

## 1 Fertility Trends during Successive Novel Infectious Disease Outbreaks: Zika and Covid-19

2

3

4

Letícia J. Marteleto<sup>1</sup>

5

Luiz Gustavo Sereno Fernandes<sup>1</sup>

6

Raquel Z. Coutinho<sup>2</sup>

7

Sandra Valongueiro Alves<sup>3</sup>

8

Molly Dondero<sup>4</sup>

9

Andrew Koepf<sup>1</sup>

10

Ryan Lloyd<sup>1</sup>

11

Ana Paula Portella<sup>1</sup>

12

13 **Abstract:** The goal of this paper is to estimate fertility trends in Brazil during the 2010s, during  
14 the ZIKV and COVID-19 crises. We use SARIMA forecast, normality test and Box Cox  
15 transformation. We find that fertility trends were stable in the 2011-2015 period, with expected  
16 seasonality. Fertility dropped during the Zika epidemic and rebounded in 2017, after the  
17 epidemic, but increases do not configure a baby-boom. Rather, peak levels never returned to  
18 those from the pre-Zika period. Correcting for Civil Registry data for the most recent period  
19 shows that fertility levels never truly rebounded after the Zika epidemic, and were already  
20 declining when the COVID-19 pandemic hit. We argue that the uncertainties generated by novel  
21 infectious disease epidemics have tangible consequences for fertility that can be even more  
22 dramatic in Brazil because of successive timing of the Zika and COVID-19 emergencies, as well  
23 as economic and political crises during these periods. In the absence of publicly available data  
24 from Brazil's Ministry of Health since 2019, and in light of the cancellation of the 2020 Census,  
25 it is imperative to devise methodologically sound ways to use available data with the goal of  
26 uncovering the consequences of all of the crises overlapping in the country. Further, policies  
27 providing women with the means to meet their fertility intentions become even more important  
28 during periods of public health crises, as resources are shifted from reproductive health care  
29 towards confronting novel disease outbreaks.

30 **Acknowledgments:** This research was funded by grant R01HD091257, Reproductive Responses  
31 to the Zika Virus Epidemic in Brazil, awarded to PI L. J. Marteleto by the Eunice Kennedy  
32 Shriver National Institute of Child Health and Human Development. This research was also  
33 supported by grant P2CHD042849, Population Research Center, awarded to the PRC at The  
34 University of Texas at Austin by the Eunice Kennedy Shriver National Institute of Child Health  
35 and Human Development. This study was conducted under Institutional Review Board approval  
36 #2018-01-0055 from the University of Texas at Austin and the Brazilian National Commission  
37 for Research Ethics (also known as CONEP, or Comissão Nacional de Ética em Pesquisa) study  
38 approval CAAE: 34032920.1.0000.5149.

39

---

<sup>1</sup> University of Texas at Austin

<sup>2</sup> Federal University of Minas Gerais

<sup>3</sup> Federal University of Pernambuco

<sup>4</sup> American University

## 40 **Introduction**

41

42 In March 2020, less than three years after the end of the Zika epidemic, Brazil was hit by Covid-  
43 19. Brazil is currently an epicenter of the pandemic, with more than 400,000 deaths due to  
44 Covid-19, and no end in sight at the time of writing in May 2021<sup>1</sup>.

45

46 Brazil was also the epicenter of the ZIKV (Zika Virus) epidemic and the devastating  
47 consequences of its accompanying surge in congenital Zika syndrome (CZS) in 2015-2017.  
48 Fertility rates declined during the ZIKV epidemic, with the steepest declines occurring roughly  
49 nine months after the link between Zika and fetal malformation was publicized<sup>2</sup>.

50

51 Covid-19 and ZIKV are distinct viruses with different modes of transmission, symptoms, and  
52 effects. However, the novelty of such diseases generates extreme uncertainty over infection risks  
53 and a chaotic prevention response, especially for typically high-risk groups like pregnant people  
54 and their babies.

