



# Insights into Clients' Choice in Preventive Health Care Facility Location Planning

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# Contents

## 1. Introduction

## 2. Facility Location Model

## 3. Minimum Workload Requirement and Participation

## 4. Computational Studies

## 5. Conclusions

## 6. Appendix 1: Lower Bound



# Problem Setting

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## Clients' Choice Behavior

$N$  set of clients

$M$  choice set: set of alternatives clients choose from

$L$  set of attributes (related to  $m \in M$ ) or characteristics (related to  $n \in N$ )





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Assumptions (Train, 2003)

- ▶  $M$  is exhaustive and alternatives are mutually exclusive
- ▶  $n \in N$  chooses exactly one alternative from choice set  $M$
- ▶  $n \in N$  chooses alternative  $j \in M$  that maximizes utility  $u_{nj}$  of alternative  $j$  for client  $n$



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⇒ **utility maximization choice rule**, i.e.  $n$  chooses  $j$ , iff

$$u_{nj} > u_{nm} \quad \forall m \in M, m \neq j. \quad (1)$$



# Operationalization of Utility

We cannot observe all attributes & characteristics: **stochastic utility**

$$u_{nj} = v_{nj} + \epsilon_{nj} \quad (2)$$

with

$v_{nj}$  deterministic part of utility

$\epsilon_{nj}$  stochastic part of utility



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Assume  $v_{nj}$  to be linear in parameters:

$$v_{nj} = \sum_{l \in L} \beta_{jl} c_{njl} \quad (3)$$

$c_{njl}$  value of attribute/characteristic  $l$  concerning individual  $n$  and alternative  $j$

$\beta_{jl}$  utility contribution per unit of attribute  $l$  related to alternative  $j$   
( $\leftarrow$  MLE)



# Client Choice Model

Stochastic utility:

$$p_{nj} = \text{Prob}(u_{nj} > u_{nm} \quad \forall m \in M, m \neq j) \quad (4)$$

Assume  $\epsilon_{nj}$  iid EV:

## Multinomial Logit Model (MNL)

$$p_{nj} = \frac{e^{u_{nj}}}{\sum_{m \in M} e^{u_{nm}}} \quad (5)$$

denotes probability that client  $n$  chooses alternative  $j$ .



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  - ▶ Assume all clients  $n$  located in demand node (i.e., zone)  $i \in I$  exhibit the same observable characteristics





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$\Rightarrow$  deterministic utility for all clients in  $i$

$$v_{ij} = \sum_{l=1}^L \beta_{jl} c_{ijl} \quad (6)$$



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Choice probability of clients located in  $i \in I$  choosing  $j \in M$  is given by the MNL

$$p_{ij} = \frac{e^{v_{ij}}}{\sum_{m \in M} e^{v_{im}}} \quad (7)$$

Note,  $\sum_{j \in J} p_{ij} < 1$ , while  $\sum_{j \in M} p_{ij} = 1$



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- ▶  $p_{ij}/p_{ik}$  is independent from a third alternative  $m \in M$ , hence if facilities are located at  $j$  and  $k$

$$p_{ij}/p_{ik} = x_{ij}/x_{ik} \quad (9)$$



# Model

$$\text{Maximize } F^A = \sum_{i \in I} g_i \sum_{j \in J} x_{ij} \quad (10)$$

subject to

$$\tilde{x}_i + \sum_{j \in J} x_{ij} \leq 1 \quad \forall i \in I \quad (11)$$

$$x_{ij} \leq \bar{p}_{ij} y_j \quad \forall i \in I, j \in J \quad (12)$$

$$\tilde{p}_i x_{ij} \leq p_{ij} \tilde{x}_i \quad \forall i \in I, j \in J \quad (13)$$

$$\tilde{p}_i x_{ij} \geq p_{ij} \tilde{x}_i + y_j - 1 \quad \forall i \in I, j \in J \quad (14)$$

$$\sum_{i \in I} g_i x_{ij} \geq R_{\min} y_j \quad \forall j \in J \quad (15)$$



## Model continued

$$\sum_{i \in I} g_i x_{ij} \leq \sum_{k=1}^K q_k w_{jk} \quad \forall j \in J \quad (16)$$

