

Comparison of Bayesian and non-Bayesian approaches to correct for measurement error in migration flow data for South America

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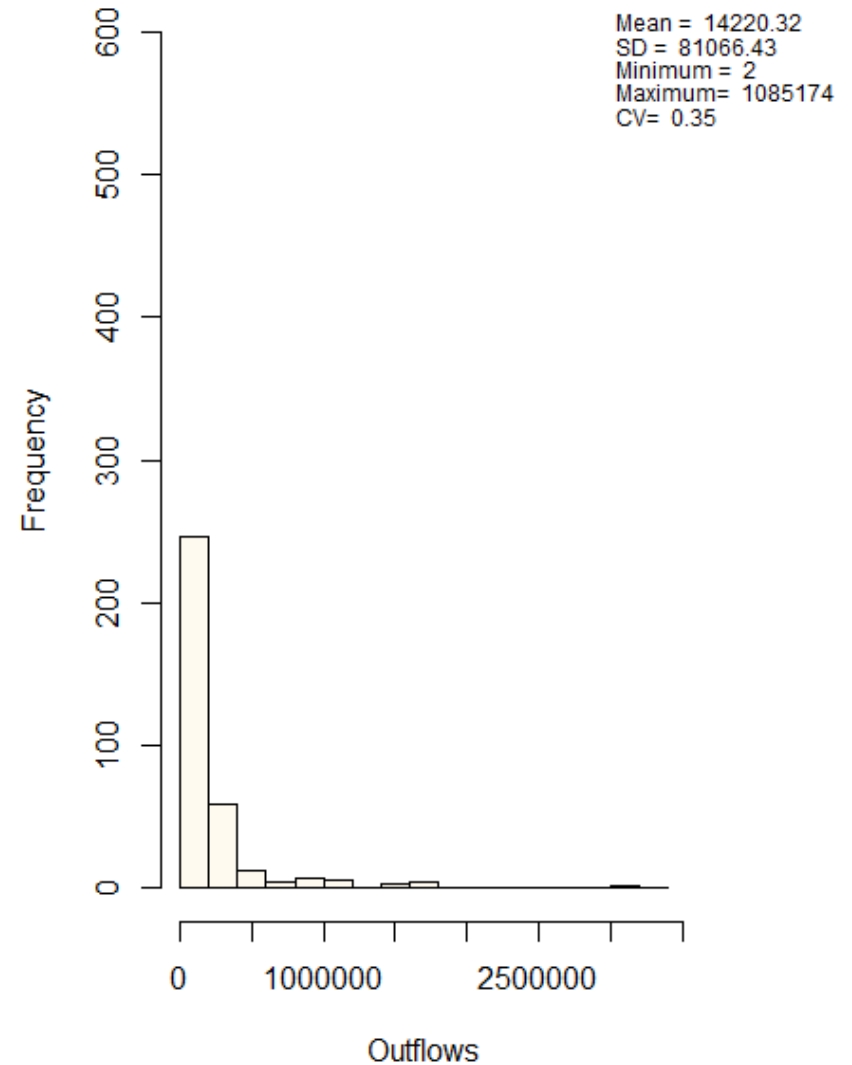
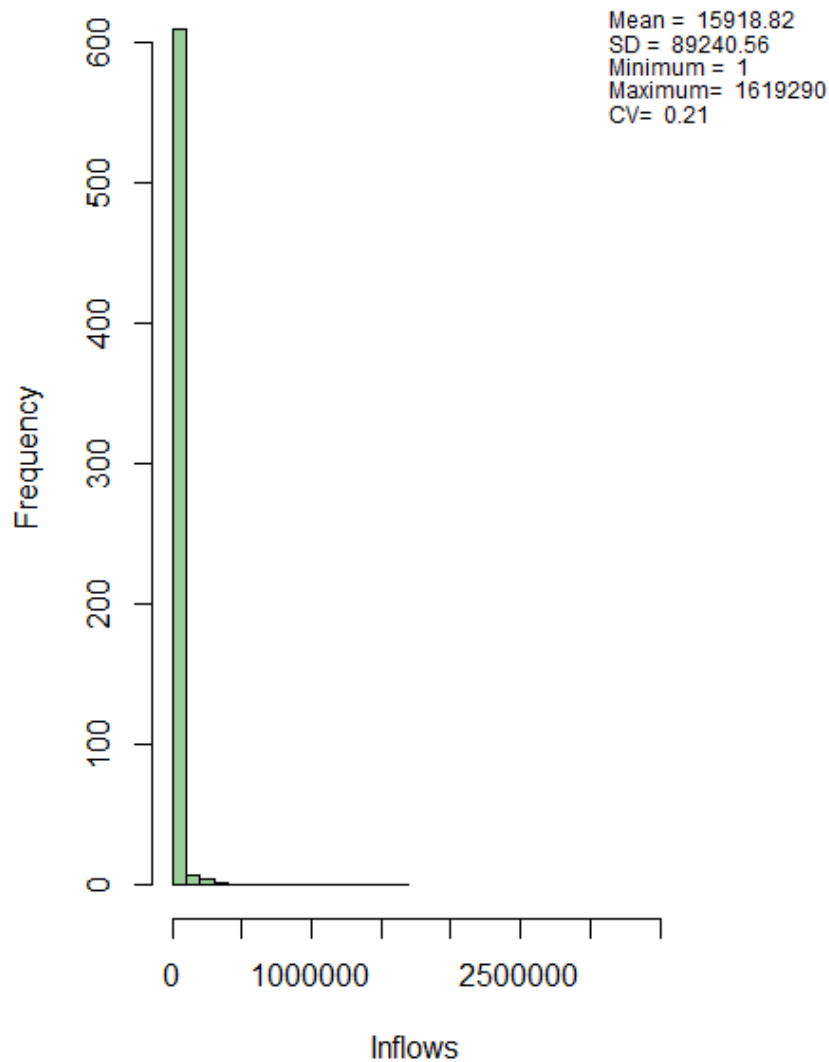
Observed South American countries (in blue)



