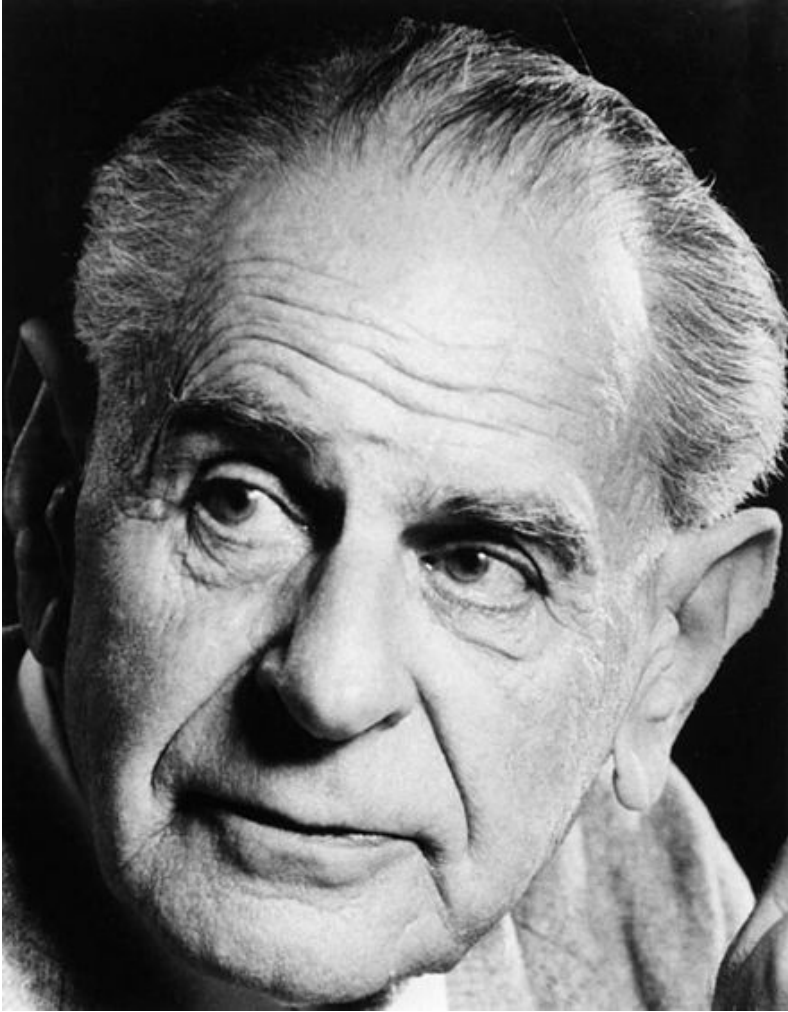


# Hypothesis Testing is a Bad Idea

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Columbia University, New York