$$\sum_{k=1}^K w_{jk} = y_j \quad \forall j \in J \quad (17)$$

$$\sum_{j \in J} \sum_{k=1}^K k \cdot w_{jk} \leq Q_{\max} \quad (18)$$

$$x_{ij} \geq 0 \quad \forall i \in I, j \in J \quad (19)$$

$$\tilde{x}_i \geq 0 \quad \forall i \in I \quad (20)$$

$$y_j \in \{0, 1\} \quad \forall j \in J \quad (21)$$

$$w_{jk} \in \{0, 1\} \quad \forall j \in J, k = 1, \dots, K \quad (22)$$



## Linear Reformulation by Zhang et al (2012)

►  $v_{i,\text{no choice}} = 0$

$M_1$  big number; = 1 for example

$M_2$  big number; = 1 for example

$z_{ijo}$  artificial continuous variable for avoiding non-linearity with  $o \in J$

$$x_{ij} + \sum_{o \in J} e^{v_{io}} z_{ijo} = e^{v_{ij}} y_j \quad i \in I, j \in J \quad (23)$$

$$z_{ijo} \leq x_{ij} \quad i \in I, j, o \in J \quad (24)$$

$$z_{ijo} \leq M_1 y_j \quad i \in I, j, o \in J \quad (25)$$

$$z_{ijo} \geq x_{ij} - M_2(1 - y_j) \quad i \in I, j, o \in J \quad (26)$$

$$z_{ijo} \geq 0 \quad i \in I, j, o \in J \quad (27)$$



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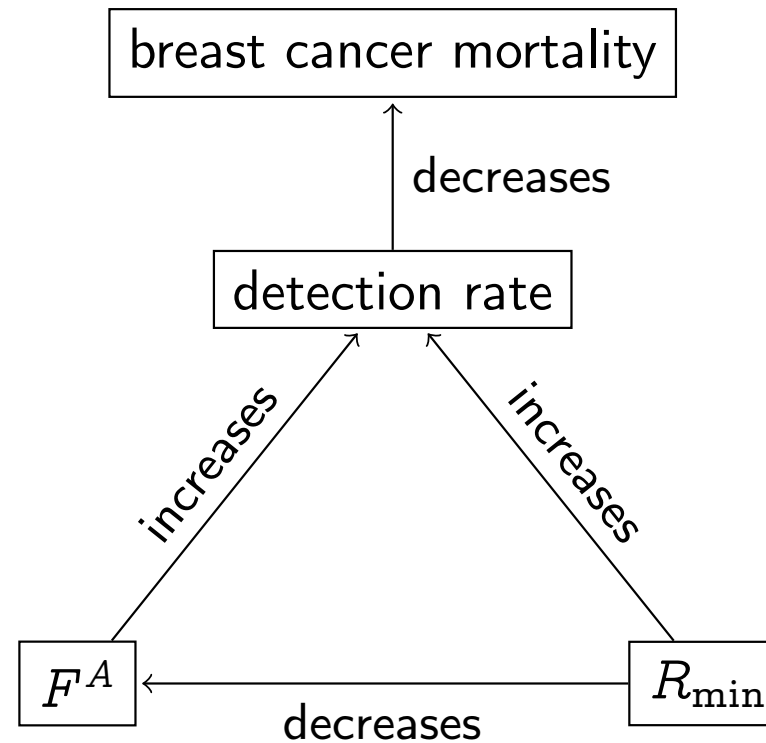


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  - ▶ participation ( $F^A$ )
  - ▶ quality of examination ( $R_{\min}$ )
- ▶ Negative trade-off between  $F^A$  and exogenously given  $R_{\min}$



# Trade Off



- Improvement: make  $R_{\min}$  endogenous to clients' choice behavior



# Numerical Example

- ▶  $j \in \{A, B, C\}$
  - ▶  $Q_{\max} = 2$
  - ▶  $R_{\min}$  be given such that  $x_{1j} > 0.15$  is required
- $\alpha_j$  true-positive-rate

	$v_{1j}$	$\{B, C\}$		$\{A, C\}$	
		$x_{1j}$	$\alpha_j$	$x_{1j}$	$\alpha_j$
facility $j = A$	-1.5	0.000		0.154	0.9
facility $j = B$	-1.0	0.231	0.9	0.000	
facility $j = C$	-1.5	0.140	0.8	0.154	0.9
no-choice $\tilde{x}_1, j = 0$	0	0.629		0.691	
participation rate	-	0.371		0.309	
detection rate	-	0.320		0.276	

poor  $\alpha_C = 0.5$ , solution  $\{B, C\}$  yields a better detection rate than solution  $\{A, C\}$