55

56 Early on in the Zika epidemic, there was a great deal of uncertainty about whether the ZIKV  
57 could be transmitted from a pregnant woman to her fetus, and whether the ZIKV was harmful to  
58 newborns in other ways. Later, fetal transmission was confirmed—along with the risk of severe  
59 fetal abnormalities.<sup>3</sup>

60

61 The specific risk of COVID-19 to pregnant people and their infants is not yet entirely clear.  
62 Early in the pandemic, evidence suggested that pregnancy posed no greater risk to COVID-19. In  
63 June 2020, however, the U.S. Centers for Disease Control added pregnancy to the list of health  
64 conditions that make COVID-19 patients more likely to be admitted to the intensive care unit.  
65 There is now evidence of increases in stillbirth and preterm delivery during the pandemic,  
66 although it is not entirely clear whether these increases result from SARS-CoV-2 infection or  
67 indirect effects such as stress or reluctance to seek medical care.<sup>4</sup> Recent evidence coming from  
68 Brazil also shows higher mortality rates for pregnant and post-partum women<sup>5</sup> and a large  
69 number of newborn deaths directly and indirectly related to Covid-19.<sup>6</sup>

70

71 The goal of this paper is to examine fertility trends in Brazil during the 2010s and early 2020s,  
 72 periods covering pre- and post- Zika epidemic, and the first wave of the COVID-19 pandemic, as  
 73 well as economic and political crises. Combined, these crises have generated a great deal of  
 74 uncertainty in the lives of women of childbearing ages.

75

## 76 **Methods**

77 To examine fertility trends in Brazil during the 2011-2021 period we calculate General  
 78 Fertility Rates (GFRs) using the formula:

$$79 \quad GFR_{ti} = \frac{\text{live births}_{ti}}{\text{pop } w \text{ } 15 - 49_{ti}}$$

80 where  $\text{pop } w \text{ } 15 - 49_{ti}$  is the number of women ages 15-49 in each month and year  $t$  in  
 81 state  $i$ . We use population projections (UFRN).

82 The  $\text{live births}_{ti}$  term is the number of live births in each month and year  $t$  in state  $i$ . We  
 83 use two publicly available datasets on live births—<sup>2</sup>Ministry of Health Sinasc data for 2011-2019  
 84 and civil registration data for 2015-2021. The Sinasc data has a documented 96% coverage of  
 85 births, but the most recent publicly available is for 2019 only, excluding recent trends. Civil  
 86 registry data is the most recent publicly available data on births in Brazil, but presents  
 87 underreporting issues.<sup>7</sup> We employ three techniques to determine and correct for underreporting  
 88 typical of civil registry data and estimate fertility during and beyond 2021.

89 First, we examine differences in live birth between civil registry and Sinasc datasets in  
 90 years these datasets overlap, 2015-2019. We confirm that civil registry data shows significant  
 91 differences from Sinasc data until 2017, and datasets become remarkably similar since then  
 92 (Appendix A).

93 We next run ADF tests for each dataset in Brazil's 27 states in 2017-2019 (Appendix B).  
 94 We determine that live births in both time series datasets are non-stationary processes. Both the  
 95 first difference and the t-1 variation are stationary. We proceed to model t-1 variation.

96 Our third step is to estimate the proportion of Sinasc live births variation (t-1) that can be  
 97 explained by civil registry live births variation (t-1). To identify the states where civil registry  
 98 live births is a reliable predictor of Sinasc live births, we regress Sinasc data on civil registry  
 99 data. Each state regression in each month and year is represented as:

100

101

$$\Delta Bsin_t = \alpha + \gamma \Delta Breg_t + \varepsilon_t$$

102

103 We define two criteria for classifying states according to these regressions modelling the  
 104 difference between datasets. The first criteria is explanatory power. That is,  $\alpha_i$  can't be  
 105 statistically different from zero (cut in 5%), the relationship between  $\gamma$  variables needs to be  
 106 statistically different from zero (cut in 5%) and the variation in civil registry data needs to  
 107 explain at least 50% of the variation in the Sinasc data ( $R^2 > 0.5$ ). The second criteria is based on  
 108 regression residuals; that is, there must be no serial autocorrelation. This is particularly important  
 109 for our analysis because measurement error can result in autocorrelation.

