Supporting figures for

Dynamic relational event modeling: Testing, exploring and applying

Marlyne Meijerink-Bosman, Roger Leenders and Joris Mulder
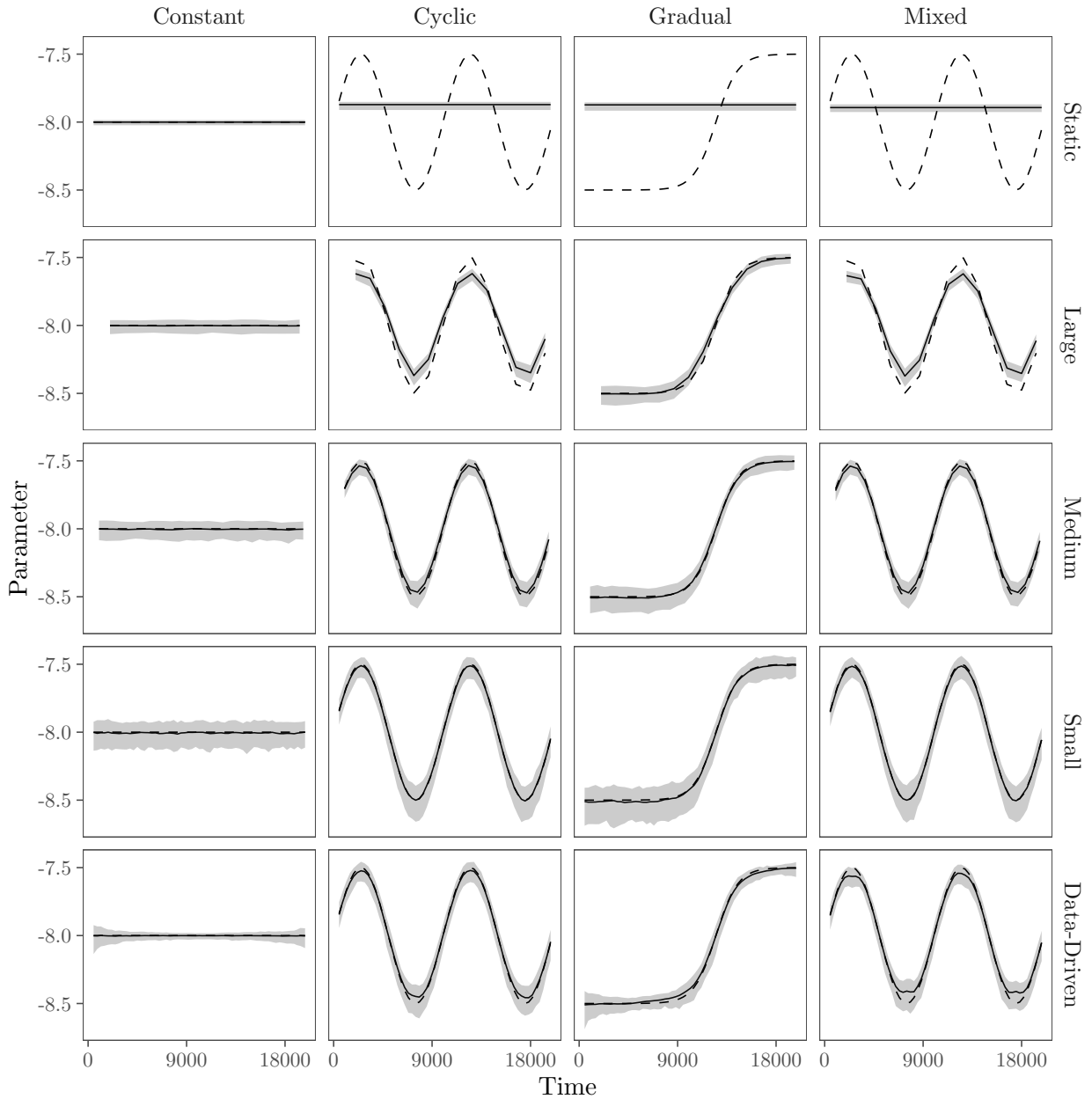
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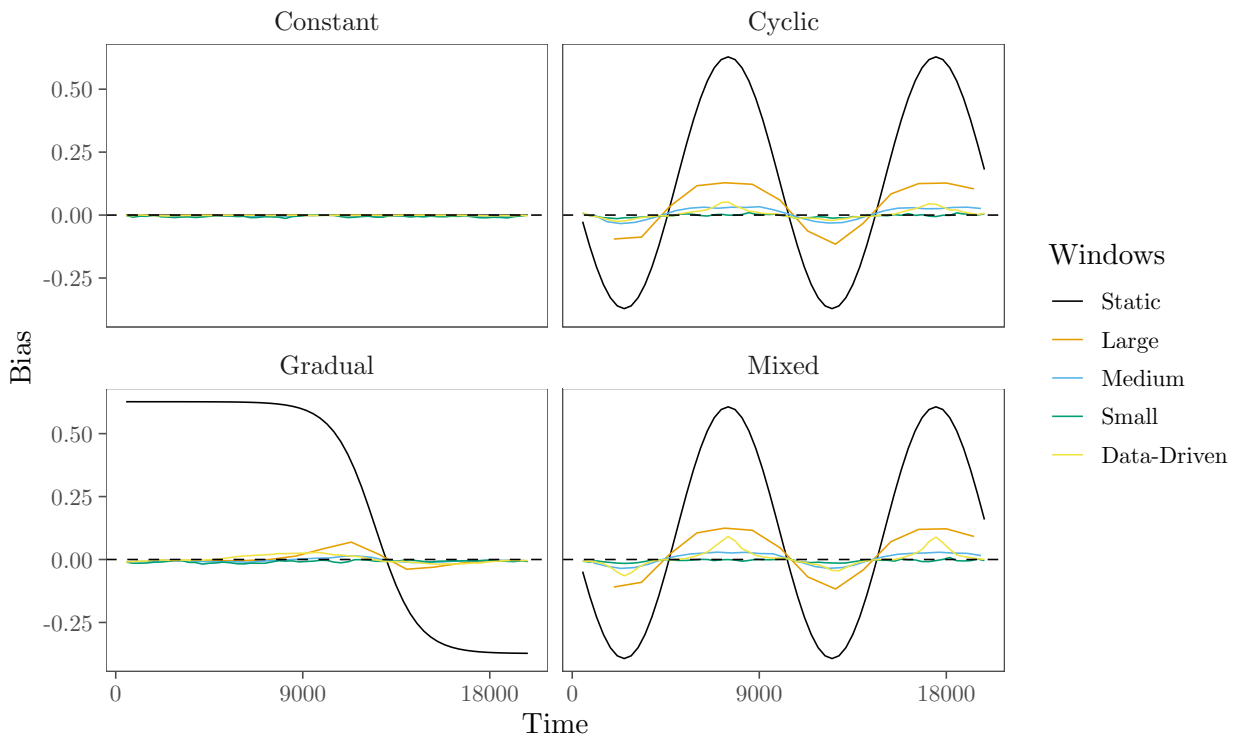
