The utility of standardized or crude weight measures in modeling of postnatal growth trajectories: Are there differences?

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Introduction

Analyses of growth trajectories are expanding in tandem with the growing interest in life course epidemiology. Z-scores are a frequent choice when modeling weight growth trajectories to standardize the sample to usually what is the CDC reference or WHO standard populations. When used for cross-sectional data, Z-scores have advantages including linear sex- and age-independent measures of weight outcomes. However, there has been no comparison of model performance for outcomes including Z-scores, or an equivalent such as percentiles, and crude weight when studying weight trajectories in infancy.

Aim

Use simulations to assess differences in power, type I error measures, and coefficient estimates of weight change differences across three different outcome measures in child anthropometric measures: weight, weight Z-score and weight percentiles.

Samples

Growth trajectory parameters for baseline exposure correspond to published estimates for three samples: Italy, Portugal and Chile.

C. Pizzi et al. "Prenatal Influences on Size, Velocity and Tempo of Infant Growth: Findings from Three Contemporary Cohorts". In: *PLoS ONE* 9.2 (Feb. 27, 2014). Ed. by G. Wang, e90291. DOI: 10.1371/journal.pone.0090291

Method

We simulated infant growth data using a Reed first order parametric model:

$$y = \beta_0 + \beta_1 \cdot t + \beta_2 \cdot \ln(t) + \frac{\beta_3}{t}$$

After simulating data, three models were fit with the simulated data:

Model 1

 $y_{ij} = \beta_0 + \beta_1 \cdot t + \beta_2 \cdot \text{group} + \beta_3 \cdot t \cdot \text{group} + e_{ij}$ Model 2

 $y_{ij} = \beta_0 + \beta_1 \cdot t + \beta_2 \cdot \operatorname{group} + \beta_3 \cdot t \cdot \operatorname{group} + \beta_4 \cdot t^2 + e_{ij}$ Model 3

 $y_{ij} = \beta_0 + \beta_1 \cdot \text{month.6} + \beta_2 \cdot \text{group} + \beta_3 \cdot \text{month.6} \cdot \text{group} + e_{ij}$

Model Terms

the outcome: weight, Z-score or percentile

group a binary exposure factor month.6 a binary variable for time with

1=month 6 and 0=month 0 e_{ij} error term with a autocorrelation structure, ρ =0.5 and σ =0.75

Results

Figure 1: Simulated weight growth curves with corresponding Z-score and percentile outcomes using Chilean growth parameters