110 Finally, we run Durbin Watson (DW) tests of the regression models and ADFs for the  
 111 residuals. To meet our criteria, the residuals must be stationary (ADF test  $p < 0.05$ ) with zero  
 112 mean and the null hypothesis of no first-order auto-correlation from the DW test must not be  
 113 rejected ( $p\text{-value} > 0.05$ ) (Appendix B).

114 We therefore proceed with calculating the GFR using both datasets. Analyses in  
 115 Appendix C show the GFR for all states, noting the eleven states that meet both of our criteria.  
 116 These states hold 66.62% of the Brazilian population in December 2019 and account for 66.79%  
 117 and 61.92% of all live births between 2017 and 2019. Our analysis shows GFRs for these states.

118

## 119 **Results**

120 The straight black line in Figure 1A shows the GFRs calculated using 2011-2019 Sinasc data,  
 121 and the straight gray line shows the GFRs calculated using 2017-2021 civil registry data for the  
 122 entire country. The dotted black line shows GFRs using Sinasc data and the dotted gray line  
 123 shows GFRs estimated through SARIMA using civil registry data for 13 states. Seven results are  
 124 noteworthy:

- 125 • There is a remarkably steady trend in fertility in the 2011-2015 pre-Zika epidemic period,  
 126 with peaks in March-May;
- 127 • This stable trend is disrupted in 2016, during the Zika epidemic;
- 128 • Fertility returns to pre-Zika levels in 2017, with no signs of a catch up or baby boom  
 129 effect. That is, the births that did not happen in 2016 were not *replaced* in 2017 or 2018.  
 130 Rather, fertility peaks in March-May 2018 and March-May 2019 are lower than peaks in  
 131 previous years, showing an overall decline trend post-Zika.

- 132 • The GFRs calculated using civil registry data show a trend similar to the Sinasc data  
133 trend for the 2017-2019 period—this is particularly true for the 13 states.
- 134 • The 2020 GFRs follow the 2019 GFRs in a decline trend.
- 135 • In the first trimester of 2021, the GFRs declined slightly when compared to the first  
136 trimester of 2020, although it is worth noting that there was already a decline trend.

137

## 138 **Discussion**

139 As we continue the urgent task of documenting the effects of the Covid-19 pandemic, it  
140 remains critical to monitor its demographic consequences beyond mortality. The ZIKV epidemic  
141 had already left recent imprints on fertility in Brazil before the Covid-19 pandemic started. We  
142 show that fertility recovered to levels pre-Zika epidemic in 2017, but that fertility started to  
143 decline again shortly after. This decline has accelerated in 2019, a year before Covid-19 was  
144 reported. We estimate that this trend persists in 2021.

145 Fertility was already in an all-time low in Brazil, following a declining trend before the  
146 Covid-19 pandemic hit. Considering that the spread of Covid-19 through the country has  
147 accelerated in 2021, we argue that the compounding effects of the uncertainties of the Zika and  
148 Covid-19 crises on fertility have the potential to be even more significant in Brazil because of the  
149 back-to-back timing of these outbreaks and its. consequent lack of appropriate time for  
150 recovering from the uncertainties they bring. The pandemic also coincides with significant  
151 economic and political crises, indicating additional effects.

152 While it is difficult to identify the precise mechanisms driving fertility responses during  
153 epidemics, it is clear that the effects go beyond the direct effect of virus mortality. Epidemics can  
154 also affect fertility by inducing behavioral changes because of generalized uncertainties due to  
155 interruptions in health care access and loss of child care provision, for example. As such,  
156 promoting policies aimed at providing women with the conditions to match their fertility  
157 behaviors to their intentions becomes even more important in periods of public health crises,  
158 when resources are shifted from reproductive care towards confronting novel outbreaks. Because  
159 such crises disproportionately affect minority populations, it is also important to focus on  
160 policies aimed at targeting minority groups.

