

1 The Model

1.1 Population and sampling parameters

- Sets of domains and strata indices in domains:

$$\mathcal{D} := \{1, \dots, D\} \quad \text{domain indices; } D \text{ is the total number of domains,}$$

$$\mathcal{H}_d := \{1, \dots, H_d\}, d \in \mathcal{D} \quad \text{strata indices in domain } d \text{ where } H_d \text{ is the total number of strata in domain } d,$$

$$\mathcal{H} := \{(d, h) : d \in \mathcal{D}, h \in \mathcal{H}_d\} \quad \text{domain-stratum indices.}$$

- Population quantities:

$$N_{d,h}, (d, h) \in \mathcal{H} \quad \text{size of stratum } h \text{ in domain } d,$$

$$S_{d,h}, (d, h) \in \mathcal{H} \quad \text{standard deviation of a given study variable in stratum } h \text{ in domain } d,$$

$$\tau_d, d \in \widetilde{\delta}(\mathcal{H}) \quad \text{total in domain } d, \text{ i.e. a sum of values of a given study variable for population elements in domain } d,$$

$$\kappa_d, d \in \widetilde{\delta}(\mathcal{H}) \quad \text{priority weight for domain } d,$$

$$\rho_d := \tau_d \sqrt{\kappa_d}, d \in \widetilde{\delta}(\mathcal{H}),$$

where function $\widetilde{\delta}$ is as defined in (1.1a).

- Sampling parameters:

$$n \quad \text{total sample size.}$$

1.2 Model definition and properties

The model serves as an abstract representation of all the population and sampling quantities defined in Section 1.1.

Definition 1.1. Let $I \subset \mathbb{N}_+^2$ and $J \subseteq I$. Then,

$$\widetilde{\delta}(I) := \{d : \pi_1(i) = d, i \in I\}, \quad (1.1a)$$

$$\widetilde{\delta}_d(J) := \{h : \pi_1(i) = d, \pi_2(i) = h, i \in I\}, \quad d \in \widetilde{\delta}(I), \quad (1.1b)$$

where

$$\pi_i(\mathbf{a}) := a_i, \quad \mathbf{a} = (a_i, i \in I) \in \mathbb{R}^{|I|}, i \in I. \quad (1.2)$$

Definition 1.2. By a model, we understand a quintuple $(\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n)$, such that

$$\mathcal{H} \subset \mathbb{N}_+^2, \quad 0 < |\mathcal{H}| < \infty, \quad (1.3a)$$

$$\mathbf{N} = (N_{d,h}, (d,h) \in \mathcal{H}) \in \mathbb{N}_+^{|\mathcal{H}|}, \quad (1.3b)$$

$$\mathbf{S} = (S_{d,h}, (d,h) \in \mathcal{H}) \in \mathbb{R}_+^{|\mathcal{H}|}, \quad (1.3c)$$

$$\boldsymbol{\rho} = (\rho_d, d \in \tilde{\delta}(\mathcal{H})) \in \mathbb{R}_+^{|\tilde{\delta}(\mathcal{H})|}, \quad (1.3d)$$

$$n \in \left(0, \sum_{(d,h) \in \mathcal{H}} N_{d,h}\right], \quad (1.3e)$$

where $\tilde{\delta}$ is defined in (1.1a).

Definition 1.3. We define the following classes of models:

$$\mathcal{P} := \{(\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n) : (1.3) \text{ hold}\}. \quad (1.4a)$$

Example of the model. Let

$$\begin{aligned} \mathcal{H} &= \{(1, 1), (1, 2), (2, 1), (2, 2), (2, 3)\}, \\ \mathbf{N} &= (N_{1,1}, N_{1,2}, N_{2,1}, N_{2,2}, N_{2,3}) = (100, 200, 150, 40, 50), \\ \mathbf{S} &= (S_{1,1}, S_{1,2}, S_{2,1}, S_{2,2}, S_{2,3}) = (10, 2, 50, 30, 20), \\ \boldsymbol{\rho} &= (\rho_1, \rho_2) = (1.26, 2.32). \\ n &= 350 \end{aligned} \quad (1.5)$$

Then $(\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n) \in \mathcal{P}$.