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# Artificial Data Generation

1. Nodes:
  - ▶ Set  $| I |$  and  $| J |$
2. Population given  $i \in I$ 
  - 2.1 Select randomly  $\phi_i \in [0, 2.4]$
  - 2.2  $g_i := \phi_i \cdot |J| / |I|$
3. Travel times, utility, choice probabilities given  $i \in I$  and  $j \in J$ 
  - 3.1 Generate randomly Cartesian coordinates of nodes in the interval  $[0, 100]$  (uniform distribution)
  - 3.2  $t_{ij} :=$  rectangle distance from node  $i \in I$  to facility  $j \in J$  divided by 60
  - 3.3  $v_{ij} := -\beta \cdot t_{ij}$  with  $\beta = 2$
  - 3.4  $v_{i0} := 0$
  - 3.5  $p_{ij} := \frac{e^{v_{ij}}}{\sum_m e^{v_{im}}}$
  - 3.6  $\tilde{p}_i := \frac{1}{1 + \sum_j e^{v_{ij}}}$
  - 3.7  $\bar{p}_{ij} := \min\{0.5, e^{v_{ij}} / (1 + e^{v_{ij}})\}$



## Artificial Data Generation (continued)

Other parameters according to Zhang et al. (2012)

- ▶  $K := 4$
- ▶  $\nabla \lambda_k := 1.05^k$
- ▶  $R_{\min} := 1.2$
- ▶  $q_k := B \cdot \nabla \lambda_k$  with  $B = 1.5 \cdot R_{\min}$
- ▶  $Q_{\max} := \lceil (|J| / 2) \rceil$
- ▶  $M_1 = M_2 := 1$



# Results

J	I	Model Z		Model A		Model A with $\underline{F}^A = LB^*$		
		"GAP"	CPU	"GAP"	CPU	"GAP"	CPU ( $LB^*$ )	$100 \frac{F^{A*} - LB^*}{F^{A*}}$
10	100	0	35	0	4	0	9 (2)	1.55
	200	0	196	0	15	0	23 (2)	4.18
	400	0	831	0	60	0	51 (7)	2.97
20	100	406.44 <sup>b</sup>	3600	0	619	0	137 (1)	2.48
	200	-	3600	1.66	2858	0	604 (2)	1.66
	400	185.34 <sup>d</sup>	3600	7.35	3600	0.22	1993 (2)	0.93 <sup>a</sup>
40	100	89.99 <sup>c</sup>	3600	8.58	3600	3.34	3600 (8)	-
	200	96.44 <sup>c</sup>	3600	27.88	3600	8.21	3600 (25)	-
	400	92.69 <sup>d</sup>	3600	55.67	3600	12.10	3600 (75)	-