Contemporary Issues in Hypothesis Testing workshop, Warwick,  
15 Sept 2016



## The logic of scientific discovery

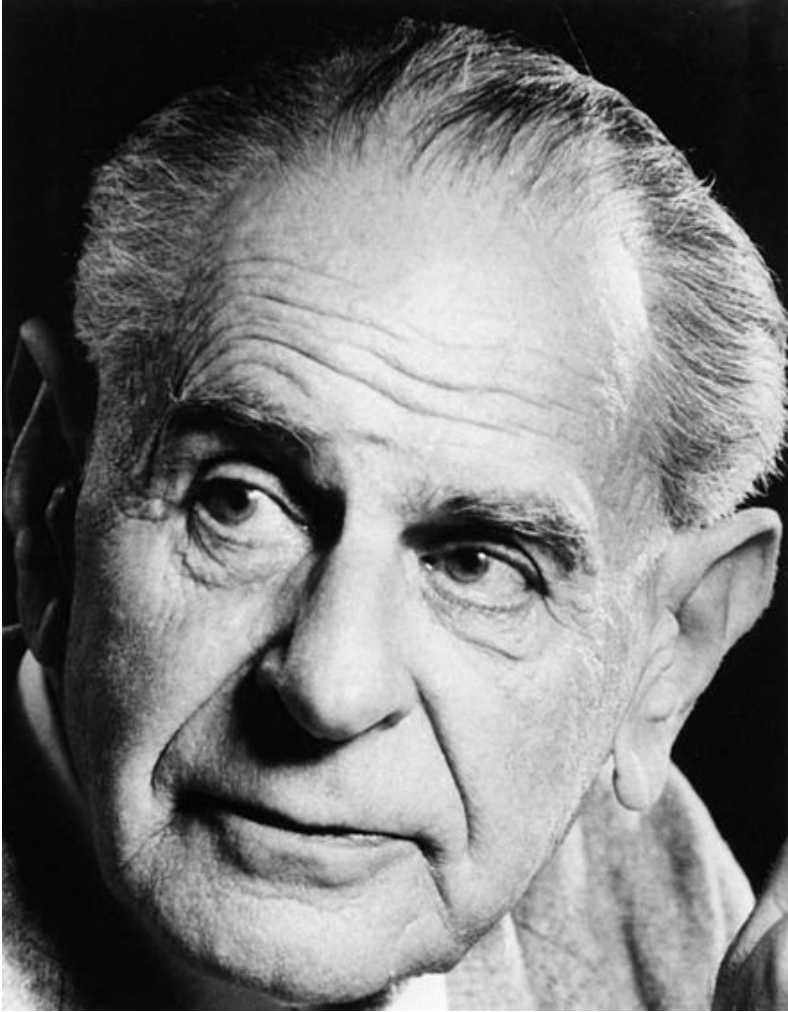
“The application of the chosen test left little doubt that the lymphocytes from household contact persons are, by and large, better breast cancer cell killers than those from the controls.” — J. Neyman, 1976

# Confirmationist vs. falsificationist reasoning

## The null hypothesis



Researchers are happy to reject this guy.

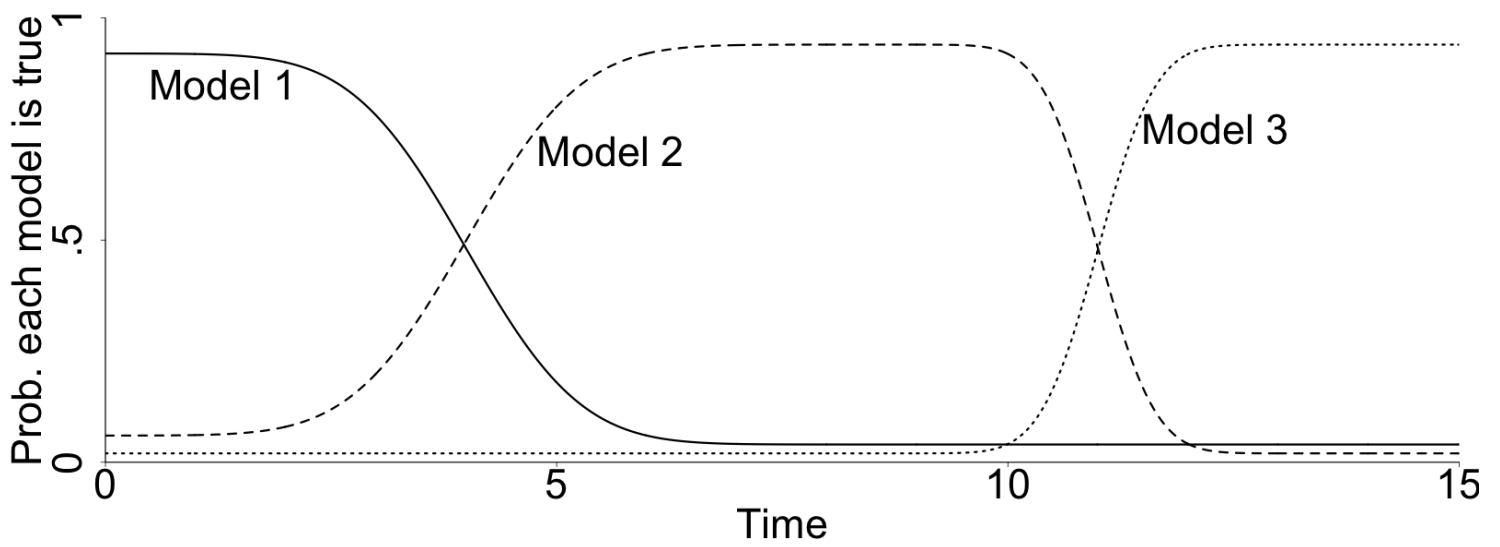


## The logic of scientific discovery



## The Structure of Scientific Revolutions

Here's a story I don't believe:





The “I’ve got statistical significance and I’m outta here” attitude

## I hate discrete models!

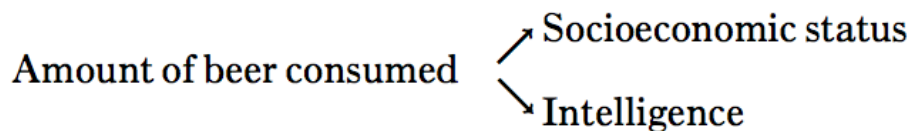
From a cognitive scientist:

If two of the variables are dependent, say, intelligence and socioeconomic status, but conditionally independent given the third variable [beer consumption], then either they are related by one of two chains:

(Intelligence  $\rightarrow$  Amount of beer consumed  $\rightarrow$  Socioeconomic status)

(Socioeconomic status  $\rightarrow$  Amount of beer consumed  $\rightarrow$  Intelligence)

or by a fork:



and then we must use some other means [other than observational data] to decide between these three possibilities. In some cases, common sense may be sufficient, but we can also, if necessary, run an experiment. If we intervene and vary the amount of beer consumed and see that we affect intelligence, that implies that the second or third model is possible; the first one is not. Or course, all this assumes that there aren't other variables mediating between the ones shown that provide alternative explanations of the depen-





What makes  
me more upset  
than binary  
thinking?

P  
X  
©  
R  
sa  
D  
P  
©

# The Fluctuating Female Vote: Politics, Religion, and the Ovulatory Cycle

**Kristina M. Durante<sup>1</sup>, Ashley Rae<sup>1</sup>, and Vldas Griskevicius<sup>2</sup>**

<sup>1</sup>College of Business, University of Texas, San Antonio, and <sup>2</sup>Carlson School of Management, University of Minnesota

## Abstract

Each month, many women experience an ovulatory cycle that regulates fertility. Although the ovulatory cycle influences women's mating preferences, we proposed that it might also change women's political views. Building on theory suggesting that political and religious orientation are linked to reproductive preferences, we tested how fertility influenced women's politics, religiosity, and voting in the 2012 U.S. presidential election. With large and diverse samples, ovulation had drastically different effects on single women and women in relationships. Ovulation led single women to become more liberal, less religious, and more likely to vote for Barack Obama. In contrast, ovulation led women in committed relationships to become more conservative and more likely to vote for Mitt Romney. In addition, ovulation-induced changes in political orientation mediated women's voting behavior. Overall, the ovulatory cycle not only influences women's politics differently for single women than for women in relationships.

## The Ancestral Logic of Politics

### Upper-Body Strength Regulates Men's Assertion of Self-Interest Over Economic Redistribution

---

Over human evolutionary history, upper-body strength has been a major component of fighting ability. Evolutionary models of animal conflict predict that actors with greater fighting ability will more actively attempt to acquire or defend resources than less formidable contestants will. Here, we applied these models to political decision making about redistribution of income and wealth among modern humans. In studies conducted in Argentina, Denmark, and the United States, men with greater upper-

Unfalsifiable!

But ...  
falsification of statistical models is not our goal.



1 quick tip to falsify any statistical model . . .