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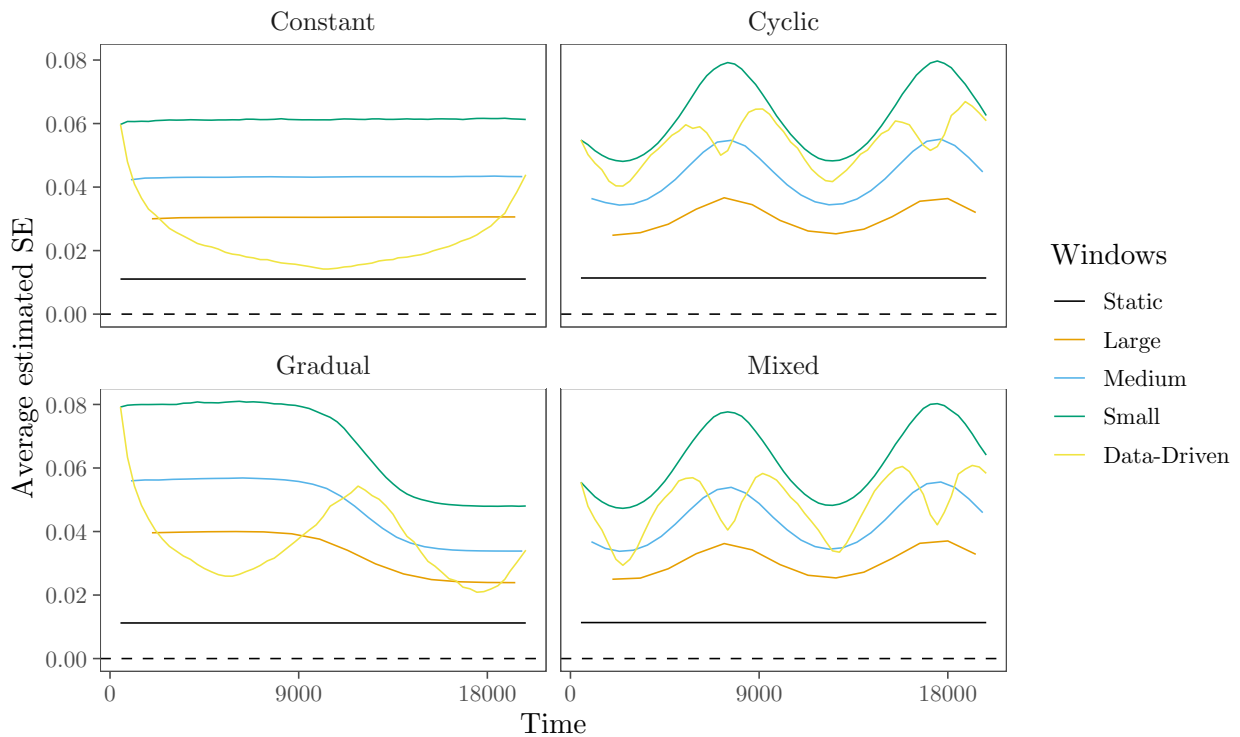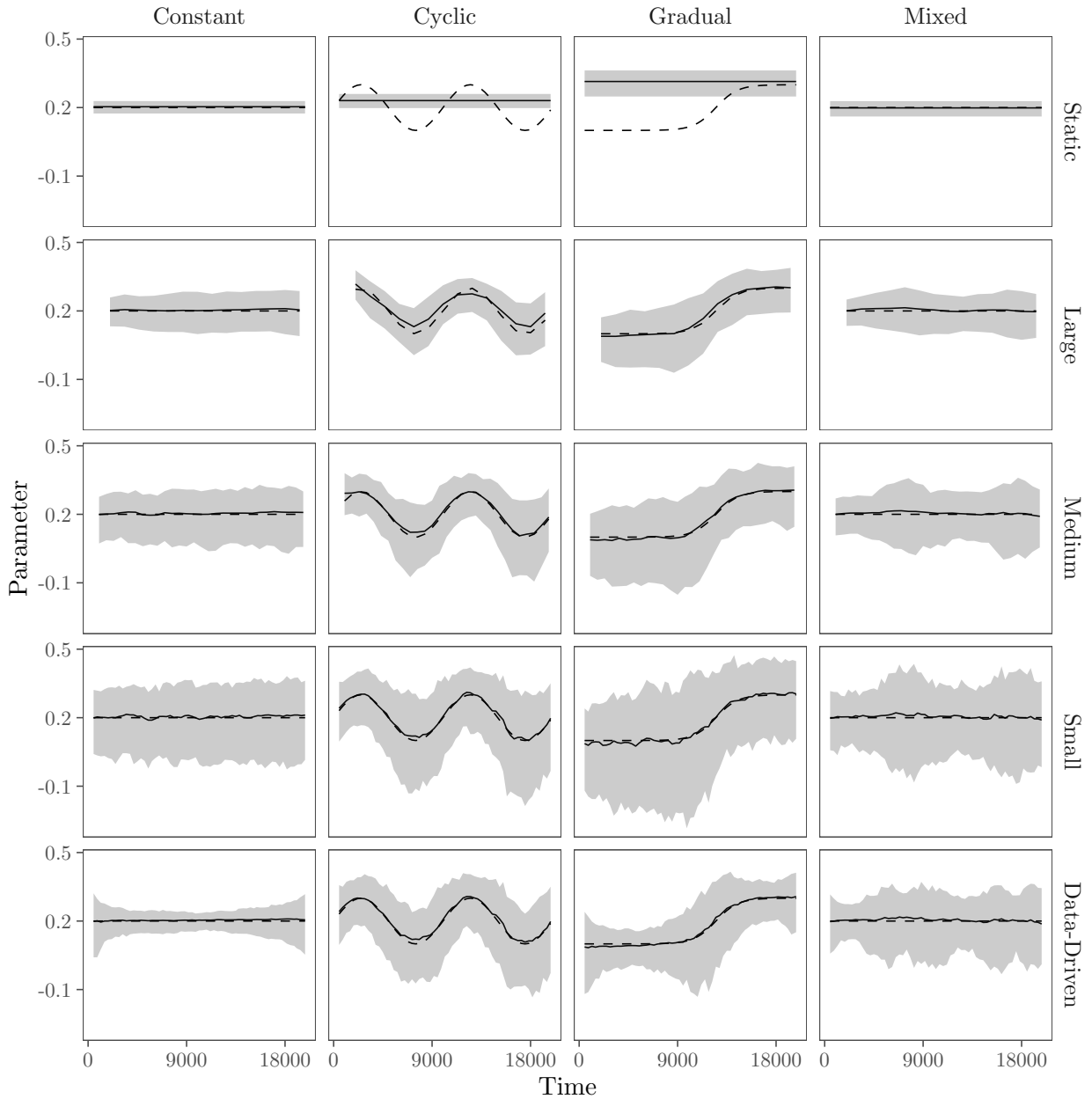
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parameters over 200 datasets over time. The gray area represents the range with 95% of the estimates for the
200 datasets. Dashed lines represent the parameters used for data generation.