161 As Brazilian women experience successive novel infectious disease outbreaks, the ages at  
162 which they experience these crises will also determine their effects. This is just another facet of

163 the long reach that these successive outbreaks will have on when and how women have children  
164 in Brazil.

165

## 166 Citations

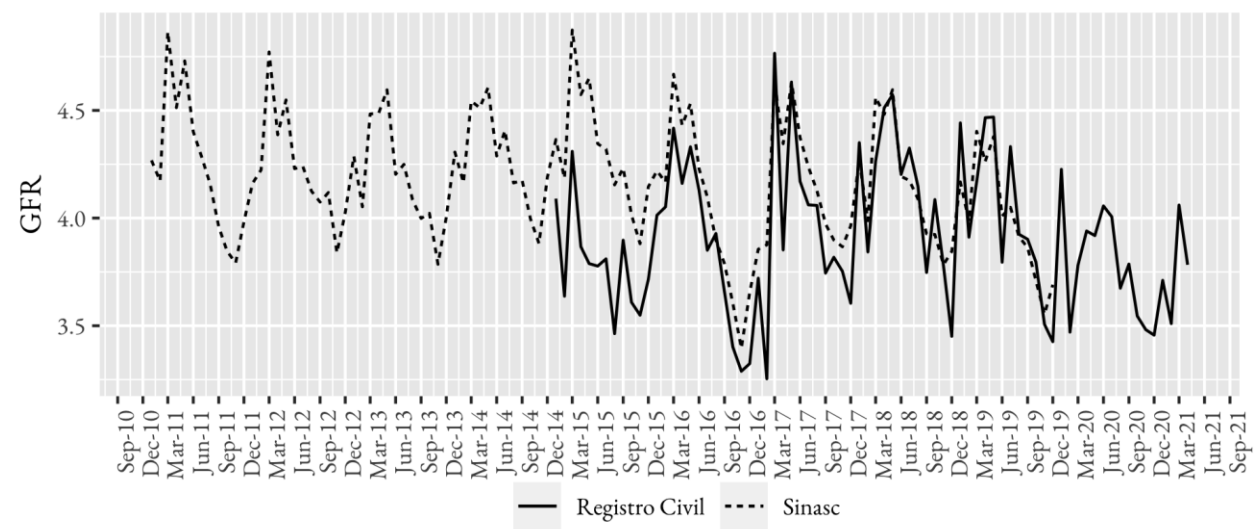
- 167 1. COVID-19 Map. Johns Hopkins Coronavirus Resource Center. Accessed April 28, 2021.  
168 <https://coronavirus.jhu.edu/map.html>
- 169 2. Marteleto LJ, Guedes G, Coutinho RZ, Weitzman A. Live Births and Fertility Amid the Zika  
170 Epidemic in Brazil. *Demography*. 2020;57(3):843-872. doi:10.1007/s13524-020-00871-x
- 171 3. Rasmussen SA, Jamieson DJ, Honein MA, Petersen LR. Zika Virus and Birth Defects —  
172 Reviewing the Evidence for Causality. *N Engl J Med*. 2016;374(20):1981-1987.  
173 doi:10.1056/NEJMSr1604338
- 174 4. Khalil A, von Dadelszen P, Draycott T, Ugwumadu A, O'Brien P, Magee L. Change in the  
175 Incidence of Stillbirth and Preterm Delivery During the COVID-19 Pandemic. *JAMA*.  
176 2020;324(7):705. doi:10.1001/jama.2020.12746
- 177 5. Alex Sandro Rolland Souza, Melania Maria Ramos Amorim. Mortalidade materna pela  
178 COVID-19 no Brasil. *Rev Bras Saúde Materno Infant*. 21(supl 1).  
179 doi:<https://doi.org/10.1590/1806-9304202100s100014>
- 180 6. Kc A, Gurung R, Kinney MV, et al. Effect of the COVID-19 pandemic response on  
181 intrapartum care, stillbirth, and neonatal mortality outcomes in Nepal: a prospective  
182 observational study. *Lancet Glob Health*. 2020;8(10):e1273-e1281. doi:10.1016/S2214-  
183 109X(20)30345-4
- 184 7. COVID-19 - PAHO/WHO Response, Report 42 (25 January 2021) - PAHO/WHO | Pan  
185 American Health Organization. Accessed April 28, 2021.  
186 <https://www.paho.org/en/documents/covid-19-pahowho-response-report-42-25-january-2021>
- 187 8. Portal da Transparência - Registro Civil. Accessed April 28, 2021.  
188 <https://transparencia.registrocivil.org.br/registros>