# Birthdays!

Social Science & Medicine 73 (2011) 1246–1248



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Contents lists available at [SciVerse ScienceDirect](#)

## Social Science & Medicine

journal homepage: [www.elsevier.com/locate/socscimed](http://www.elsevier.com/locate/socscimed)



### Short report

## Influence of Valentine's Day and Halloween on Birth Timing

Becca R. Levy\*, Pil H. Chung, Martin D. Slade

Yale University, School of Public Health, Division of Social & Behavioral Sciences, 60 College Street, New Haven, CT 06520-8034, United States

### ARTICLE INFO

*Article history:*  
Available online 28 July 2011

*Keywords:*  
United States  
Culture  
Birth timing  
Holidays  
Pregnancy  
Biocultural  
Birth

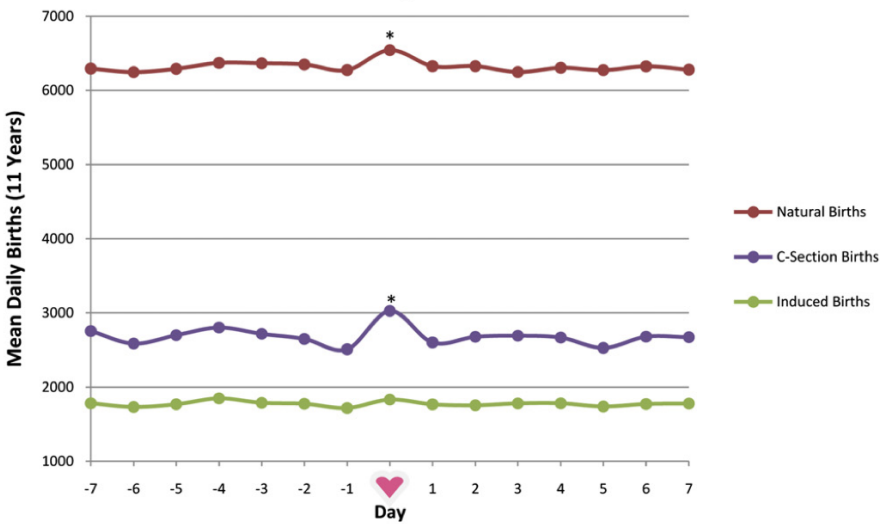
### ABSTRACT

It is known that cultural representations, in the form of stereotypes, can influence functional health. We predicted that the influence of cultural representations, in the form of salient holidays, would extend to birth timing. On Valentine's Day, which conveys positive symbolism, there was a 3.6% increase in spontaneous births and a 12.1% increase in cesarean births. Whereas, on Halloween, which conveys negative symbolism, there was a 5.3% decrease in spontaneous births and a 16.9% decrease in cesarean births. These effects reached significance at  $p < .0001$ , after adjusting for year and day of the week. The sample was based on birth-certificate information for all births in the United States within one week on either side of each holiday across 11 years. The Valentine's-Day window included 1,676,217 births and the Halloween window included 1,809,304 births. Our findings raise the possibility that pregnant women may be able to control the timing of spontaneous births, in contrast to the traditional assumption, and that scheduled births are also influenced by the cultural representations of the two holidays.

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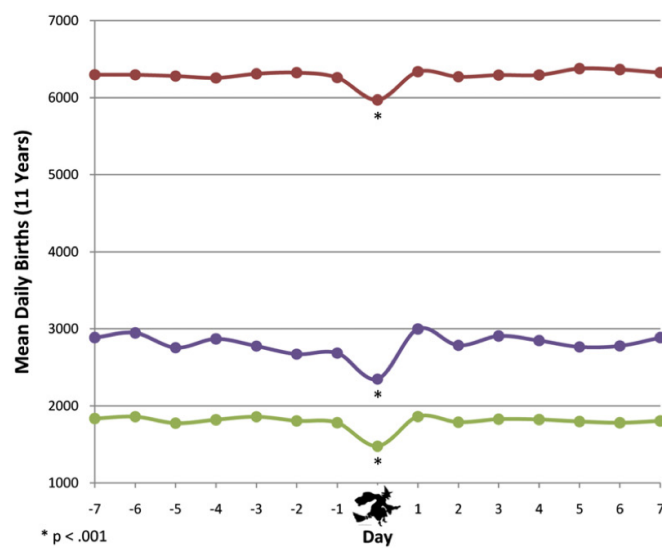
The published graphs show data from 30 days in the year

