

Visualisation of multinomial multilevel time-series modelling

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Research Goal

- Survey data are reliable at the national level, but detailed sub-national estimates are often unreliable due to low or zero sample sizes.
- Small area estimation (SAE) is a statistical technique for producing reliable, detailed-level statistics when data are limited.
- We illustrate the approach using antenatal care (ANC) coverage in Bangladesh, categorizing pregnant women by 0, 1–3, or 4+ ANC visits, with WHO recommending at least four visits.
- A multinomial multilevel model captures the probabilities of each ANC category across districts and years, accounting for both spatial and temporal correlations.
- Incorporating covariates such as night-time light intensity improves prediction and highlights local disparities over time.
- A spatio-temporal SAE model is applied to estimate district-level ANC trends from 1994–2022, demonstrating the flexibility of multinomial multilevel time-series modelling.

Parameter Estimates

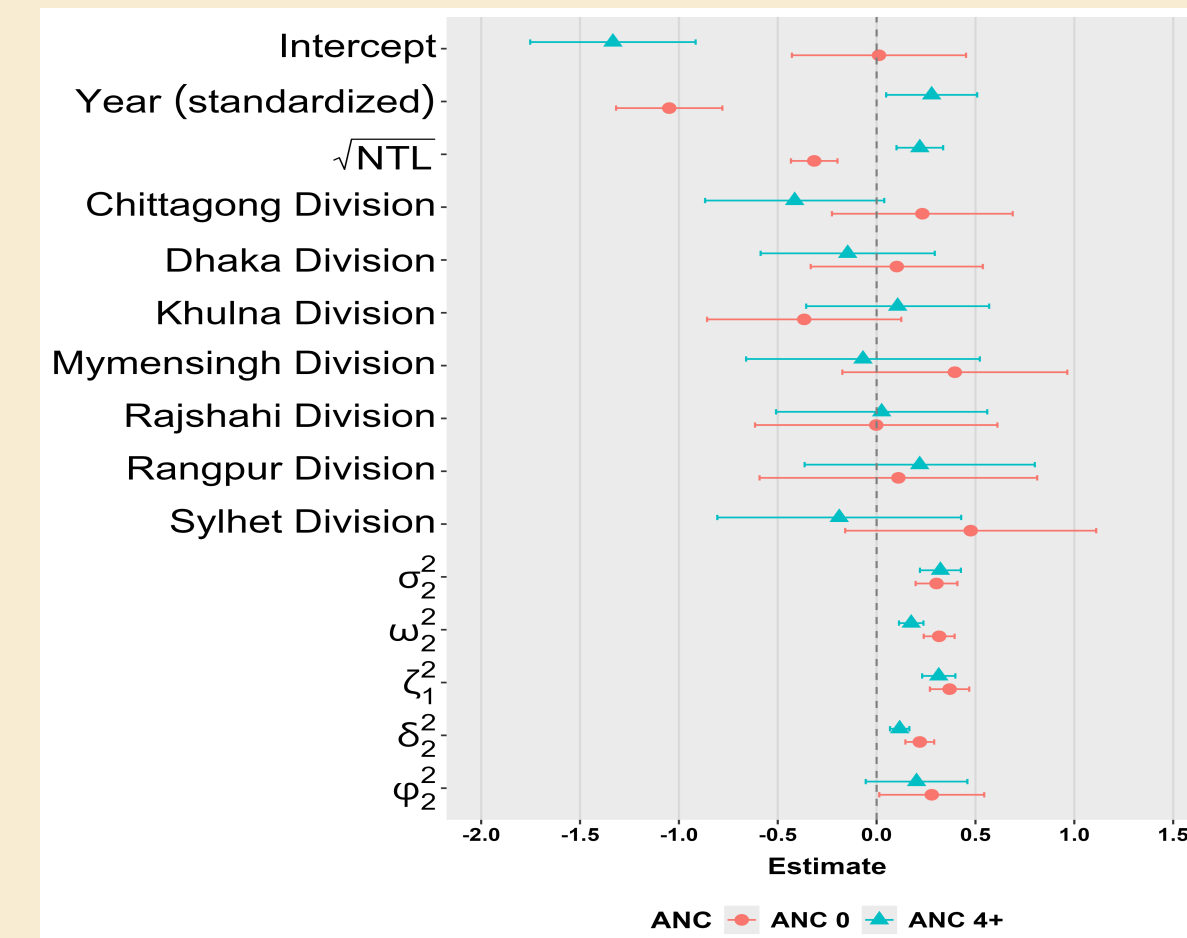


Fig. 2: Fixed (log odds scale) and random effects for ANC0 and ANC4+ under the multinomial multilevel model. Variance components correspond to district- and division-level temporal, spatial, and year-district random effects.

Level	Component 1	Component 2	Mean	t-value	$\hat{R}$
District Int.: ANC0	Int.: ANC4+	-0.709	-5.519	1.007	
Dist Int.: ANC0	Slope: ANC0	0.360	2.198	1.000	
Dist Int.: ANC4+	Slope: ANC0	-0.155	-0.855	1.000	
Dist Int.: ANC0	Slope: ANC4+	-0.194	-0.829	1.000	
Dist Int.: ANC4+	Slope: ANC4+	-0.004	-0.019	1.000	
Div Slope: ANC0	Slope: ANC4+	-0.497	-2.476	1.000	
Div RW1: ANC0	RW1: ANC4+	-0.297	-1.722	1.003	
Dist RW2: ANC0	RW2: ANC4+	-0.796	-8.476	1.022	

Tab. 1: Correlation between random effects under different random effects components defined for the two outcome variables $ANC0$ and $ANC4+$ under the multinomial multilevel model. Posterior means, t -values, and Gelman–Rubin $\hat{R}$ statistics are reported. Statistical significance (in **bold**) is indicated by $|t| > 1.96$, and convergence of MCMC chains is assumed confirmed when $\hat{R} < 1.1$.