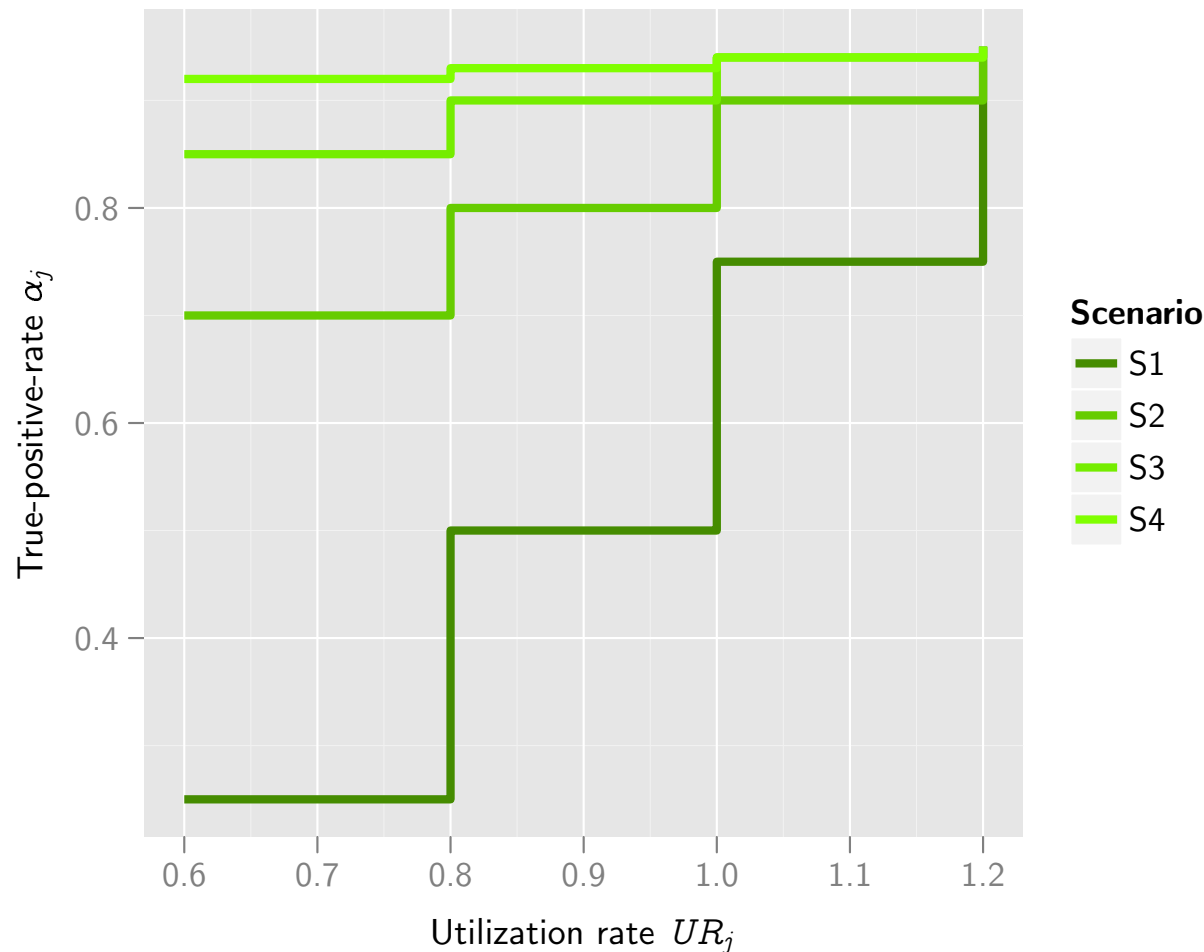
Number of instances where an optimal solution is obtained:  $a: 9, b: 5, c: 3, d: 1$

- ▶ Averages over ten random instances
- ▶ CPU: CPLEX time in seconds
- ▶ (): time to compute LB
- ▶ GAP provided by CPLEX



# Numerical Investigation of Trade-Off between $F^A$ and $R_{\min}$

$$\text{Utilization rate } UR_j = \sum_{i \in I} g_i x_{ij} \quad \forall j \in J,$$