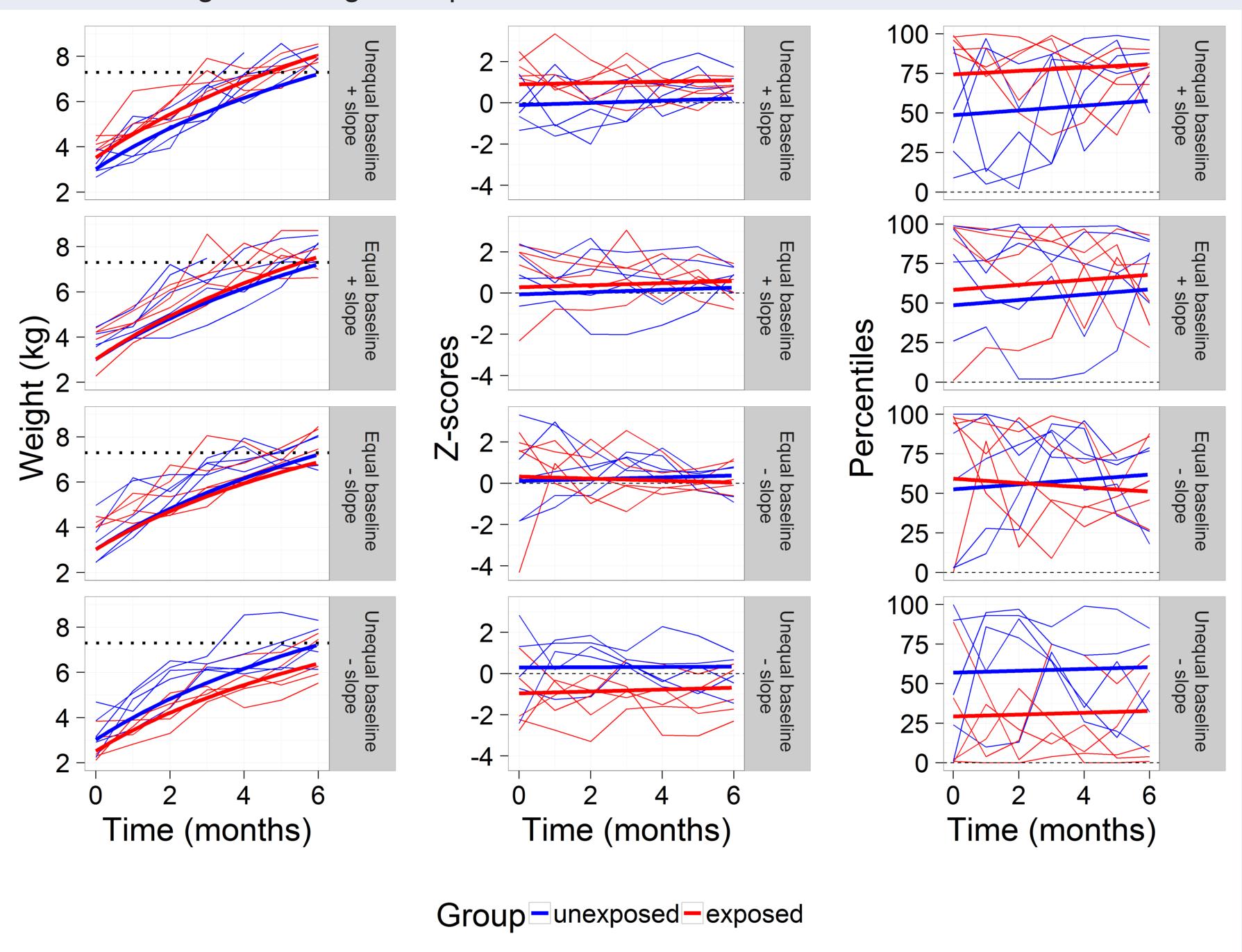
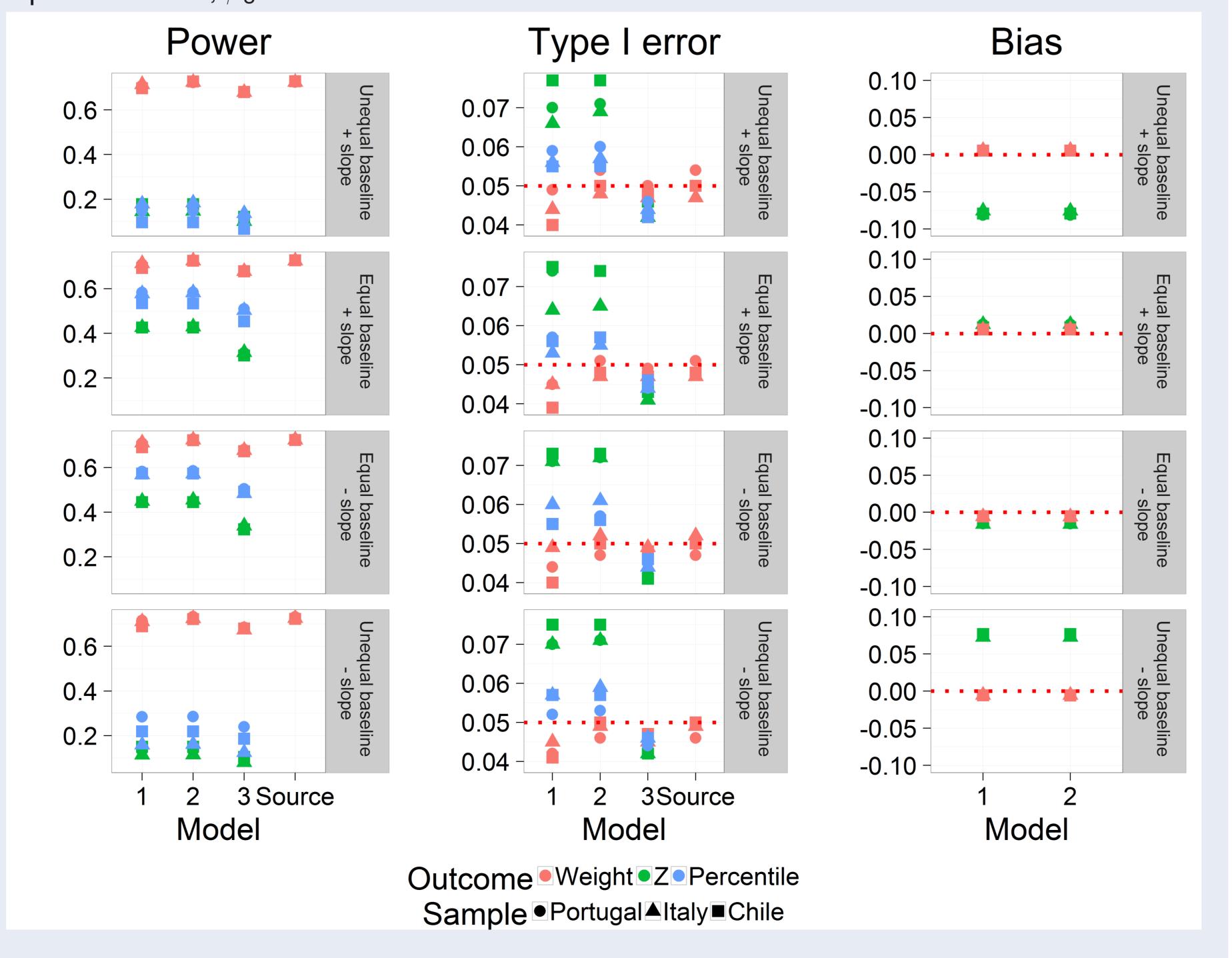