Valentine's Day: Two-Week Window



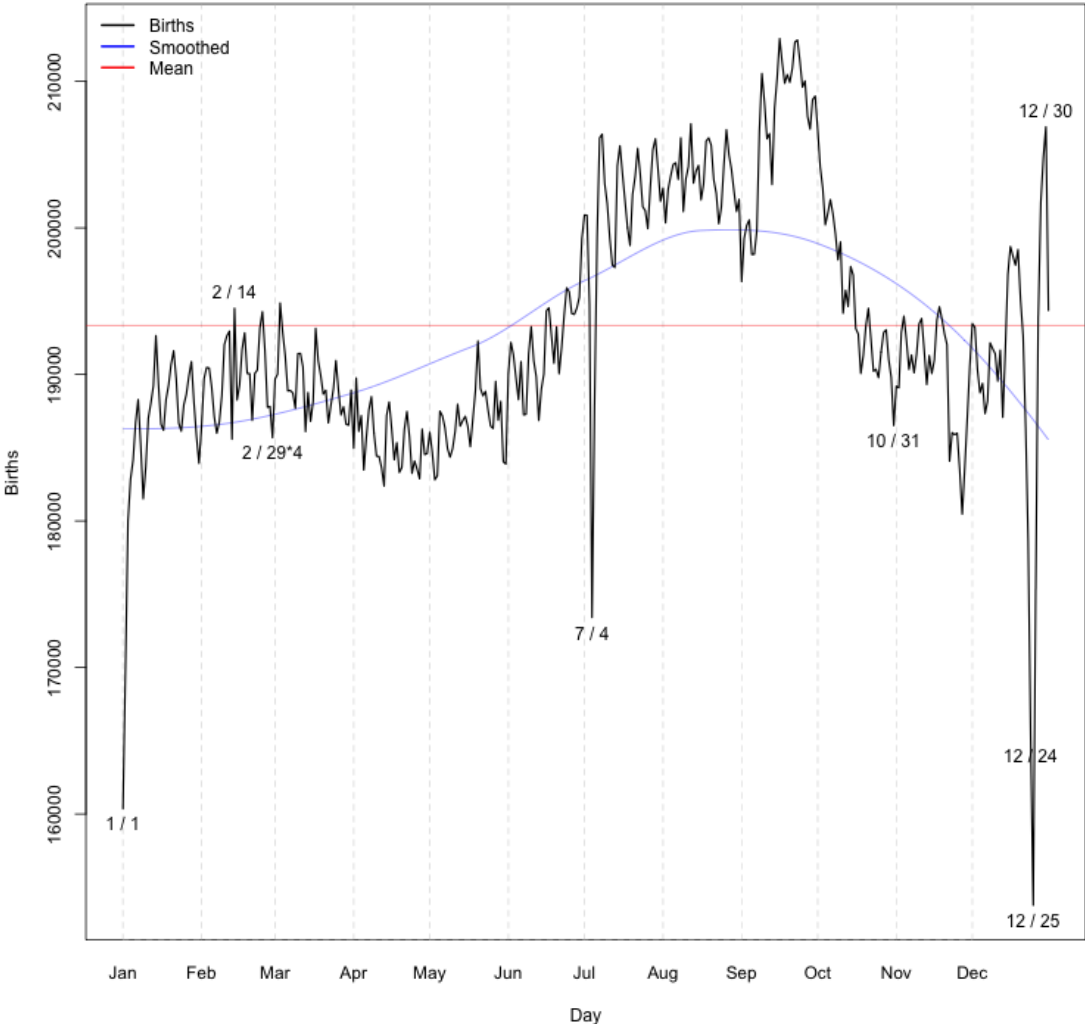
\* p < .001

Halloween: Two-Week Window



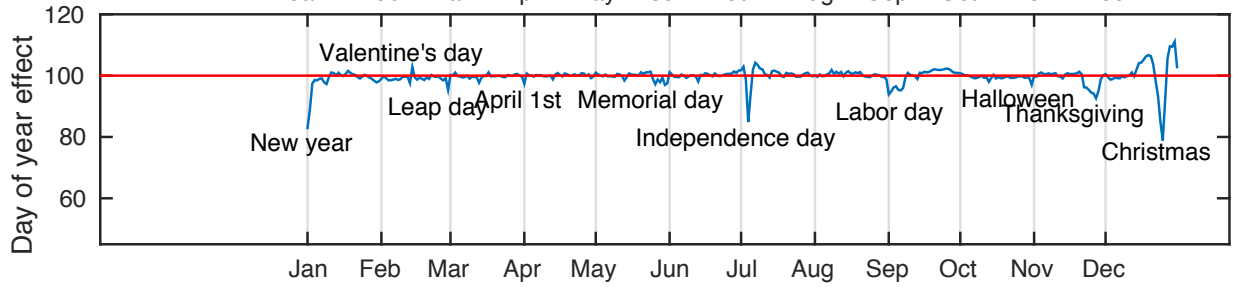
