

Before-and-after analysis:

An application of structural break testing to the determination of economic damages.

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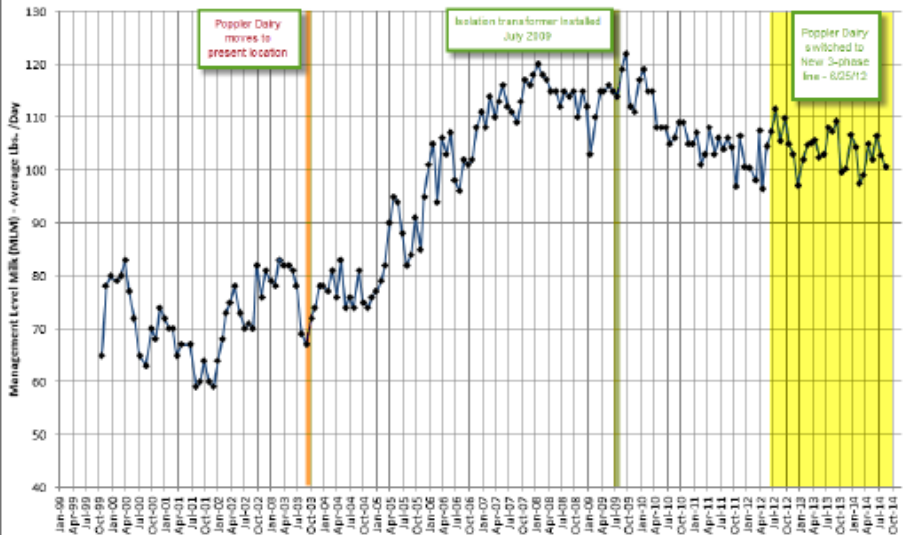
2018 ASSA Annual Meeting



Outline

- 1 Introduction
- 2 The problems
 - Overt omitted variable problem
 - Multiple comparisons problem
 - Latent omitted variable problem
- 3 Empirical example
- 4 Objections
 - Post hoc ergo propter hoc
 - Nearly simultaneous causal events
 - Causal lag
- 5 Conclusion

Poppler - Management Level Milk (MLM) - DHI



Motivation

- Before-and-after method is highly reliable and widely accepted, BUT it is not fool proof
- Tomlin and Merrell (2006) demonstrate, “simple” forms of the before-and-after method may yield “phantom” damages
- The simplistic approach, however, may run into three problems that we label:
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 - **more data to estimate**
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Nevada Bob's Golf of Nebraska, Inc. counterclaimed that the plaintiff denied its right of first refusal to lease and occupy adjacent space

- Nevada Bob's expert estimated lost profit by simply calculating profit per square foot at the alternative location
- The court concluded,
 - *"The expert assumed that the only difference between the two locations was the square footage"*
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- Especial relevant in providing support for causation or damage period
- Testing for a single break assumes only one external cause

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In this case, Clear Channel Communications, Inc. et al. was alleged to have “engaged in anticompetitive...practices in an effort to...extend its monopoly power in the market for live rock concerts tickets

- The plaintiff's expert predict the expected average ticket prices from 2000 through 2006 using ticket prices from
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In rebuttal the plaintiff's expert

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- *“Moreover, the ‘pooled sample’ analysis fails even to consider whether a so-called ‘structural break’ occurred in any year other than 2000...”*
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Solution to the latent omitted variable problem

- To account for an unobserved factors we must allow for multiple unknown breakpoints
- Bai (1997) and Bai and Perron (1998) further extend the Quandt-Andrews test to allow for multiple unknown breakpoints
- Perron (2006) recommends
 - Double maximum test to determine if any breaks are present
 - Sequential test starting at some value of greater than zero to determine the number of breakpoints

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Dairy Litigation

In this case, the dairy claimed their business was harmed by stray voltage, and sued electrical utility. The court found in favor the dairy ordered repairs and awarded damages.

- A partial remedy was implemented July 2009
- The definitive remedy was implemented June 2012

