facility j

$$\begin{array}{ll} \text{minimize} & \sum_{j \in J} f_j x_j + \sum_{i \in I} \sum_{j \in J} h_i c_{ij} y_{ij} \\ \text{subject to} & \sum_{j \in J} y_{ij} = 1, \quad \forall i \in I \quad \longrightarrow \quad \text{Assignment Constraint} \\ & y_{ij} \leq x_j, \quad \forall i \in I, \forall j \in J \quad \longrightarrow \quad \text{Linking Constraint} \\ & x_j \in \{0,1\}, \quad \forall j \in J \\ & y_{ij} \geq 0, \quad \forall i \in I, \forall j \in J \end{array}$$

Python + Gurobi Time!





Lagrangian Relaxation

• Introduce a penalty term: $\sum_{i \in I} \lambda_i \left(1 - \sum_{i \in J} y_{ij} \right)$

$$\begin{aligned} & \text{minimize} & & \sum_{j \in J} f_j x_j + \sum_{i \in I} \sum_{j \in J} h_i c_{ij} y_{ij} + \sum_{i \in I} \lambda_i \left(1 - \sum_{j \in J} y_{ij}\right) \\ & = \sum_{j \in J} f_j x_j + \sum_{i \in I} \sum_{j \in J} \left(h_i c_{ij} - \lambda_i\right) y_{ij} + \sum_{i \in I} \lambda_i \\ & \text{subject to} & & y_{ij} \leq x_j, & \forall i \in I, \forall j \in J \\ & & x_j \in \{0, 1\}, & \forall j \in J \\ & & y_{ij} \geq 0, & \forall i \in I, \forall j \in J \end{aligned}$$

Lagrangian Relaxation

It turns out that this relaxiation is easy to solve:

minimize
$$= \sum_{j \in J} f_j x_j + \sum_{i \in I} \sum_{j \in J} (h_i c_{ij} - \lambda_i) y_{ij} + \sum_{i \in I} \lambda_i$$
 subject to
$$y_{ij} \le x_j, \quad \forall i \in I, \forall j \in J$$

$$x_j \in \{0, 1\}, \quad \forall j \in J$$

$$y_{ij} \ge 0, \quad \forall i \in I, \forall j \in J$$

• Observe that if we open facility j, i.e., set $x_j = 1$

$$\beta_j = \sum_{i \in I} \min \left\{ 0, h_i c_{ij} - \lambda_i \right\} \quad \longrightarrow \quad \text{``benefit''}$$

• So we should open *j* if and only if $\beta_j + f_j < 0$

Lagrangian Relaxation

minimize
$$= \sum_{j \in J} f_j x_j + \sum_{i \in I} \sum_{j \in J} (h_i c_{ij} - \lambda_i) y_{ij} + \sum_{i \in I} \lambda_i$$
subject to
$$y_{ij} \le x_j, \quad \forall i \in I, \forall j \in J$$

$$x_j \in \{0, 1\}, \quad \forall j \in J$$

$$y_{ij} \ge 0, \quad \forall i \in I, \forall j \in J$$

Solution:

$$\bar{x}_{j} = \begin{cases} 1, & \text{if } \beta_{j} + f_{j} < 0 \\ 0, & \text{otherwise} \end{cases}$$

$$\bar{y}_{ij} = \begin{cases} 1, & \text{if } \bar{x}_{j} = 1 \text{ and } h_{i}c_{ij} - \lambda_{i} < 0 \\ 0, & \text{otherwise} \end{cases}$$

Objective value:

$$z_{LR}(\lambda) = \sum_{j \in J} \min \{0, \beta_j + f_j\} + \sum_{i \in I} \lambda_i$$

Now Construct UB

- Use the "opened" facility in the relaxed solution
- Assign each customer to its nearest open facility

Updating Multipliers

Adjust the penalty multipliers:

If $\sum_{j\in J} y_{ij} = 0$, then λ_i is too small; it should be increased. If $\sum_{j\in J} y_{ij} > 1$, then λ_i is too large; it should be decreased. If $\sum_{j\in J} y_{ij} = 1$, then λ_i is just right; it should not be changed.

Then a natural updating rule becomes

$$\lambda_i^{t+1} = \lambda_i^t + \Delta^t \left(1 - \sum_{j \in J} y_{ij} \right)$$
 where the step-size is

$$\Delta^{t} = \frac{(\mathrm{UB} - z_{\mathrm{LR}}(\lambda^{t}))}{\sum_{i \in I} \left(1 - \sum_{j \in J} y_{ij}\right)^{2}}$$