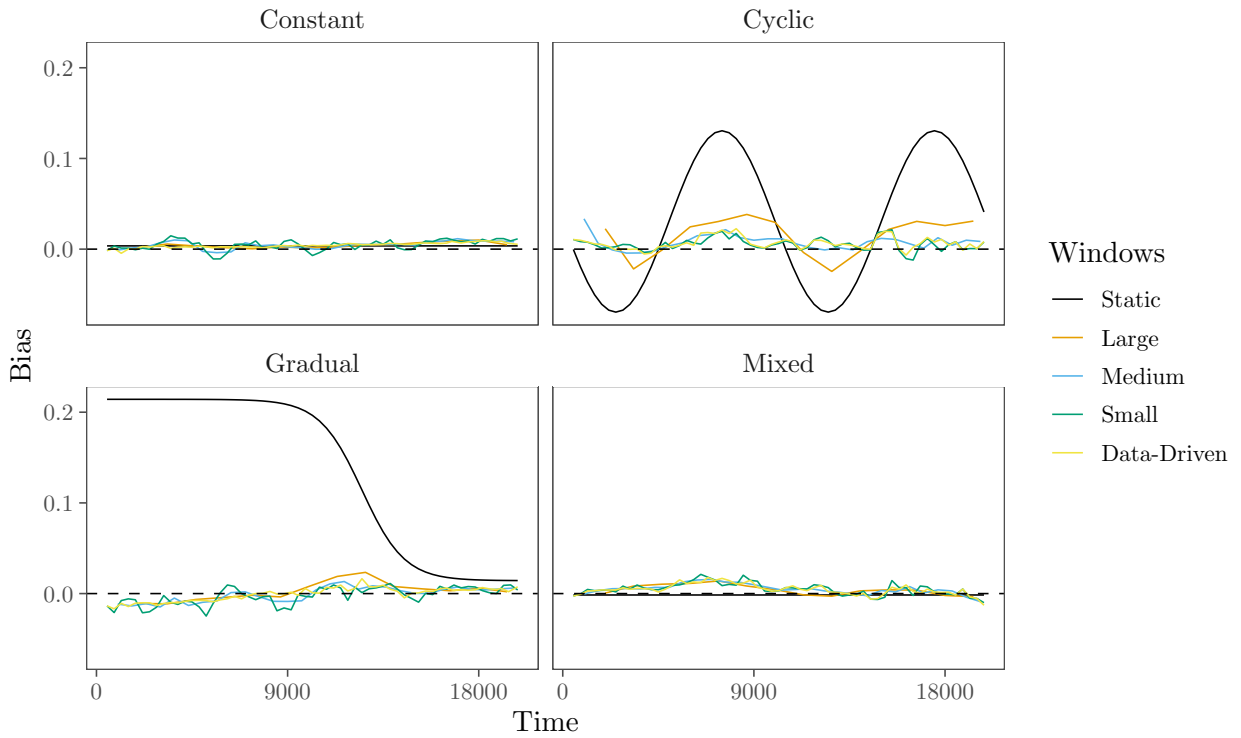
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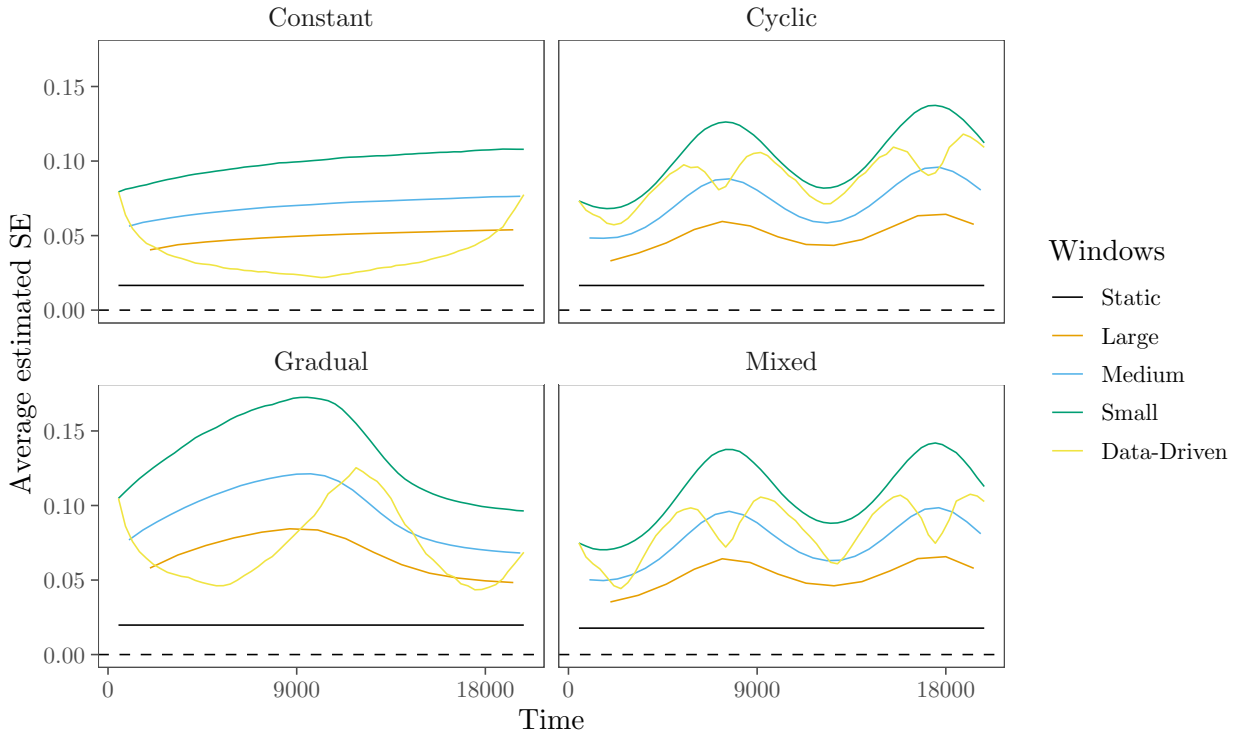
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Panels refer to the four time-varying effect scenarios. Solid lines represent the average estimated standard error of the parameter estimates over time, with colors representing estimation with large $(4000t)$, medium $(2000t)$, small $(1000t)$ and data-based window widths.
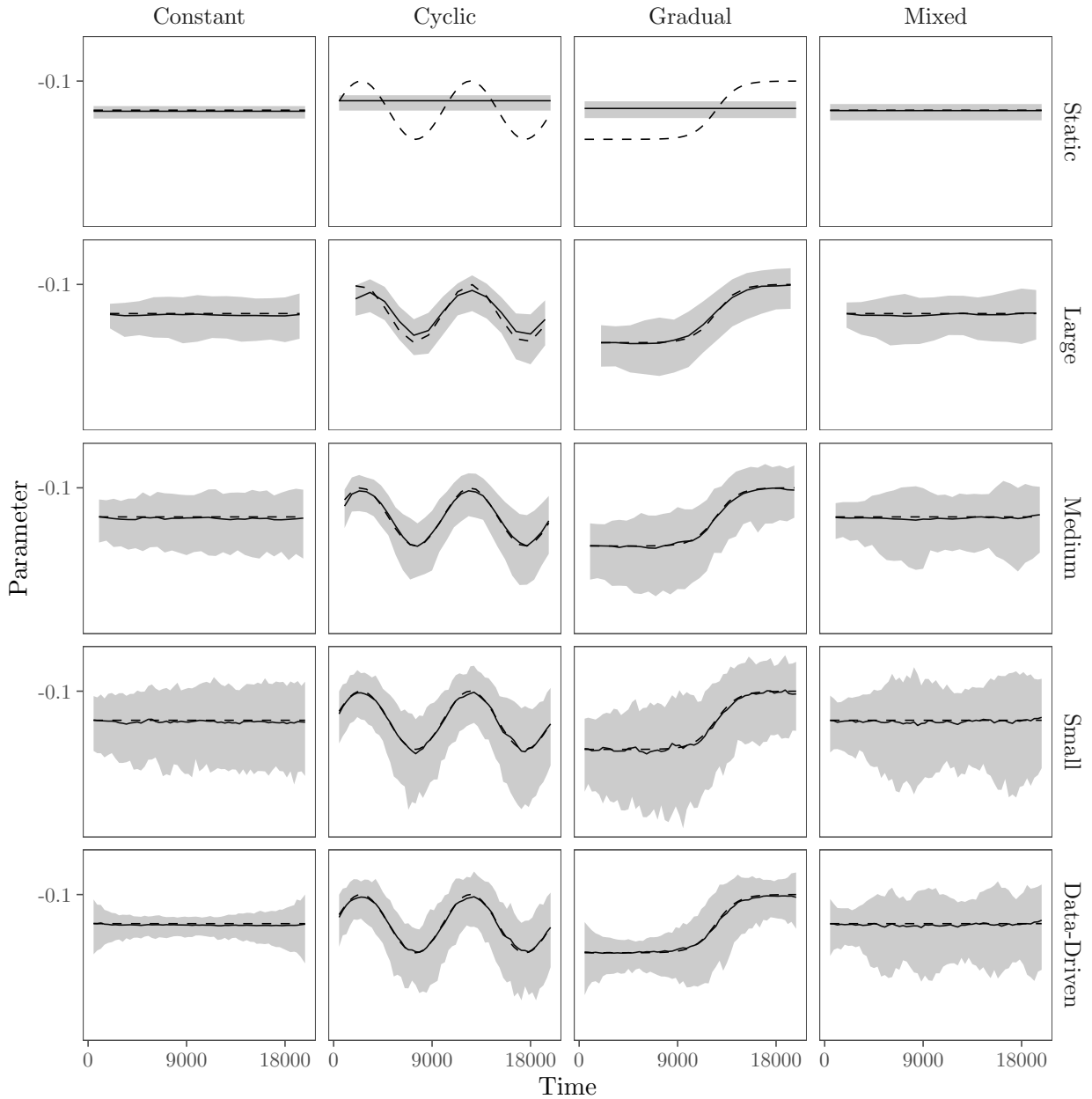
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200 datasets. Dashed lines represent the parameters used for data generation.