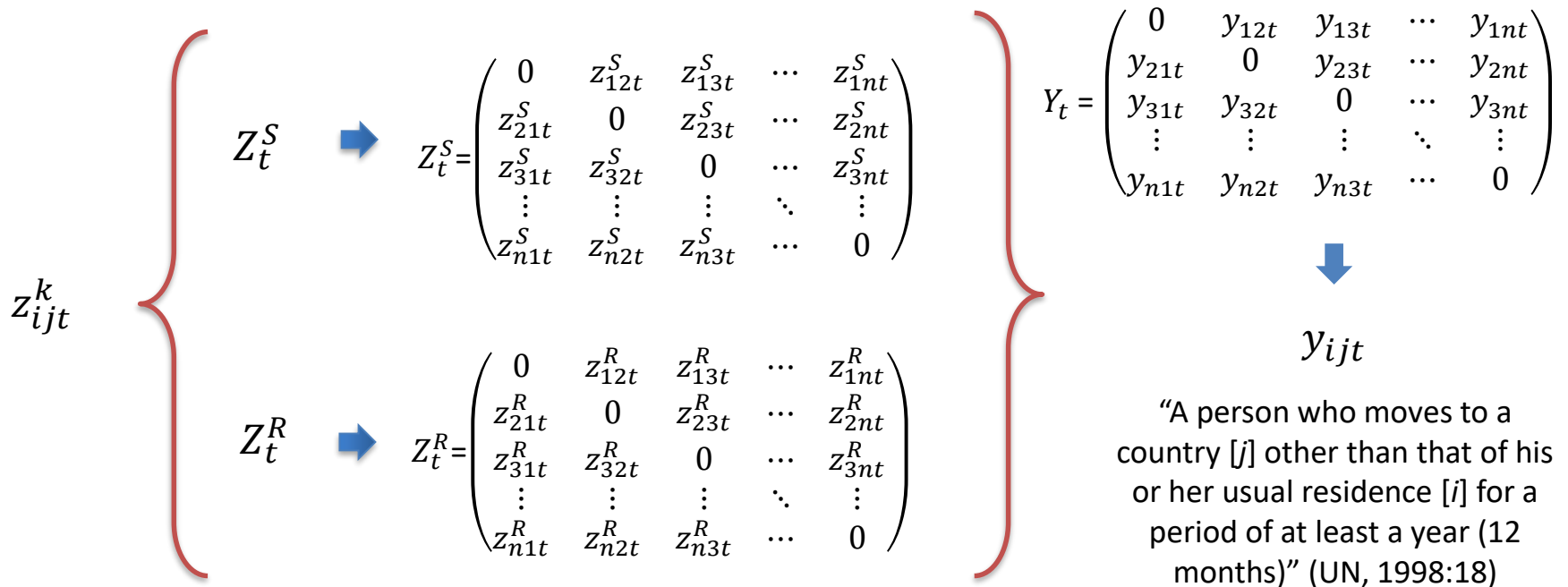
Data

$$\left. \begin{array}{l} z_{ijt}^k \end{array} \right\} \begin{array}{l} Z_t^S \rightarrow Z_t^S = \begin{pmatrix} 0 & z_{12t}^S & z_{13t}^S & \cdots & z_{1nt}^S \\ z_{21t}^S & 0 & z_{23t}^S & \cdots & z_{2nt}^S \\ z_{31t}^S & z_{32t}^S & 0 & \cdots & z_{3nt}^S \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ z_{n1t}^S & z_{n2t}^S & z_{n3t}^S & \cdots & 0 \end{pmatrix} \\ \\ Z_t^R \rightarrow Z_t^R = \begin{pmatrix} 0 & z_{12t}^R & z_{13t}^R & \cdots & z_{1nt}^R \\ z_{21t}^R & 0 & z_{23t}^R & \cdots & z_{2nt}^R \\ z_{31t}^R & z_{32t}^R & 0 & \cdots & z_{3nt}^R \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ z_{n1t}^R & z_{n2t}^R & z_{n3t}^R & \cdots & 0 \end{pmatrix} \end{array}$$

Data: comparison between inflows and outflows in South America



Data



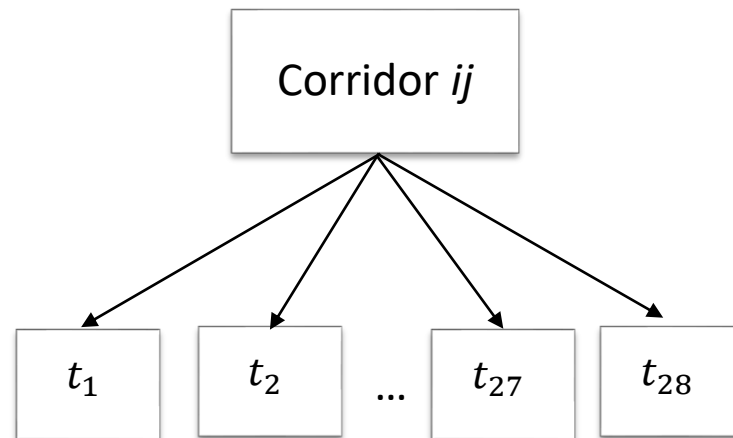
Data can be seen as ...

| Origin i | Destination j | Corridor ij | Year t | Counts z_{ijt}^k |
|------------|-----------------|---------------|----------|--------------------|
| ARG | BOL | ARG -> BOL | 1991 | ... |
| ARG | BRA | ARG -> BRA | 1991 | ... |
| ARG | CHL | ARG -> CHL | 1991 | ... |
| ⋮ | ⋮ | ⋮ | ⋮ | ⋮ |

Level 2 (90 corridors)

Corridor ij

Level 1 (2520 observations)



Adapting data to the UN definition

y_{ijt}



$$z_{ijt}^S \sim \text{Poisson}(\mu_{ijt}^S)$$

$$z_{ijt}^R \sim \text{Poisson}(\mu_{ijt}^R)$$



$$\log(\mu_{ijt}^k) = \log(y_{ijt}) + q_{ijt}^k$$