Model Diagnostics

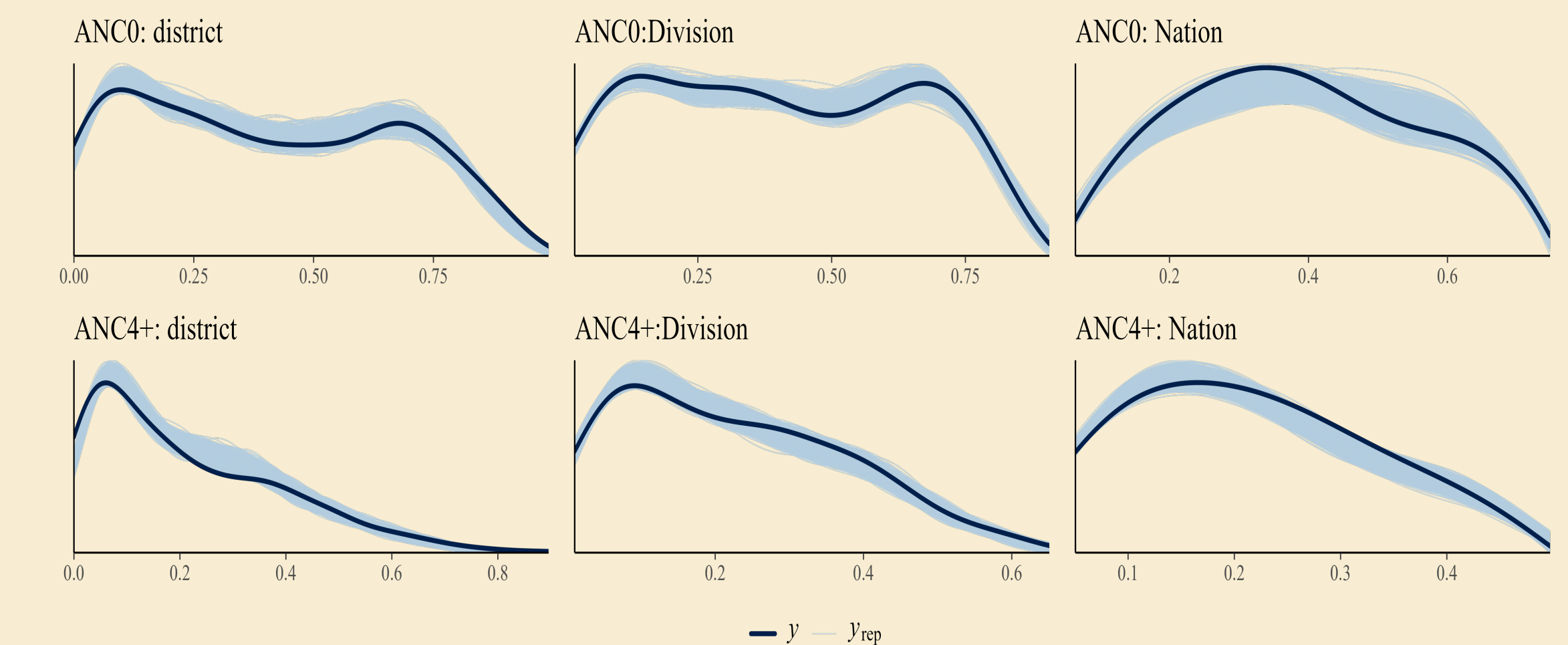


Fig. 5: Posterior predictive check (PPC) comparing model-based simulated draws (thin lines) with kernel densities of direct estimates (thick lines) at district, division, and national levels. The model smooths noisy district-level trends while remaining consistent with higher-level direct estimates.