Definition 1.4. Let functions $\tilde{\delta}$ and $\tilde{\delta}_d$ be as defined in (1.1). We define the following model functions:

$$\delta(p) := \tilde{\delta}(\mathcal{H}), \quad p \in \mathcal{P}, \quad (1.6a)$$

$$\delta_d(p) := \tilde{\delta}_d(\mathcal{H}), \quad d \in \delta(p), \quad p \in \mathcal{P}, \quad (1.6b)$$

$$A_{d,h}(p) := \frac{N_{d,h} S_{d,h}}{\rho_d}, \quad (d, h) \in \mathcal{H}, \quad p \in \mathcal{P}, \quad (1.6c)$$

$$c_d(p) := \sum_{h \in \delta_d(p)} \frac{[A_{d,h}(p)]^2}{N_{d,h}}, \quad d \in \delta(p), \quad p \in \mathcal{P}, \quad (1.6d)$$

where $p = (\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n)$.

2 Allocation Problem

Problem 2.1 (FPDA). *For a given $p = (\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n) \in \mathcal{P}$, the FPDA(p) problem is defined as follows:*

$$\underset{(T, \mathbf{x})}{\text{minimize}} \quad T \tag{2.1}$$

$$\text{subject to} \quad \sum_{(d,h) \in \mathcal{H}} x_{d,h} - n = 0 \tag{2.2a}$$

$$\sum_{h \in \delta_d(p)} \frac{[A_{d,h}(p)]^2}{x_{d,h}} - c_d(p) - T = 0, \quad d \in \delta(p) \tag{2.2b}$$

$$x_{d,h} - N_{d,h} \leq 0, \quad (d, h) \in \mathcal{H}, \tag{2.2c}$$

where $(T, \mathbf{x}) = (T, (x_{d,h}, (d, h) \in \mathcal{H})) \in \mathbb{R} \times \mathbb{R}_+^{|\mathcal{H}|}$ is an optimization variable, and $\delta, \delta_d, A_{d,h}, c_d$ are the functions as defined in (1.6).

3 Set functions and some population matrix

Definition 3.1. Let $p = (\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n) \in \mathcal{P}$. Suppose that $\mathcal{A} \subsetneq \mathcal{H}$ and $\mathbf{v} = (v_d, d \in \tilde{\delta}(\mathcal{A}^c)) \in \mathbb{R}_+^{|\tilde{\delta}(\mathcal{A}^c)|}$. Then set functions s_d , for $d \in \tilde{\delta}(\mathcal{A}^c)$, are defined as follows:

$$s_d(\mathcal{A}, \mathbf{v} | p) := \frac{n - \sum_{(i,h) \in \mathcal{A}} N_{i,h}}{\sum_{(i,h) \in \mathcal{A}^c} v_i A_{i,h}(p)} v_d, \quad d \in \tilde{\delta}(\mathcal{A}^c), \quad (3.1)$$

where $\mathcal{A}^c := \mathcal{H} \setminus \mathcal{A}$.

Definition 3.2. Let $p = (\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n) \in \mathcal{P}$. Suppose that $\mathcal{B} \subseteq \mathcal{H}$ is such that $\sum_{(d,h) \in \mathcal{B}^c} N_{d,h} < n$. Then matrix-valued function \mathbf{D} is defined by

$$\mathbf{D}(\mathcal{B} | p) := \frac{1}{n - \sum_{(d,h) \in \mathcal{B}^c} N_{d,h}} \mathbf{a} \mathbf{a}^\top - \text{diag}(\mathbf{c}), \quad (3.2)$$

where $\mathcal{B}^c := \mathcal{H} \setminus \mathcal{B}$, and

$$\mathbf{a} = (a_d, d \in \tilde{\delta}(\mathcal{B}))^\top := \left(\sum_{h \in \tilde{\delta}_d(\mathcal{B})} A_{d,h}(p), d \in \tilde{\delta}(\mathcal{B}) \right)^\top, \quad (3.3a)$$

$$\mathbf{c} = (c_d, d \in \tilde{\delta}(\mathcal{B}))^\top := \left(\sum_{h \in \tilde{\delta}_d(\mathcal{B})} \frac{[A_{d,h}(p)]^2}{N_{d,h}}, d \in \tilde{\delta}(\mathcal{B}) \right)^\top, \quad (3.3b)$$

Symbol \top denotes matrix transpose (i.e. \mathbf{a}, \mathbf{c} are column vectors) and $\text{diag}(\mathbf{c})$ is a diagonal matrix with vector \mathbf{c} on the diagonal.

Definition 3.3. Let $p = (\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n) \in \mathcal{P}$. Suppose that $\mathcal{B} \subseteq \mathcal{H}$ is such that $\sum_{(d,h) \in \mathcal{B}^c} N_{d,h} < n$, and $\tilde{\delta}(\mathcal{B}) = \{d_1, \dots, d_D\}$, where $D := |\tilde{\delta}(\mathcal{B})|$. Then the *Eigen operator* is defined as follows:

$$\text{Eigen}(\mathcal{B} | p) := (\lambda, \mathbf{v}), \quad (3.4)$$

where (λ, \mathbf{v}) is the unique eigenpair for matrix $\mathbf{D}(\mathcal{B} | p)$ such that $\mathbf{v} = (v_{d_1}, \dots, v_{d_D}) \in \mathbb{R}_+^D$ is a unit vector.