$$\log(\mu_{ijt}^k) = \log(y_{ijt}) + \delta_{m(k)} + \lambda_{n(k)} + \beta_{q(i)} + \beta_{q(j)} + q_{ijt}^k$$



$$\log(\mu_{ijt}^S) = \log(y_{ijt}) + \delta_{m(i)} + \lambda_{n(k)} + \beta_{q(i)} + q_{ijt}^S$$

$$\log(\mu_{ijt}^R) = \log(y_{ijt}) + \delta_{m(j)} + \lambda_{n(k)} + \beta_{q(j)} + q_{ijt}^R$$

Adapting data to the UN definition

$$\log(\mu_{ijt}^S) = \log(y_{ijt}) + \delta_{m(i)} + \lambda_{n(k)} + \beta_{q(i)} + q_{ijt}^S ;$$

$$q_{ijt}^S \sim N(0, \sigma_{q(i)}^2)$$

$$\log(\mu_{ijt}^R) = \log(y_{ijt}) + \delta_{m(j)} + \lambda_{n(k)} + \beta_{q(j)} + q_{ijt}^R ;$$

$$q_{ijt}^R \sim N(0, \sigma_{q(j)}^2)$$

$$\delta_{m(k)} = \begin{cases} \delta_1 & \text{if criterion is "0 months"} \\ \delta_2 & \text{if criterion is "3 months"} \\ \delta_3 & \text{if criterion is "6 months"} \\ 0 & \text{if criterion is "12 months"} \end{cases}$$

$$\lambda_{n(k)} = \begin{cases} \lambda_1 & \text{if k source undercounts immigrants} \\ \lambda_2 & \text{if k source undercounts emigrants} \end{cases}$$

$$\delta_1 = \delta_1 + \delta_2 + \delta_3$$

$$\delta_2 = \delta_2 + \delta_3$$

$$\delta_3 = \delta_3$$

$$\delta_4 = -\delta_4$$

$$\beta_{q(k)} = \begin{cases} \beta_1 & \text{Good registers} \\ \beta_2 & \text{Less reliable registers} \\ 0 & \text{Excellent register} \end{cases}$$

Non-Bayesian and Bayesian approaches

Non-Bayesian approach => Maximum Likelihood

vs.