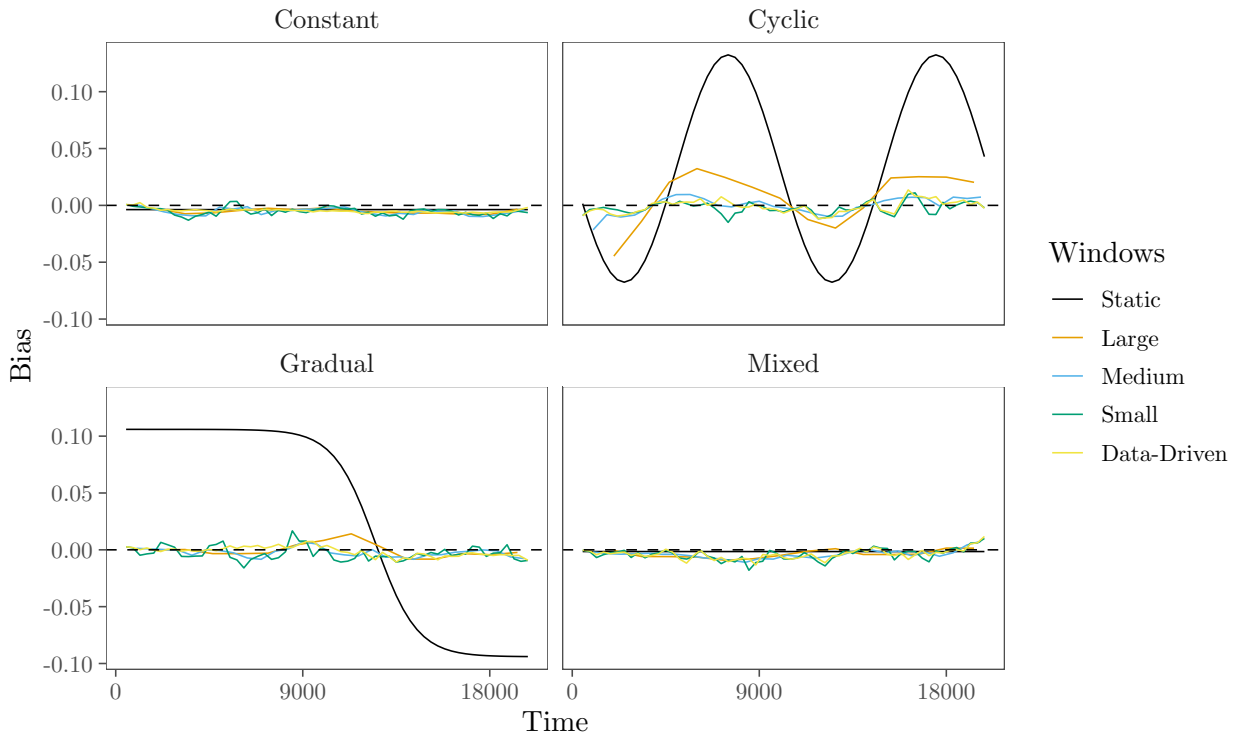
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Panels refer to the four time-varying effect scenarios. Solid lines represent the average estimated standard
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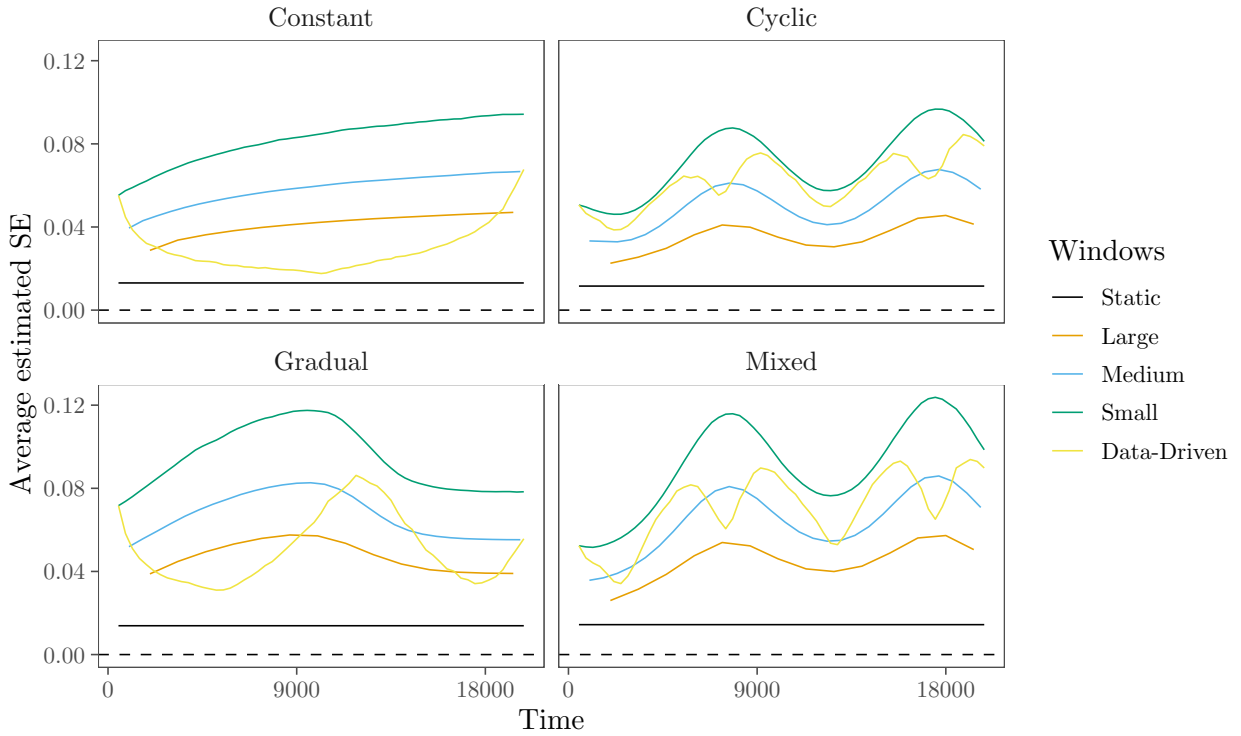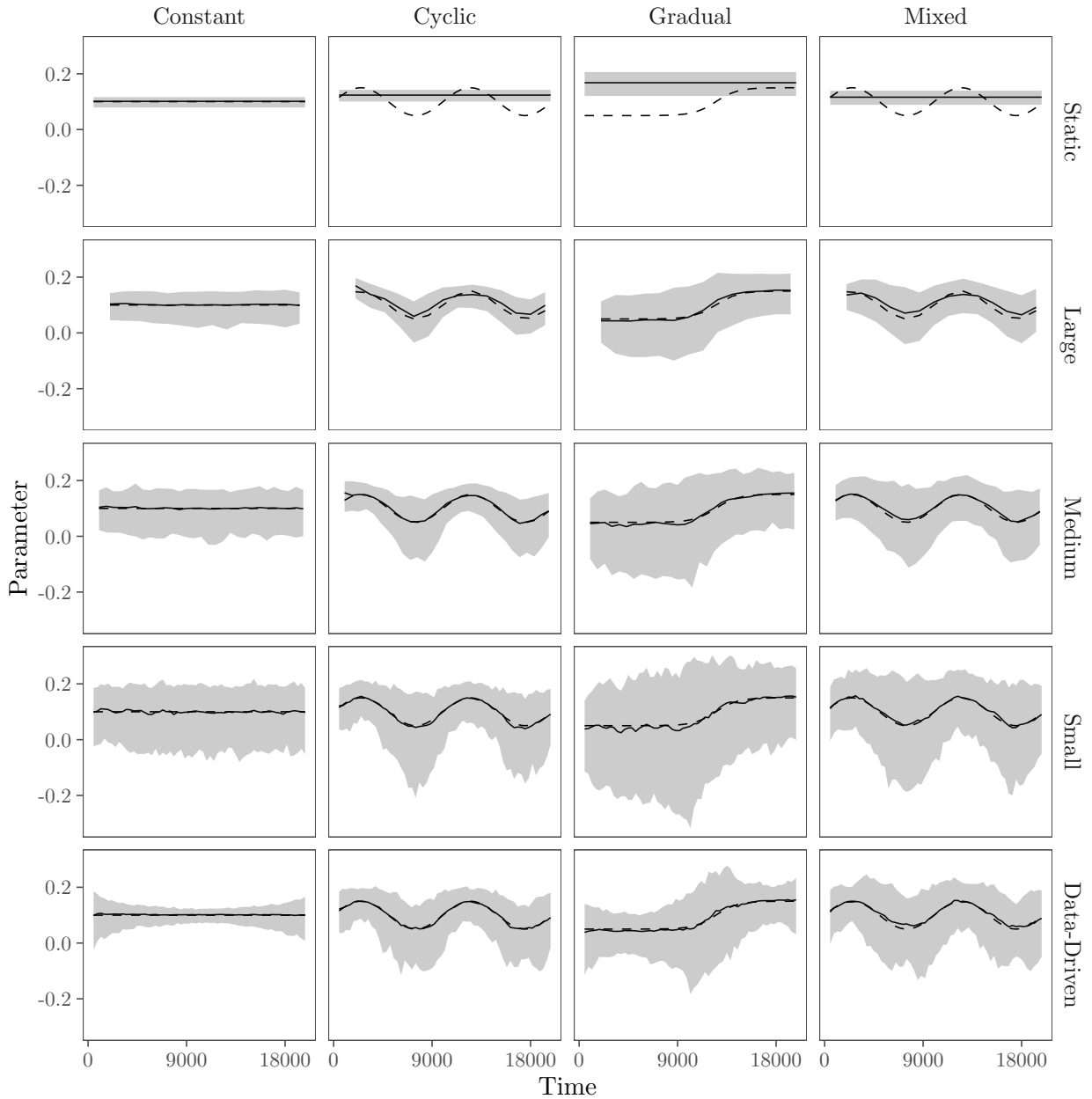(2000t), small (1000t) and data-based window widths.

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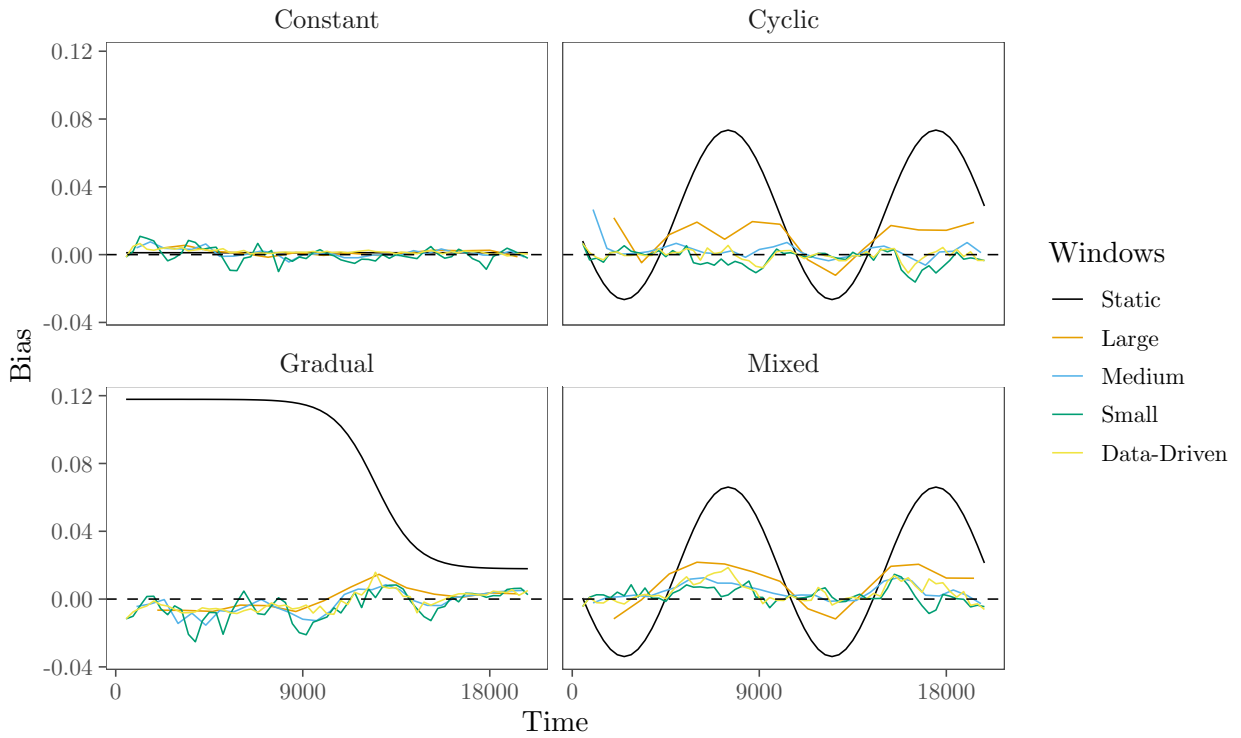
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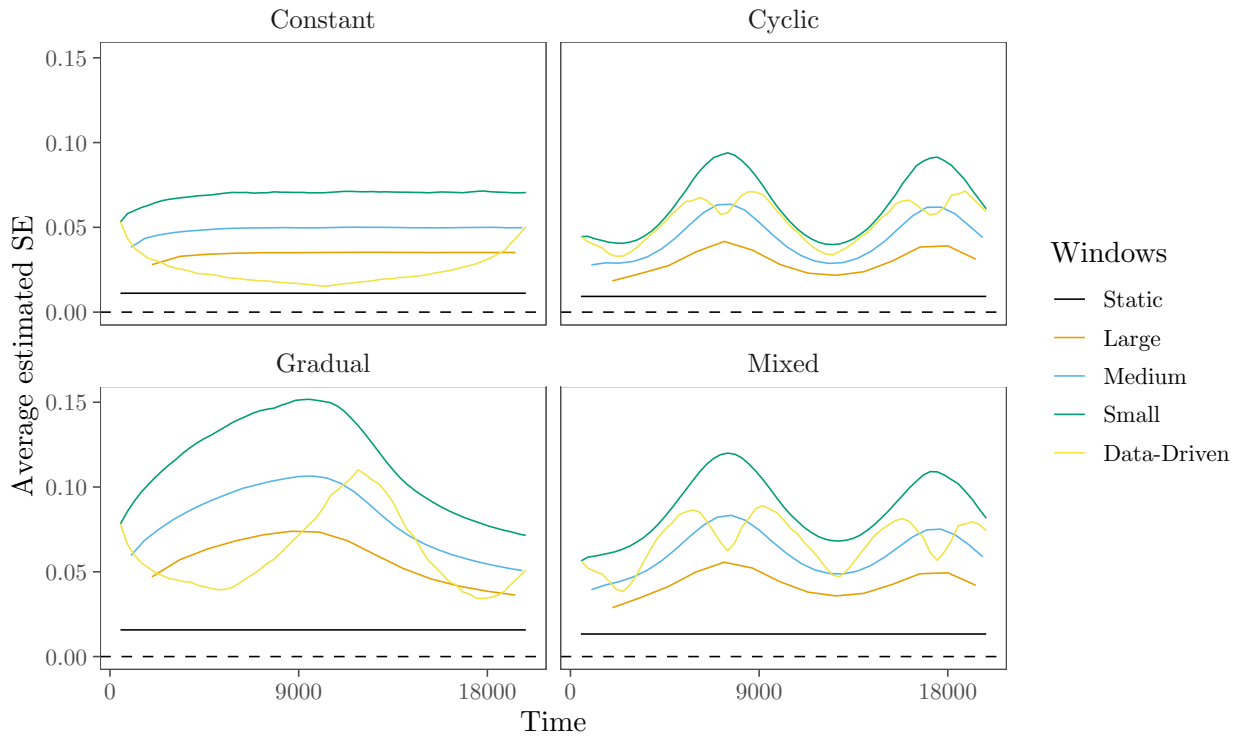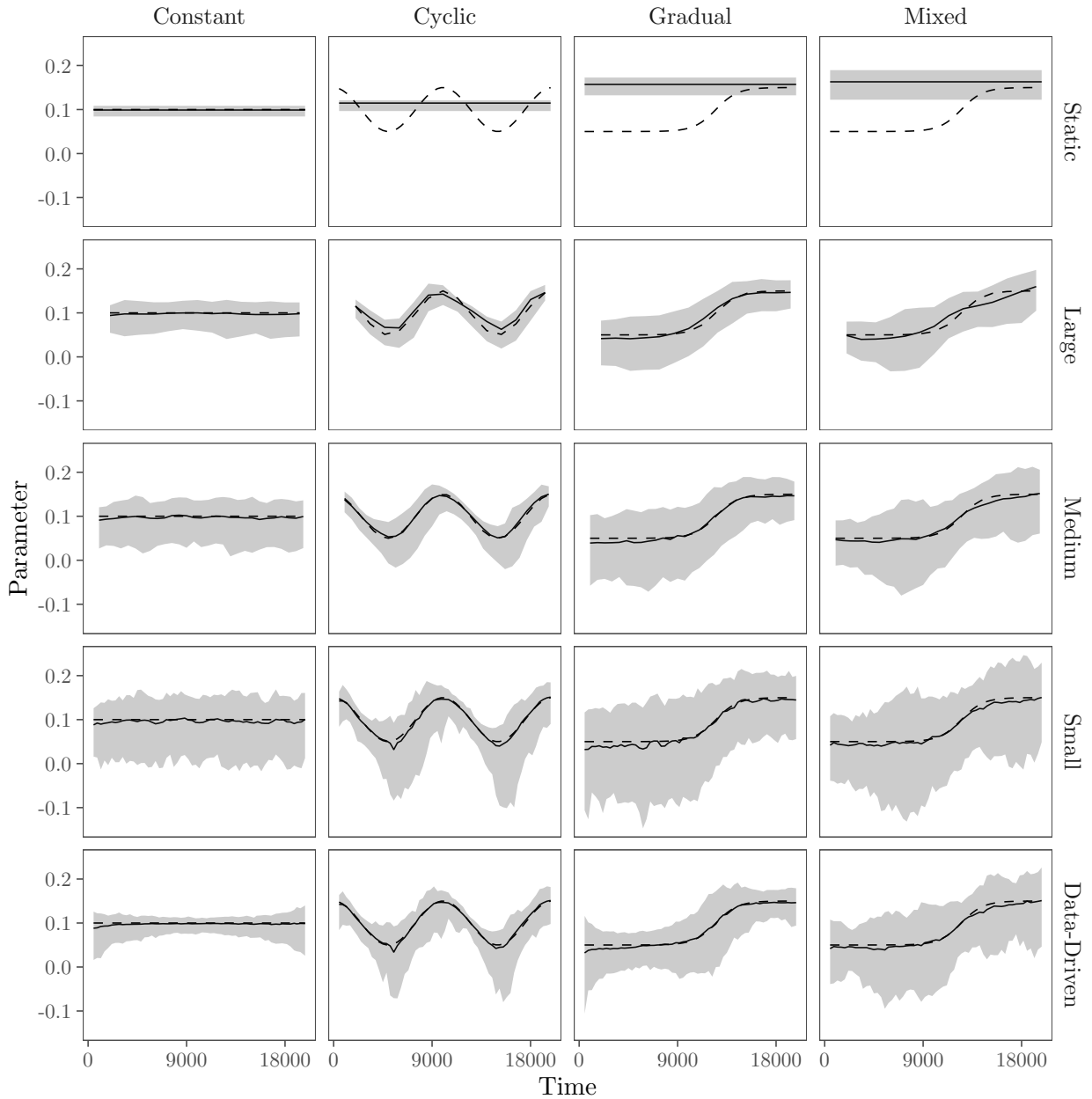