189

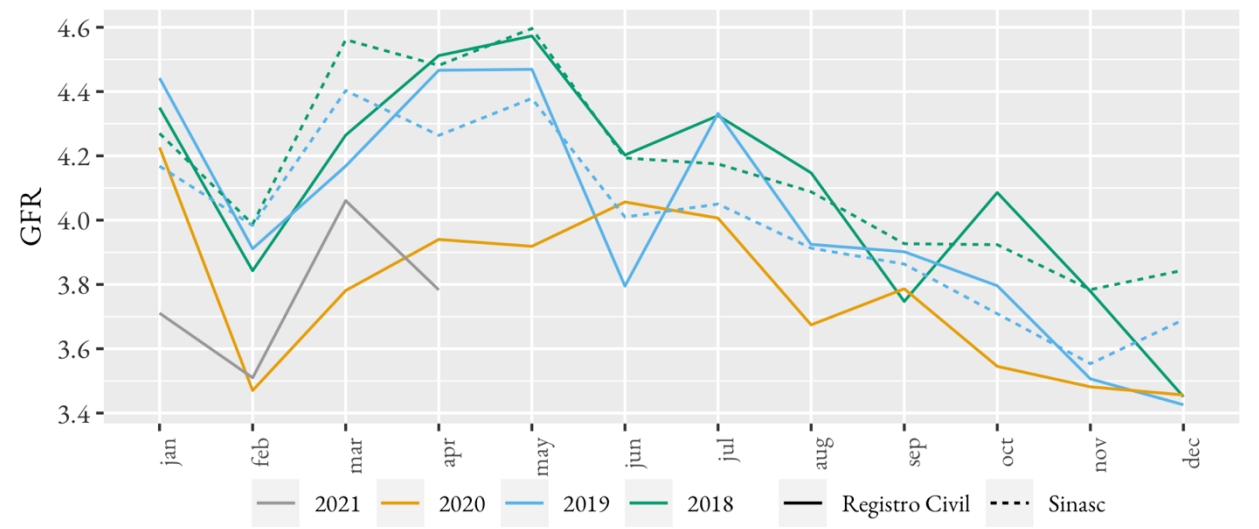
190

191

192 **Figure 1** General Fertility Rate by Month, 2010-2021. Sinasc & Civil Registry Data. Brazil.



193



194  
195

196  
197  
198  
199

Appendix A

		<p>...</p>

200  
201  
202  
203  
204  
205

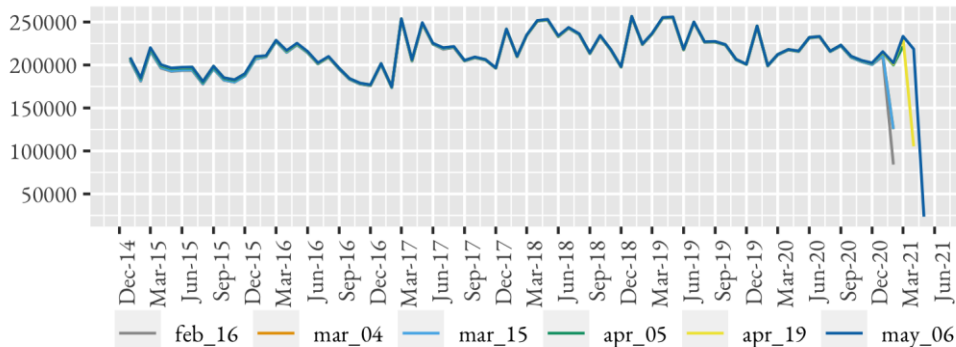
Figure 1 for state?