Bayesian approach => MCMC

Prior of $\log(y_{ijt})$

$$\log(y_{ijt}) = \alpha_0 + \alpha_1 t + \alpha_2 \log(P_{it}) + \alpha_3 \log(P_{jt}) + \alpha_4 B_{ijt} + u_{0ij} + u_{1ij}t + e_{ijt}$$

$$\begin{pmatrix} u_{0ij} \\ u_{1ij} \end{pmatrix} \sim N \left[\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_{u0}^2 & \\ 0 & \sigma_{u1}^2 \end{pmatrix} \right]$$

$$e_{ijt} \sim N(0, \sigma_e^2)$$

Variances of non-adapted and adapted estimates

| Parameters | | Maximum likelihood | | MCMC | |
|-------------------------|-----------------|--------------------|-----------------|--------|-----------------|
| | | Estimate | 95% CI | Mean | 95% HDI region |
| Non-adapted data | σ_{qS}^2 | 0.665 | (0.161, 0.846) | 0.423 | (0.002, 0.664) |
| | σ_{qR}^2 | 49.741 | (2.419, 98.098) | 50.109 | (2.797, 97.073) |
| Adapted data | σ_{qS}^2 | 0.602 | (0.046, 0.836) | 0.4770 | (0.074, 0.676) |
| | σ_{qR}^2 | 50.099 | (2.717, 97.494) | 49.837 | (3.335, 97.693) |

Measurement model parameters of adapted data

| Parameters | Maximum likelihood | | | MCMC | | |
|-------------|--------------------|--------|-------------------|--------|--------|-------------------|
| | Estimate | SE | 95% CI | Mean | SD | 95% HDI region |
| δ_1 | 0.196 | 0.086 | (0.056, 0.395) | 0.143 | 0.067 | (0.035, 0.291) |
| δ_2 | 0.130 | 0.076 | (0.019, 0.302) | 0.090 | 0.053 | (0.0133, 0.215) |
| δ_3 | 0.065 | 0.058 | (0.002, 0.221) | 0.044 | 0.039 | (0.001, 0.148) |
| λ_1 | 4.925 | 1.180 | (2.944, 6.990) | 11.112 | 2.178 | (7.271, 13.354) |
| λ_2 | -0.535 | 30.931 | (-62.519, 59.292) | 0.263 | 32.588 | (-63.626, 63.548) |
| β_1 | -0.098 | 0.077 | (-0.244, 0.052) | -0.100 | 0.057 | (-0.211, 0.013) |
| β_2 | -0.053 | 0.046 | (-0.142, 0.037) | -0.042 | 0.039 | (-0.118, 0.032) |

$$\delta_3 = 0.044 \Rightarrow \exp(-\delta_3) = 0.9569$$

For countries with a minimum duration of stay of 6 months, the true flows represent 95.69% of the observed/reported flows.

***Sensitivity of variances to the partial removal of the data
(MCMC estimates)***

| Parameters | | Complete data | | Partial removal of the data | |
|-------------------------|-----------------|---------------|-----------------|-----------------------------|-----------------|
| | | Mean | 95% HDI region | Mean | 95% HDI region |
| Non-adapted data | σ_{qS}^2 | 0.423 | (0.002, 0.664) | 0.535 | (0.115, 0.687) |
| | σ_{qR}^2 | 50.109 | (2.797, 97.073) | 50.115 | (2.895, 97.339) |
| Adapted data | σ_{qS}^2 | 0.477 | (0.074, 0.676) | 0.524 | (0.081, 0.709) |
| | σ_{qR}^2 | 49.837 | (3.335, 97.693) | 50.325 | (2.799, 97.606) |

Conclusions

- Despite the differences between MLE and MCMC estimates, the HDIs contain the values obtained in the foremost.

Further work

- Improving the criteria to consider the accuracy of the data sources.
- Including additional measurement error variables to capture other differences (e.g. one-year and five-year definitions).
- Testing non-linear treatments of time and dependency between intercepts and slopes should be tested.

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Thanks!

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