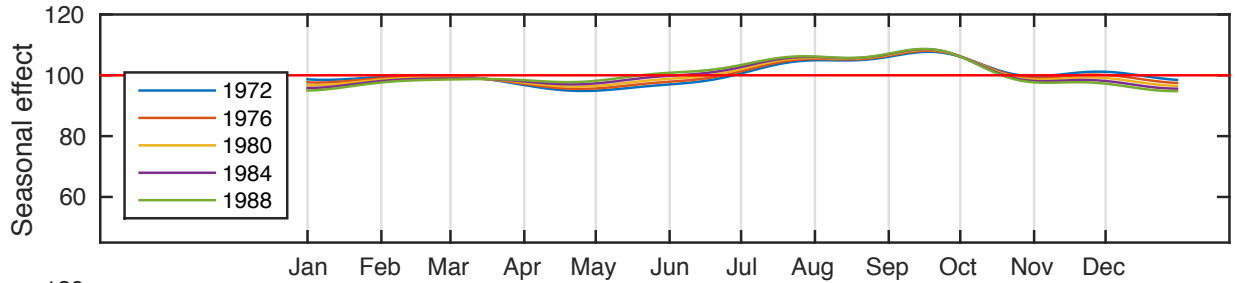
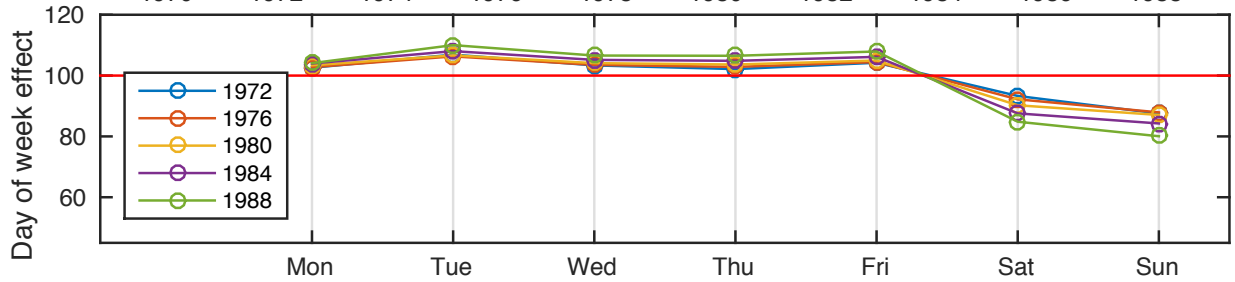
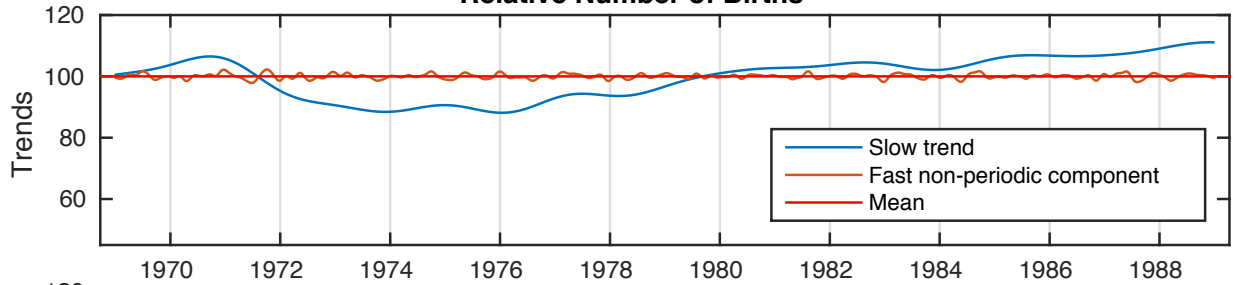
\* p < .001