ANC Coverage and Nightlights

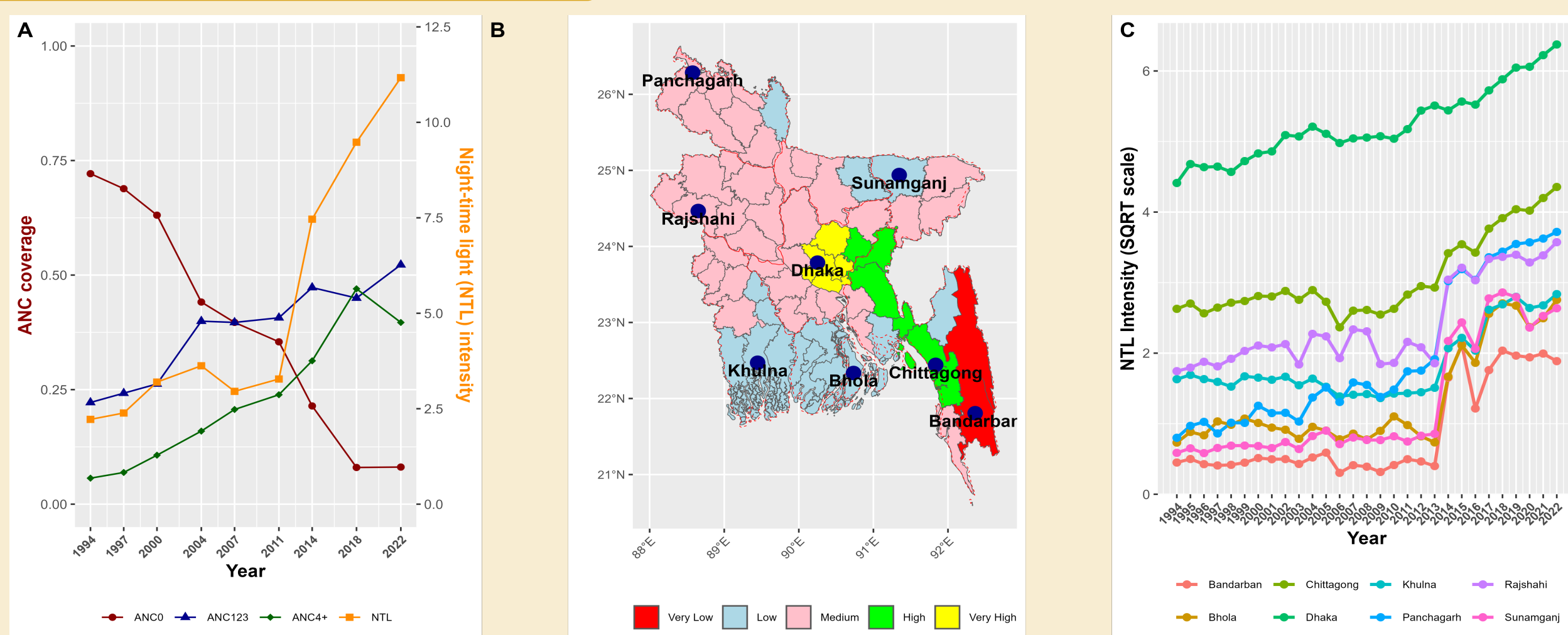


Fig. 1: Country-level trends in ANC coverage and night-time light (NTL) intensity (A); district-level NTL intensity in 2024 (B); and temporal trends for selected districts (C) in Bangladesh. The country comprises 8 divisions (admin-1) and 64 districts (admin-2).