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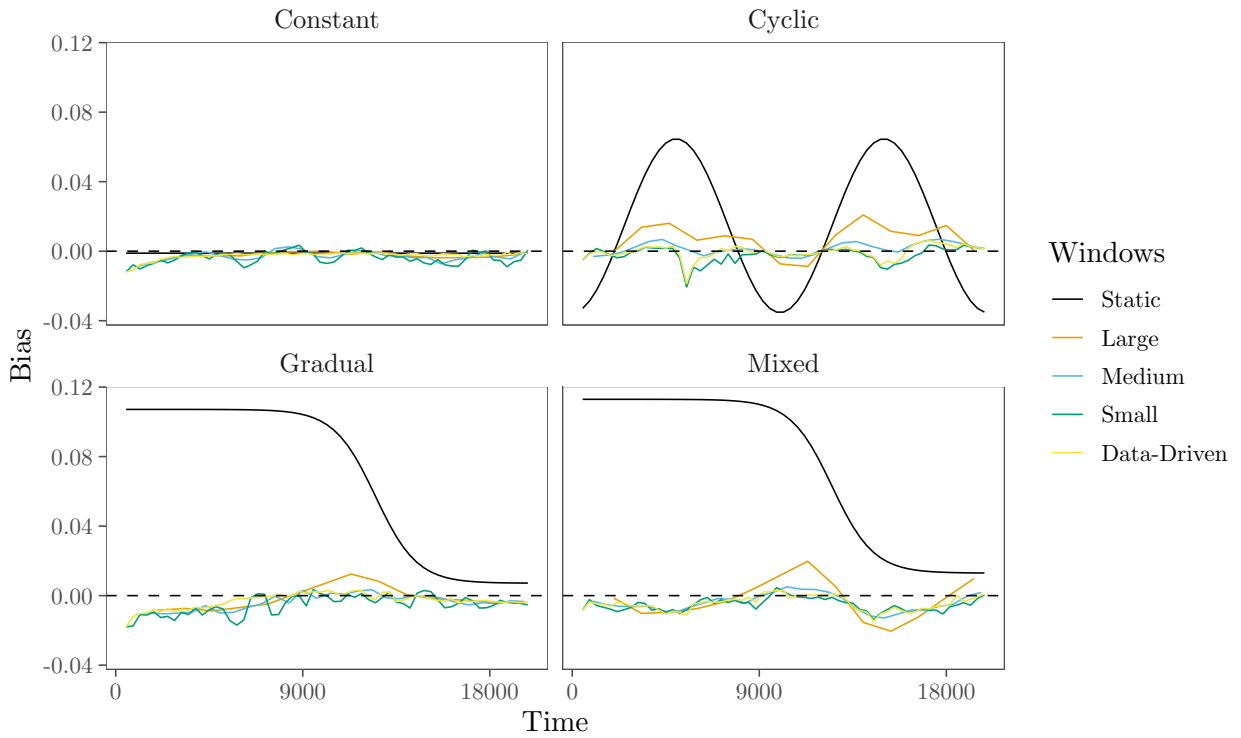
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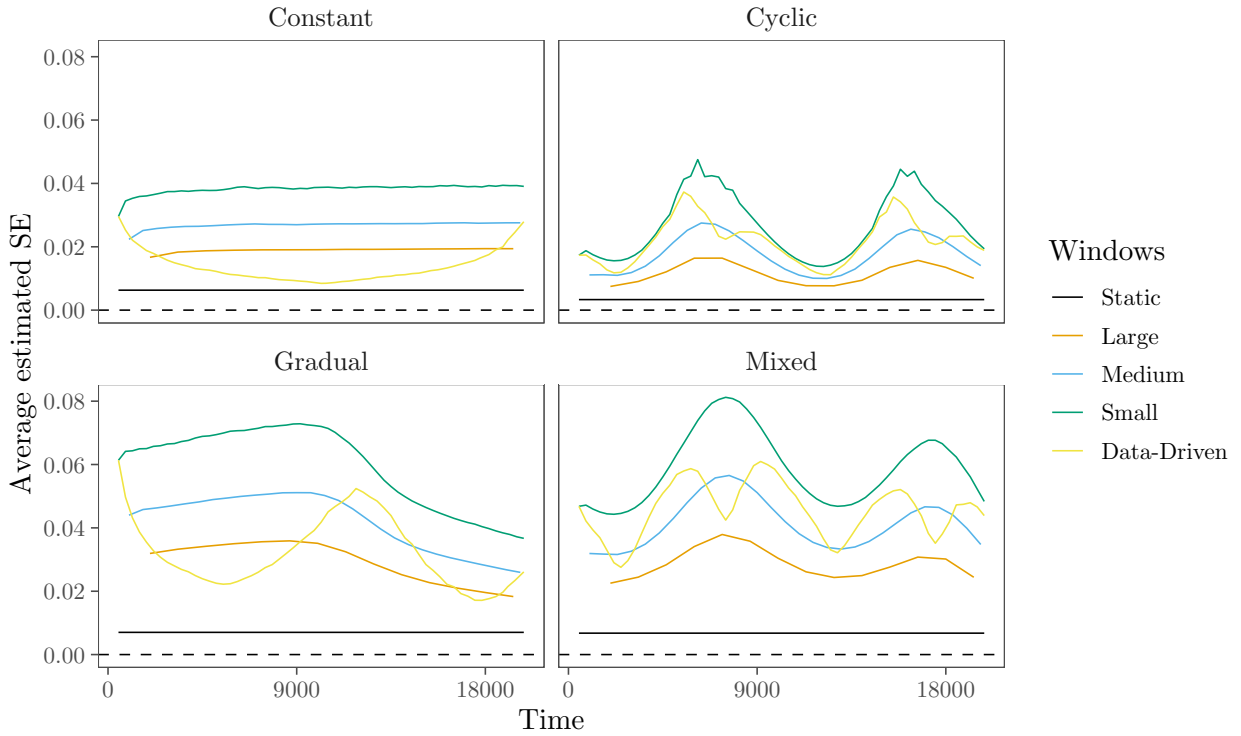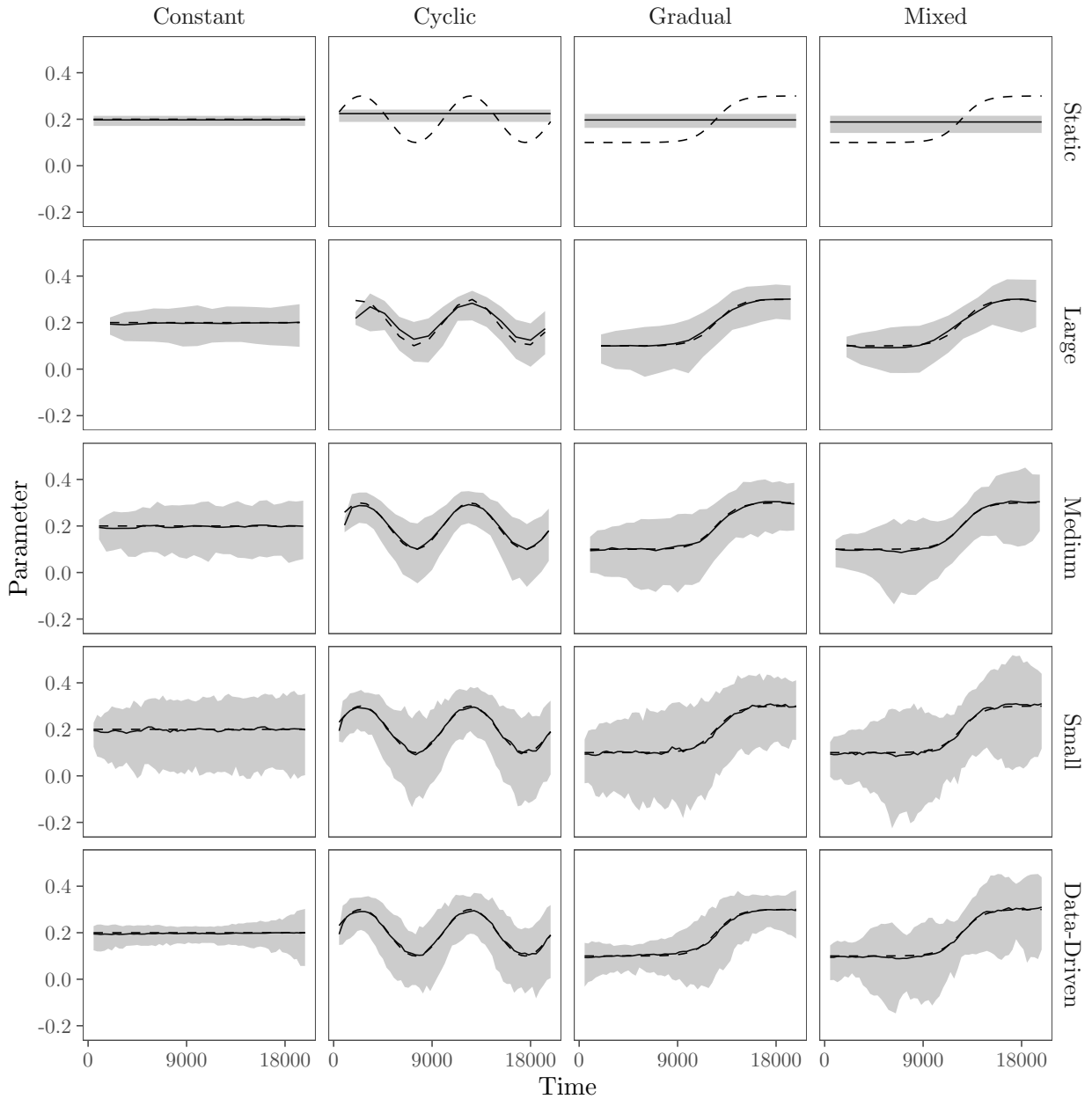
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the 'transitivity' effect in the four time-varying effects scenarios. Solid lines represent the mean estimated