# Data

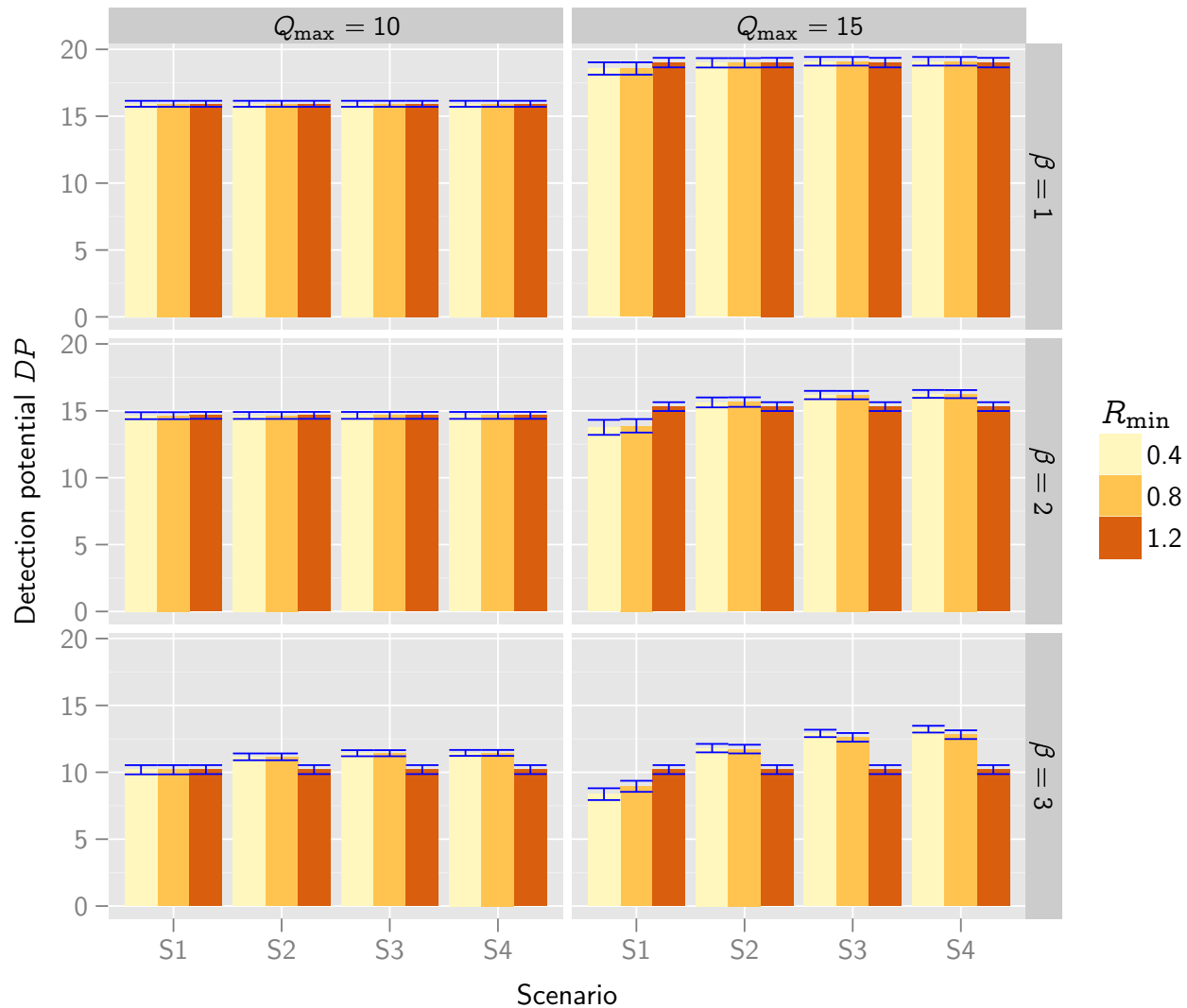
Total detection potential

$$DP = \sum_{i \in I} g_i \sum_{j \in J} \alpha_j x_{ij} \quad (28)$$

- ▶  $|J| = 20, |I| = 100$
- ▶  $B = 1.2 \cdot 1.5 = 1.8$
- ▶ 10 random instances for  $R_{\min} = \{1.2, 0.8, 0.4\}$