Figure 2: Comparison of power, type-I error and bias for product term for time and exposure effect, β_3 .



Results, cont...

Power Consistently greater with crude weight.Weight difference at baseline results in larger

Type I error Crude weight measure closest to nominal Type I error of 0.05.

 Z-scores have greatest type I error for models 1 and 2.

Bias Baseline weight differences lead to opposite direction of effect for Z-score vs crude weight outcomes.

• No weight differences at baseline leads to similar estimates between Z-score and original weight outcomes.

Summary

differential.

Z-scores or percentiles These measures provide a measure of weight relative to a referent population.

• Trajectory group differences using Z-scores or percentiles as outcomes can lead to estimates in opposite direction compared crude weight outcomes.

 Using these measures also lead to lower power to detect differences in slope and inflated type I error.

Conclusions

Each measure examined in this project, Z-scores, percentiles and crude weight measures, serves a purpose as needed in research. However, these outcomes in growth trajectory analyses are not equivalent.

• Transforming crude weight measures into Z-scores, as is commonly done in the literature, may provide benefits such as a reference point to a standard and comparability across genders, but this measure does not function the same as crude weight when examining growth trajectories.

• Simulations presented here provide evidence for this conclusion when baseline differences in weight exists across groups.

• If reference to a standard is not a priority when examining group differences in growth trajectory patterns, the best approach is to use the crude weight measures.

 In the context of growth trajectories serving as exposures in life course analyses, use of Z-scores can lead to exposure misspecification and conflicting inference.