Modeling the Damage Period

- Dependent Variable

MHA Annualized average monthly production per cow

- Breaking Regressors

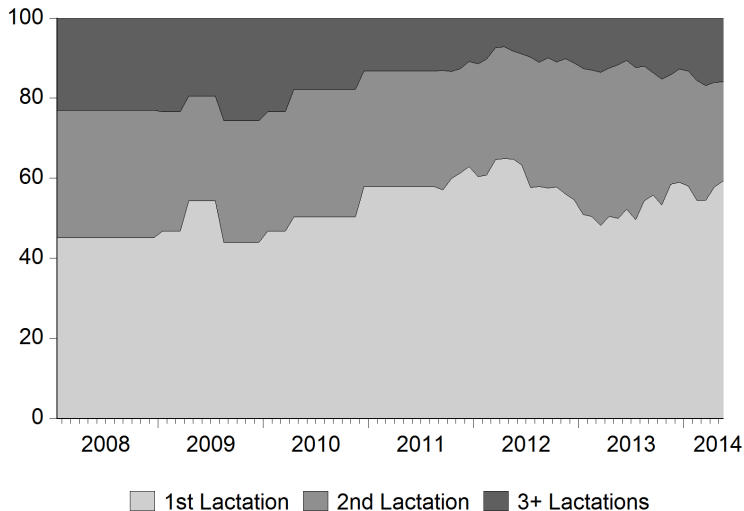
Constant

Trend Deterministic time trend

- Non-breaking Regressors

L1 Percent of heard that has had only on calf

Percent of herd in 1st, 2nd or 3rd+ lactation



Modeling the Damage Period

- Perform the double maximum structural break testing procedure as suggested by Perron (2006) with
 - trimming is set to 20%
 - maximum number of breaks set to 3
 - WDMax test statistic = 24.344 (C.V. = 12.15)
- Number of Breaks
 - Significant F-statistic largest breaks: 3
 - UDmax = 19.97 (C.V. = 11.16) determined breaks: 2
 - WDmax = 24.344 (C.V. = 12.15) determined breaks: 2
- Estimated break dates:
 - 1: 2013M02
 - 2: 2009M04, 2013M02
 - 3: 2009M10, 2011M12, 2013M03

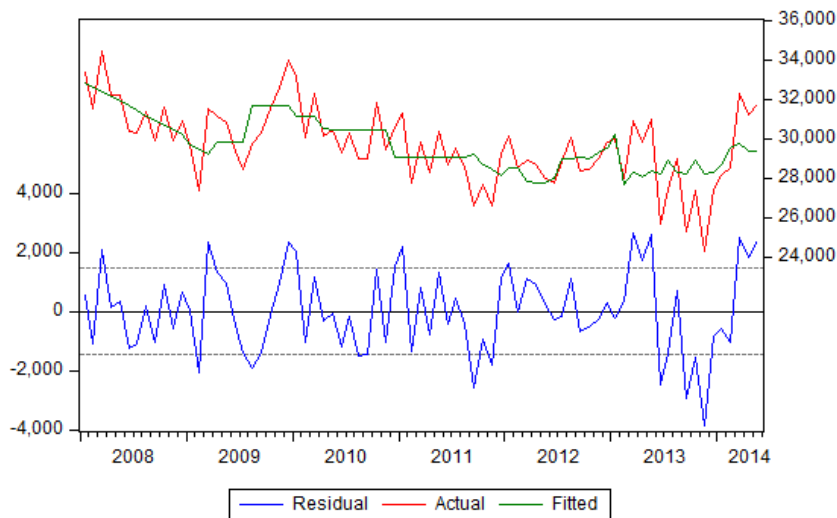
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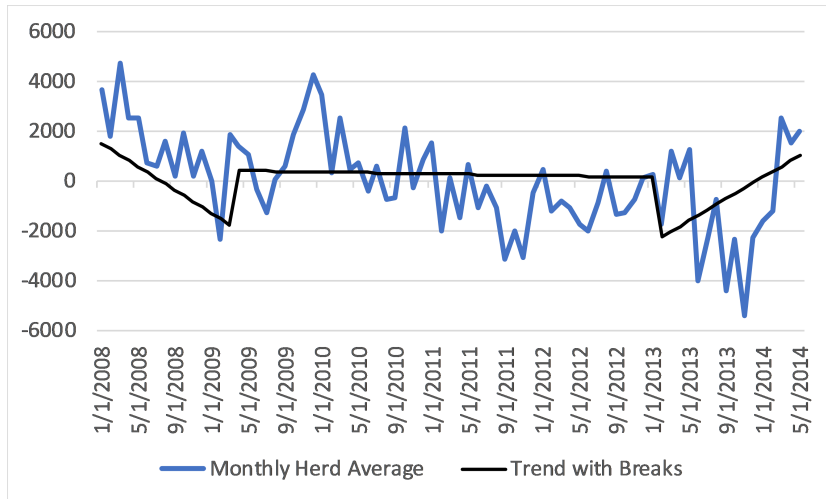
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Trend with Breaks Full Model Fit



Trend with Breaks



Three objections to structural break analysis

- ① Post hoc ergo propter hoc fallacy
- ② Nearly simultaneous causal events
- ③ Causal lag

Post hoc ergo propter hoc

- Objection
 - Structural break analysis relies on timing evidence
 - This line of reasoning may well suffer either the post hoc ergo propter hoc or the cum hoc ergo propter hoc fallacies.
- Answer
 - All before-and-after analysis relies on timing evidence (see e.g. *Young v. Hickory Business Furniture* 2000).
 - Structural break analysis allows for control of
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- Objection
 - An issue is whether more than one casual factor could have occurred within the same time period
 - This challenge is most apparent when working with annual data
- Answers
 - Non-statistical supporting evidence/testimony
 - Higher frequency data

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- However, as Gaughan states,
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