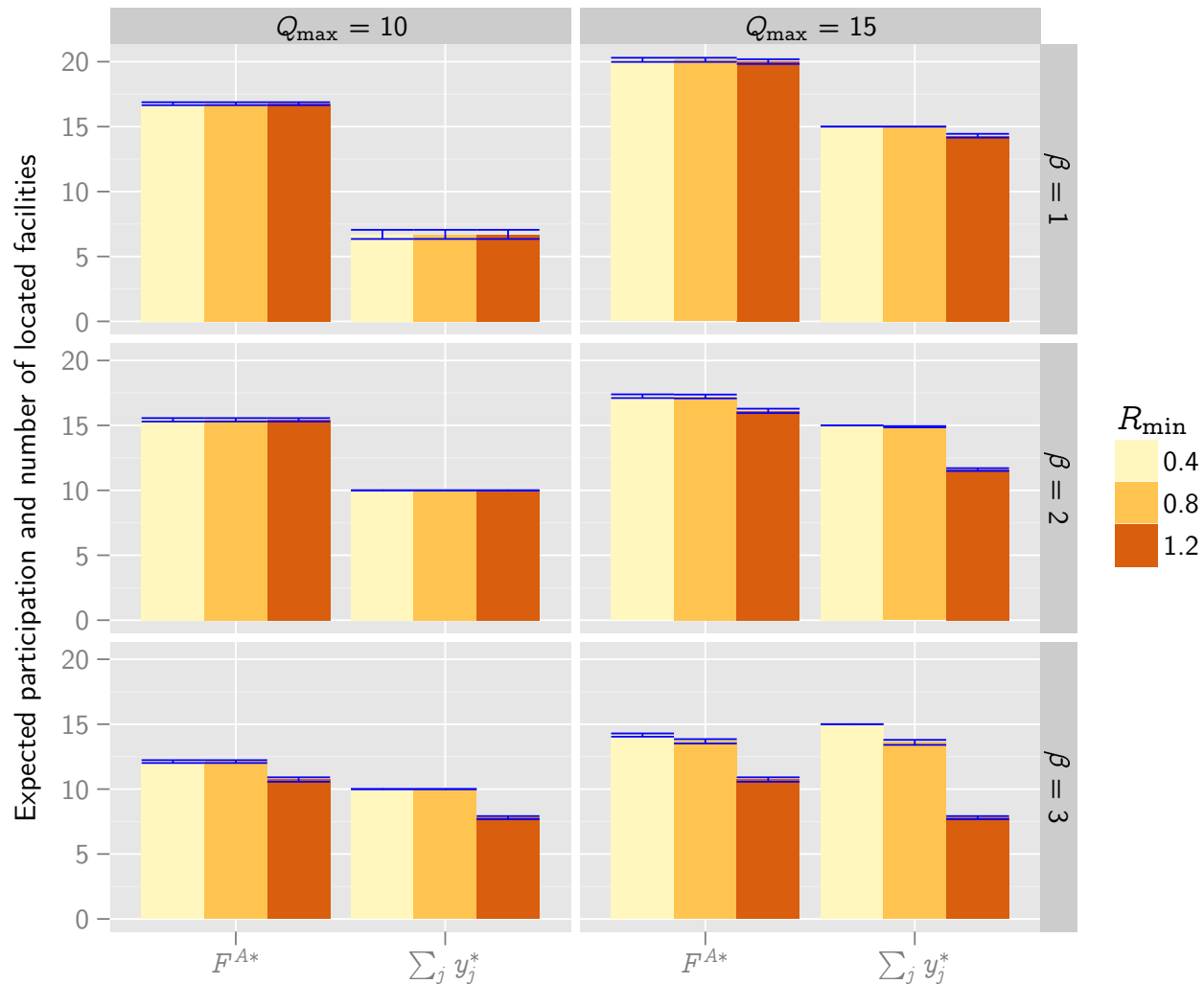


# Resulting Detection Potential





# Expected Participation





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- ⇒ incorporation of expected waiting times and quality of care (true-positive rate, for example) in clients' utility function



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# Contents

1. Introduction
2. Facility Location Model
3. Minimum Workload Requirement and Participation
4. Computational Studies
5. Conclusions
6. Appendix 1: Lower Bound



## Step 1

$\Omega_{ij} = 1$ , if node  $i \in I$  is assigned to facility location  $j \in J$  (0, otherwise),

$\Psi_j = 1$ , if location  $j \in J$  provides a health care facility (0, otherwise),

$F^B$  ofv indicating the cumulative choice probabilities,

$F^C$  ofv indicating the attractiveness of the located facilities, and

$LB$  lower bound to Model A.

Model B

$$\text{Maximize } F^B = \sum_{i \in I} \sum_{j \in J} p_{ij} \Omega_{ij} \quad (29)$$

subject to

$$\sum_{j \in J} \Omega_{ij} = Q_{\max} \quad \forall i \in I \quad (30)$$

$$\Omega_{ij} \in \{0, 1\} \quad \forall i \in I, j \in J \quad (31)$$

determines for each demand node  $i$  the  $Q_{\max}$  most attractive locations.



## Step 2

Attractiveness

$$b_j = \sum_{i \in I} g_i \Omega_{ij}^* \quad \forall j \in J \quad (32)$$

Model C

$$\text{Maximize } F^C = \sum_{j \in J} b_j \Psi_j \quad (33)$$

subject to

$$\sum_{j \in J} \Psi_j = Q_{\max} \quad (34)$$

$$\Psi_j \in \{0, 1\} \quad \forall j \in J \quad (35)$$

determines the  $Q_{\max}$  most attractive facility locations.



## Step 3

### Model D

$$\text{Maximize } LB = \sum_{i \in I} g_i \sum_{j \in J} x_{ij} \quad (36)$$

subject to (11) - (22) and

$$y_j \leq \Psi_j^* \quad \forall j \in J \quad (37)$$

determines a lower bound to Model A