National Trends

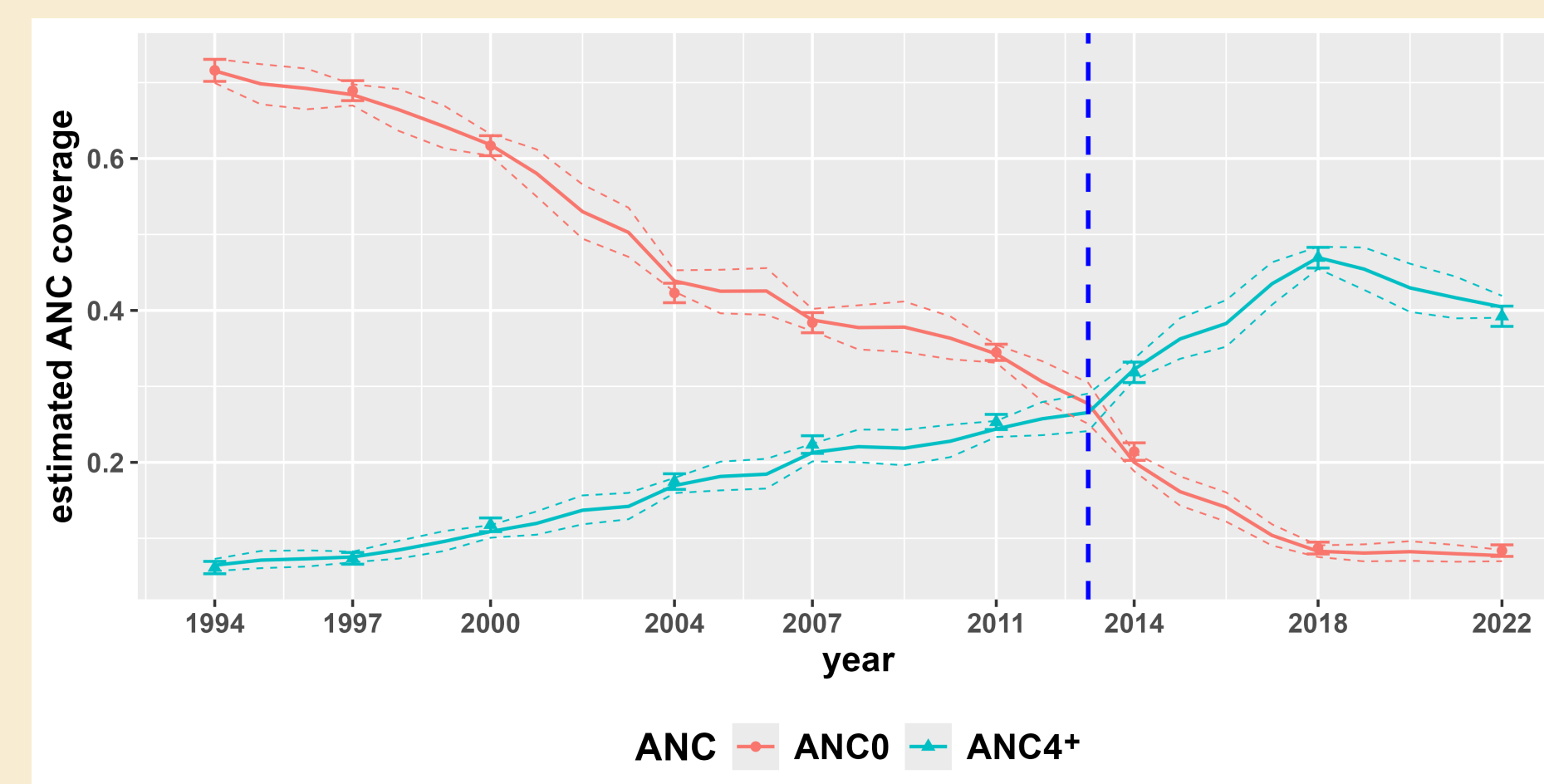


Fig. 3: National level trends in the coverage of 0 and 4+ ANC visits for pregnant women in Bangladesh during 1994–2022. Solid line shows mean of posterior distribution and 95% credible interval. Direct estimates and 95% confidence intervals shown by dots and whiskers.

Conclusion

- The **multinomial multilevel time-series model** enables the joint analysis of multiple correlated outcomes ($ANC0$, $ANC123$, and $ANC4+$), capturing dependencies across outcomes.
- Its hierarchical structure borrows strength across space, time, and correlated outcomes, producing stable estimates even for data-sparse districts.
- Model-based estimates maintain numerical consistency across aggregation levels and offer greater precision compared to direct survey estimates.
- Visualising ANC trends within a multinomial framework enhances interpretation of complex spatio-temporal patterns in maternal healthcare access.
- The Bangladesh ANC case study highlights districts exhibiting progress, stagnation, or decline in coverage, with patterns shaped by socioeconomic gradients proxied through night-time light intensity.

Model