parameters over 200 datasets over time. The gray area represents the range with 95% of the estimates for the
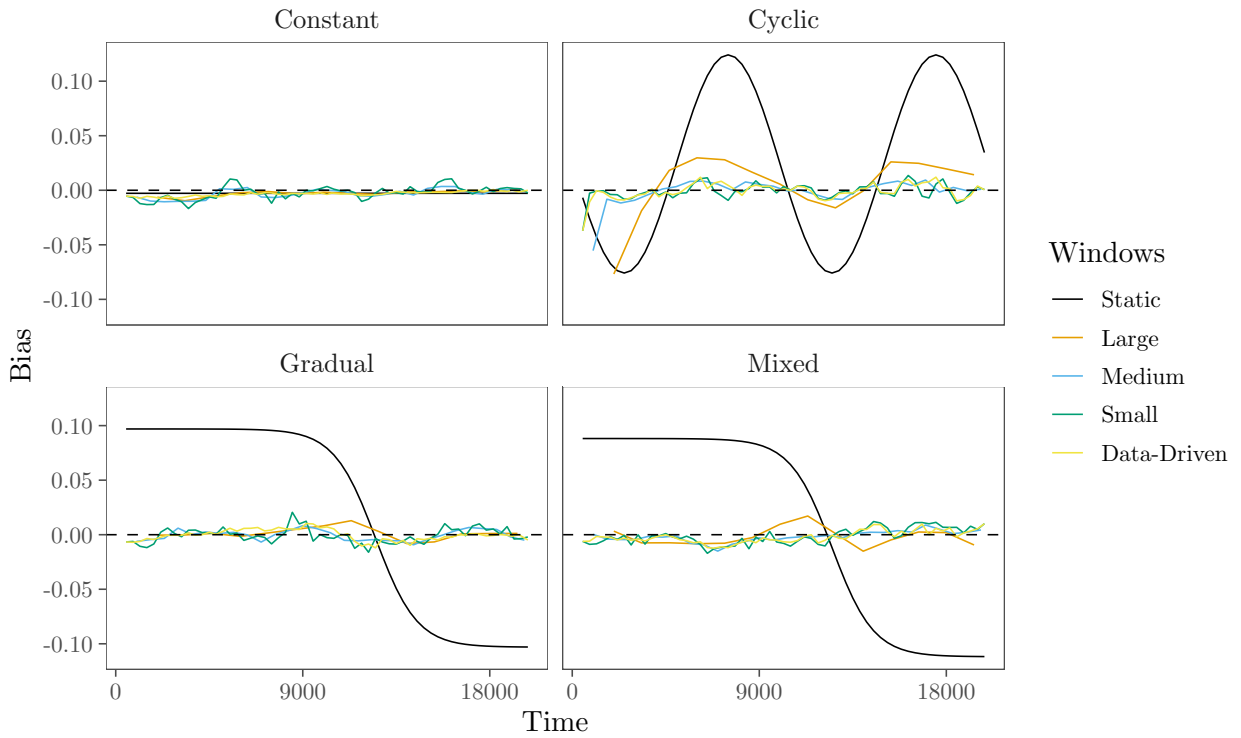200 datasets. Dashed lines represent the parameters used for data generation.

