

Causal Treatment Effect Aggregation in Boundary Discontinuity Designs*

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Abstract

This paper analyzes two distinct approaches for estimating aggregated average treatment effects in boundary discontinuity designs: the widely-used *pre-aggregation* method, and a novel *post-aggregation* method. The pre-aggregation approach bundles all units near a treatment assignment boundary first, and then estimates a single, aggregated local average treatment effect. While commonly used in empirical work, its formal statistical properties remain underexplored. In contrast, the post-aggregation approach first estimates heterogeneous treatment effects along the assignment boundary, and then aggregates them, offering flexibility in weighting and robustness to boundary irregularities. We provide a precise definition of the aggregated causal parameters identified by each approach, and establish valid estimation and inference methods based on them. The theoretical results elucidate the relative merits of each aggregation approach, providing practical guidance for past and future applications of boundary discontinuity designs. General-purpose companion software is provided, and illustrated using an empirical application.

Keywords: regression discontinuity, treatment effects estimation, causal inference.

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