- The multinomial multilevel model was specified for counts $\hat{Y}_{it}^{(k)}$ for domains $i = 1$ to 64, years $t = 1994$ to 2022, and outcome categories $k = 1, 2, 3$ corresponding to $ANC0$, $ANC123$, and $ANC4+$, with category $k = 2$ chosen as the reference. The model can be expressed in its simplest form for domain i as

$$\hat{Y}_{it} \sim \text{Multinomial}(\eta_{it}, P_{it}), P_{itk} = \frac{e^{\eta_{itk}}}{\sum_{k'=1}^K e^{\eta_{itk'}}}, \eta_{itk} = X_{it}\beta_k + \sum_{\alpha} Z_{it}^{\alpha} v^{\alpha}$$

$$\log \left(\frac{P_{it}^{(1)}}{P_{it}^{(k)}} \right) = \beta' X_{it} + \nu_i + \nu_i^{(gr)} + u_{it}^{(Div)} + u_{it}^{(Dist)} + s_i$$

- Random intercepts** at district-year level: $\nu_i \sim N(0, \sigma_{\nu}^2)$,
- Random slopes** (time trends) at district level: $\nu_i^{(gr)} \sim N(0, \omega_{\nu}^2)$,
- Temporal effects** at division level $u_{it}^{(Div)} \sim N(0, \zeta_1^2)$ & at district level $u_{it}^{(Dist)} \sim N(0, \delta_2^2)$,
- Spatial effects** at district level: $s_i \sim N(0, \varphi_2^2)$.
- The model allowed correlated district- and division-level random effects to capture dependence between outcome categories $ANC0$ and $ANC4+$.