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Panels refer to the four time-varying effect scenarios. Solid lines represent the average estimated standard
error of the parameter estimates over time, with colors representing estimation with large ($4000t$), medium
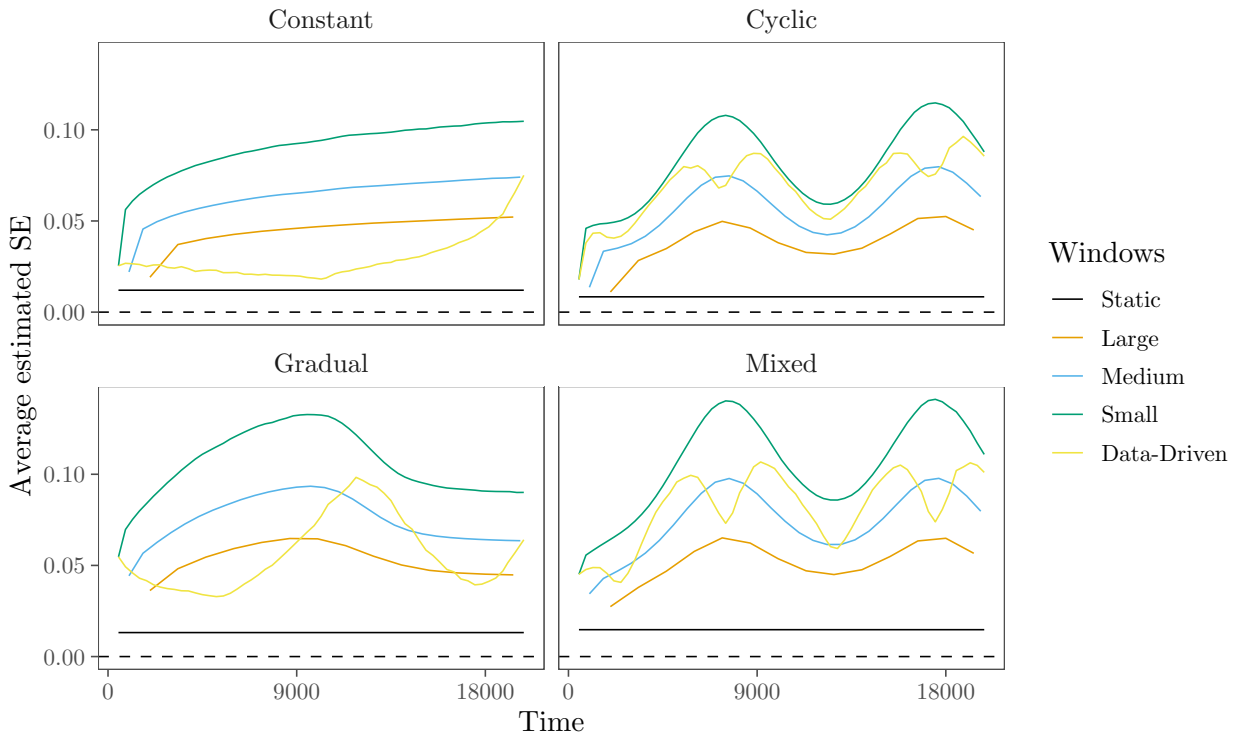($2000t$), small ($1000t$) and data-based window widths.