

Antecedents and Spatial Variation of the Dual Burden of Childhood Stunting and Underweight in India: A Copula Geoadditive Modelling Approach

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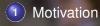


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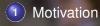


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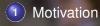


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Motivation

- India has one of the highest burdens of childhood undernutrition in the world.
- Current estimates ⇒ 36% stunting, 17% underweight & 6% wasting among under 6 children.
- Extant research: analysis of antecedents & spatial heterogeneity for each of these metrics separately.
- Above dimensions can be significantly correlated in a particular population ⇒ valuable insights regarding penetration and spread of dual burdens of malnutrition.
- Proposed research: quantification of spatial distribution and critical drivers of the dual burden of stunting and underweight using copula-geoadditive modeling approach.

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Methodology

- Copulas ⇒ separate marginal model specification of responses from the joint distribution governing their dependence structure.
- $Y_{is}(Y_{iu}) = 1(0)$ if i^{th} child is stunted (underweight) \Rightarrow

$$P(Y_{is} = 1, Y_{iu} = 1 | \boldsymbol{x}_{is}, \boldsymbol{x}_{iu}) = C(P(Y_{is} = 1 | \boldsymbol{x}_{is}), P(Y_{iu} = 1 | \boldsymbol{x}_{iu}); \nu)$$

 $C: [0,1]^2 \rightarrow [0,1]$: copula function, ν : copula parameter \Rightarrow quantifies stunting-underweight association.

• Copula geoadditive models accommodates the following effects:

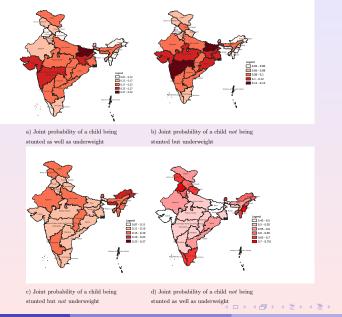
- Linear as well as flexible, non-linear effects of predictors.
- Within-region (unstructured) & between-region (unstructured) spatial effects of stunting & underweight.
- Spatial heterogeneity in the copula parameter.
- Non-linear effects ⇒ thin-plate regression splines; structured spatial effects ⇒ Markov random-field smoother.

Results

- CGM was applied on data from 1,04,021 children from National Family Health Survey-5 (NFHS-5) dataset.
- χ^2 test \Rightarrow significant stunting-underweight association.
- Significant spatial heterogeneity in dual burden of stunting and underweight ⇒ highest (lowest) prevalence in eastern & western states (northern & southern states).
- Significant between-state variation in underweight ⇒ lowest in north-east, highest in central and western India.
- Boys, low birthweight, normal delivery, no prenatal checkup, maternal short-stature & underweight, anemia ⇒ significantly higher odds of stunting & underweight.
- Child age, age at first birth & duration of breast feeding ⇒ significant non-linear association with stunting and underweight.

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Joint probability maps



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Non-Linear Associations

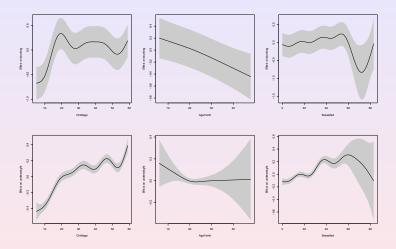


Figure 1: Estimated non-linear effects of child age, maternal age at first birth and duration of breastfeeding on the likelihood of stunting (top row) and underweight (bottom row). Shaded regions correspond to 95% confidence bands

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Discussion

- One of the largest studies that jointly models the dual burden and geographical variation of stunting & underweight in India.
- Moderately strong stunting-underweight association and significant spatial heterogeneity in this association across India.
- Significant within-state variation in the prevalence of stunting & underweight ⇒ district-specific analysis.
- Necessity of a nuanced, region-specific nutritional intervention plan to effectively tackle childhood undernutrition in India.



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