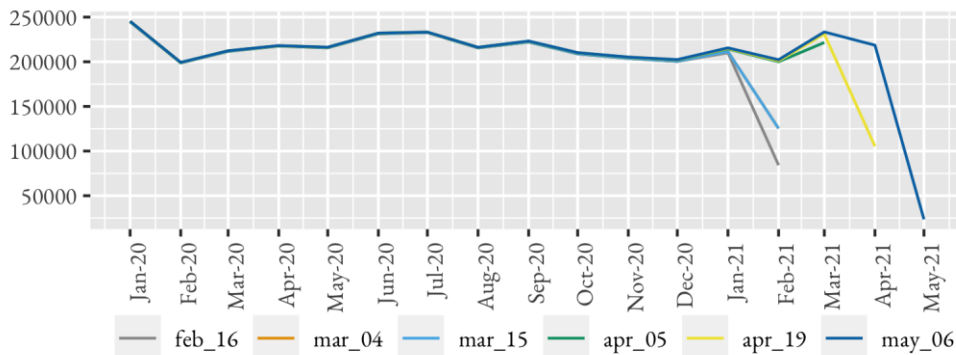
206  
207  
208  
209

Appendix B

Number of Live Birth from Civil Registry on different data collection dates, Brazil



210



211  
212

213

214

215 Estimações para os 27 estados

	Sinasc	Civil Registry	$\alpha$	$\gamma$		Residuals	Residuals	Res(t-1) p-	
UF	ADF	ADF	(p value)	(p value)	R2	mean	ADF	values	Escolha
11	0.010	0.010	0.706	0.000	0.616	0.000	0.010	0.120	1
12	0.010	0.010	0.965	0.000	0.421	0.000	0.010	0.001	0
13	0.010	0.010	0.657	0.014	0.170	0.000	0.010	0.130	0
14	0.010	0.010	0.634	0.446	0.018	0.000	0.010	0.018	0
15	0.010	0.010	0.708	0.000	0.479	0.000	0.010	0.102	0
16	0.035	0.010	0.890	0.011	0.182	0.000	0.010	0.007	0
17	0.010	0.010	0.959	0.000	0.496	0.000	0.010	0.642	0
21	0.010	0.010	0.888	0.001	0.296	0.000	0.010	0.281	0
22	0.010	0.010	0.858	0.031	0.133	0.000	0.010	0.247	0
23	0.013	0.010	0.818	0.000	0.482	0.000	0.010	0.346	0
24	0.010	0.010	0.740	0.000	0.465	0.000	0.010	0.146	0
25	0.011	0.010	0.870	0.000	0.642	0.000	0.010	0.053	1
26	0.022	0.010	0.876	0.000	0.595	0.000	0.010	0.087	1
27	0.016	0.010	0.865	0.000	0.569	0.000	0.010	0.404	1
28	0.041	0.010	0.954	0.000	0.352	0.000	0.010	0.405	0
29	0.046	0.010	0.560	0.000	0.560	0.000	0.010	0.094	1
31	0.025	0.010	0.900	0.000	0.582	0.000	0.010	0.210	1
32	0.020	0.010	0.940	0.000	0.576	0.000	0.010	0.004	0
33	0.042	0.010	0.815	0.000	0.723	0.000	0.010	0.307	1
35	0.037	0.010	0.893	0.000	0.661	0.000	0.025	0.116	1
41	0.027	0.010	0.870	0.000	0.712	0.000	0.010	0.023	0
42	0.010	0.010	0.837	0.000	0.430	0.000	0.010	0.064	0
43	0.010	0.010	0.667	0.000	0.602	0.000	0.010	0.052	1
50	0.050	0.010	0.923	0.000	0.667	0.000	0.019	0.468	1
51	0.010	0.010	0.596	0.000	0.523	0.000	0.010	0.150	1
52	0.010	0.010	0.918	0.000	0.587	0.000	0.010	0.047	0
53	0.010	0.010	0.312	0.000	0.702	0.000	0.010	0.000	0

216