Division and District Trends

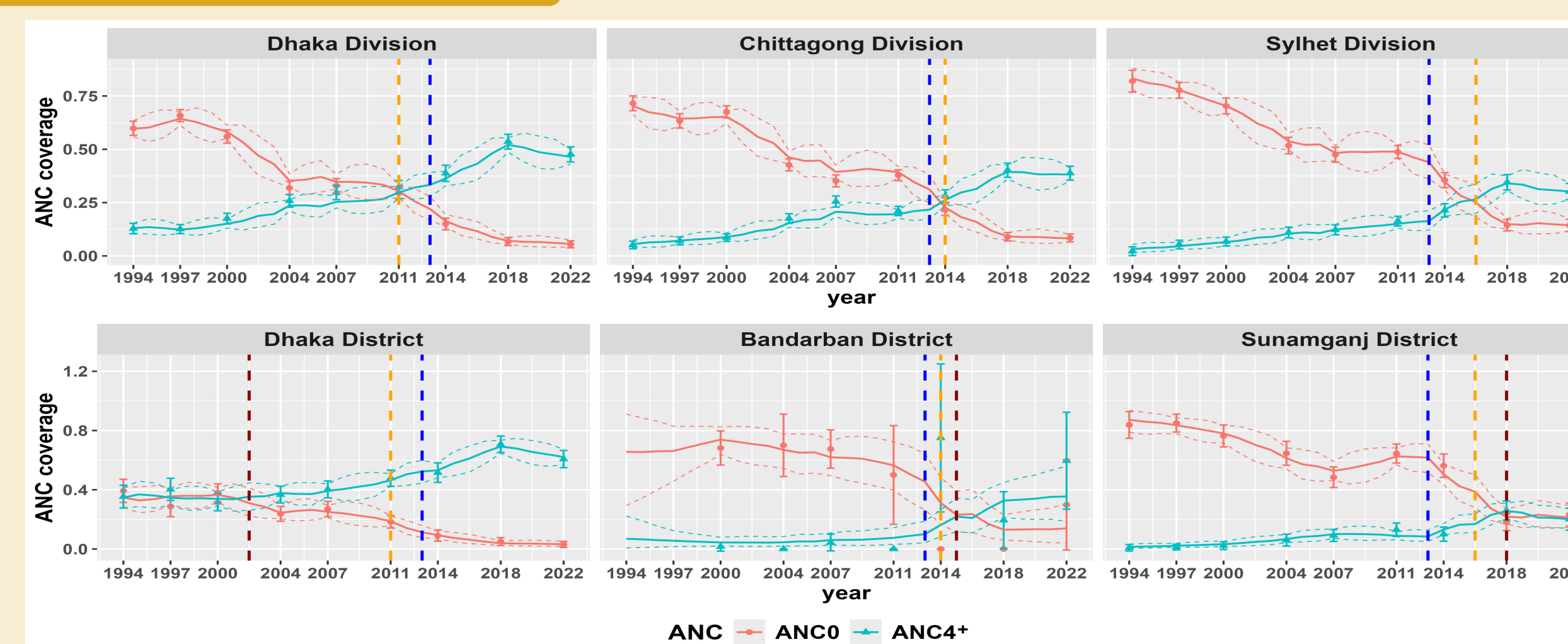


Fig. 4: Selected division and district level trends in the coverage of 0 and 4+ ANC visits for pregnant women in Bangladesh during 1994–2022. Vertical lines represent National (blue), Division (orange) and District (purple) level achievement.

More Information



Health worker Rashida checking an expecting mother's health condition in Gaibandha, Bangladesh. (C) BRAC /Sumon Yusuf 2018.

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