To illustrate the functioning of the *Eigen* operator, consider $p \in \mathcal{P}$ with $\mathcal{B} \subseteq \mathcal{H}$, such that $\delta(p) = \{1, 2, 3\}$ and $\tilde{\delta}(\mathcal{B}) = \{2, 3\}$. Suppose that $(\lambda, (w_1, w_2))$ is an eigenpair for the matrix $\mathbf{D}(\mathcal{B} | p)$. Then, $\text{Eigen}(\mathcal{B} | p) = (\lambda, \mathbf{v})$, where $\mathbf{v} = (v_2, v_3) = (w_1, w_2)$. This example demonstrates that $\tilde{\delta}(\mathcal{B}) \subseteq \delta(p)$, and hence, the dimension of the vector \mathbf{v} may be less than or equal to $|\delta(p)|$.

4 RFIXPREC Algorithm

Algorithm 1 *RFIXPREC*

Input: $p = (\mathcal{H}, \mathbf{N}, \mathbf{S}, \boldsymbol{\rho}, n) \in \mathcal{P}$, $\mathcal{U} \subseteq \mathcal{H}$, $\mathcal{J} \subseteq \delta(p)$. ▷ see (1.4a) and (1.6a)

Require: $(\sum_{(d,h) \in \mathcal{U}} N_{d,h} < n \text{ or } \mathcal{U} = \mathcal{H})$ and $(\tilde{\delta}(\mathcal{U}) \not\subseteq \mathcal{J} \text{ if } \mathcal{U} \neq \emptyset)$ and $\mathcal{J} \neq \emptyset$. ▷ see (1.1)

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1: function RFIXPREC( $p, \mathcal{U}, \mathcal{J}$ )
2:    $j \in \mathcal{J}$  ▷ chosen arbitrarily
3:    $\mathcal{H}_j \leftarrow \{(d, h) \in \mathcal{H} : d = j\}$ 
4:   do
5:     if  $|\mathcal{J}| = 1$  then
6:        $(T, \mathbf{x}) \leftarrow \text{FIXPREC}(\mathcal{U})$  ▷ see the next page
7:     else
8:        $(T, \mathbf{x}) \leftarrow \text{RFIXPREC}(p, \mathcal{U}, \mathcal{J} \setminus \{j\})$ 
9:     end if
10:     $\mathcal{U}_j \leftarrow \{(d, h) \in \mathcal{H}_j : x_{d,h} \geq N_{d,h}\}$  ▷  $\mathbf{x} = (x_{d,h}, (d, h) \in \mathcal{H})$ 
11:    if  $\mathcal{U}_j \neq \emptyset$  then
12:       $\mathcal{U} \leftarrow \mathcal{U} \cup \mathcal{U}_j$ 
13:       $\mathcal{H}_j \leftarrow \mathcal{H}_j \setminus \mathcal{U}_j$ 
14:    end if
15:    while  $\mathcal{U}_j \neq \emptyset$ 
16:    return  $(T, \mathbf{x})$ 
17: end function

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Algorithm 2 *FIXPRECACT*

Input: $p = (\mathcal{H}, N, S, \rho, n) \in \mathcal{P}$, $\mathcal{U} \subseteq \mathcal{H}$.

▷ see (1.4a)

Require: $\sum_{(d,h) \in \mathcal{U}} N_{d,h} < n$ or $\mathcal{U} = \mathcal{H}$.

1: **function** FIXPRECACT(p, \mathcal{U})

2: **if** $\mathcal{U} = \mathcal{H}$ **then**

3: $T \leftarrow 0$

4: **else**

5: $(\lambda, \mathbf{v}) \leftarrow \text{Eigen}(\mathcal{H} \setminus \mathcal{U} \mid p)$

▷ see Definition 3.3 of *Eigen*

6: $T \leftarrow \lambda$

7: **end if**

8: **return** (T, \mathbf{x}) where $x_{d,h} = \begin{cases} N_{d,h}, & (d, h) \in \mathcal{U} \\ s_d(\mathcal{U}, \mathbf{v} \mid p) A_{d,h}(p), & (d, h) \in \mathcal{H} \setminus \mathcal{U} \end{cases}$ ▷ see Definition 3.1 of s_d

9: **end function**