### Births by Day of Year

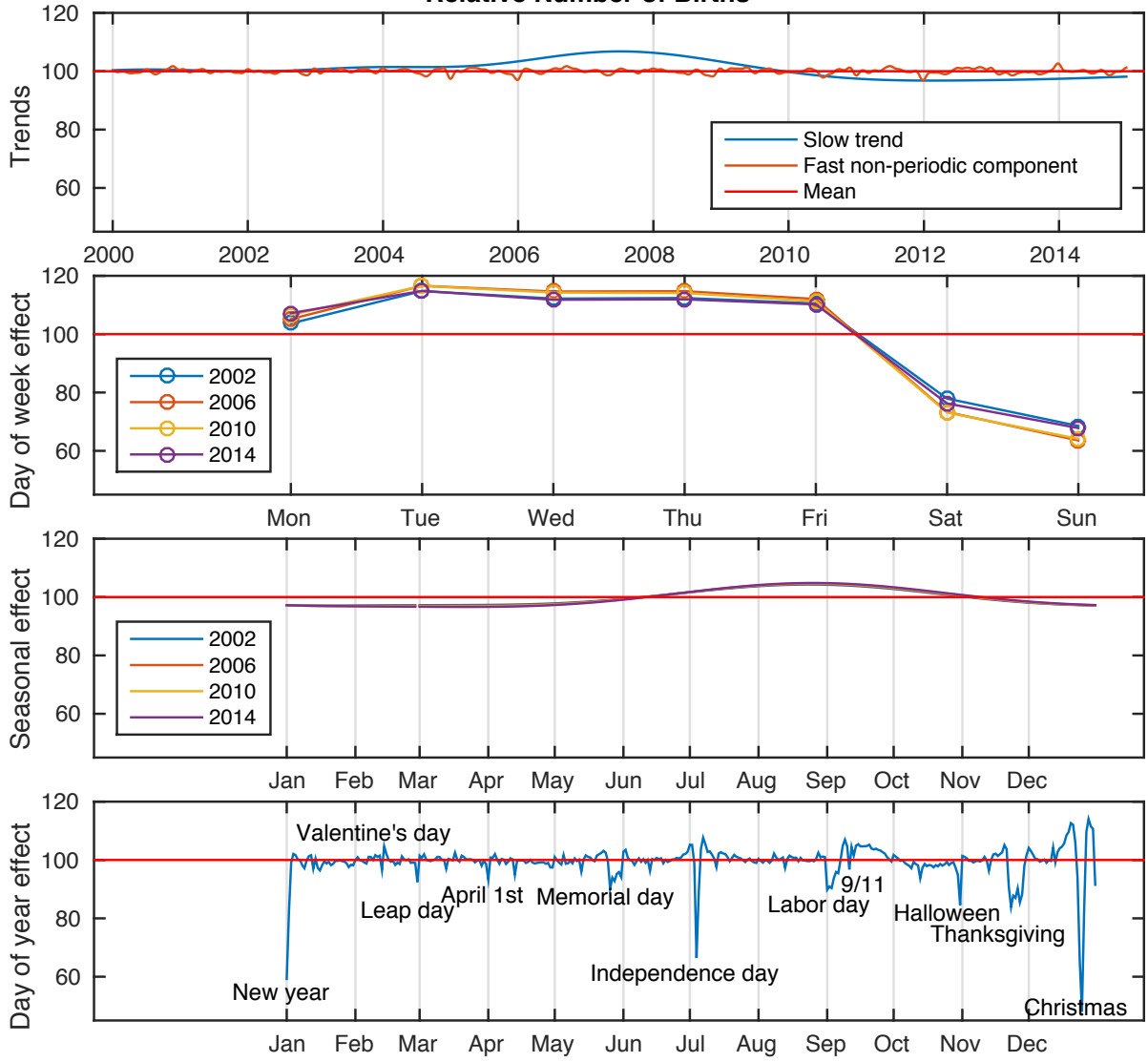


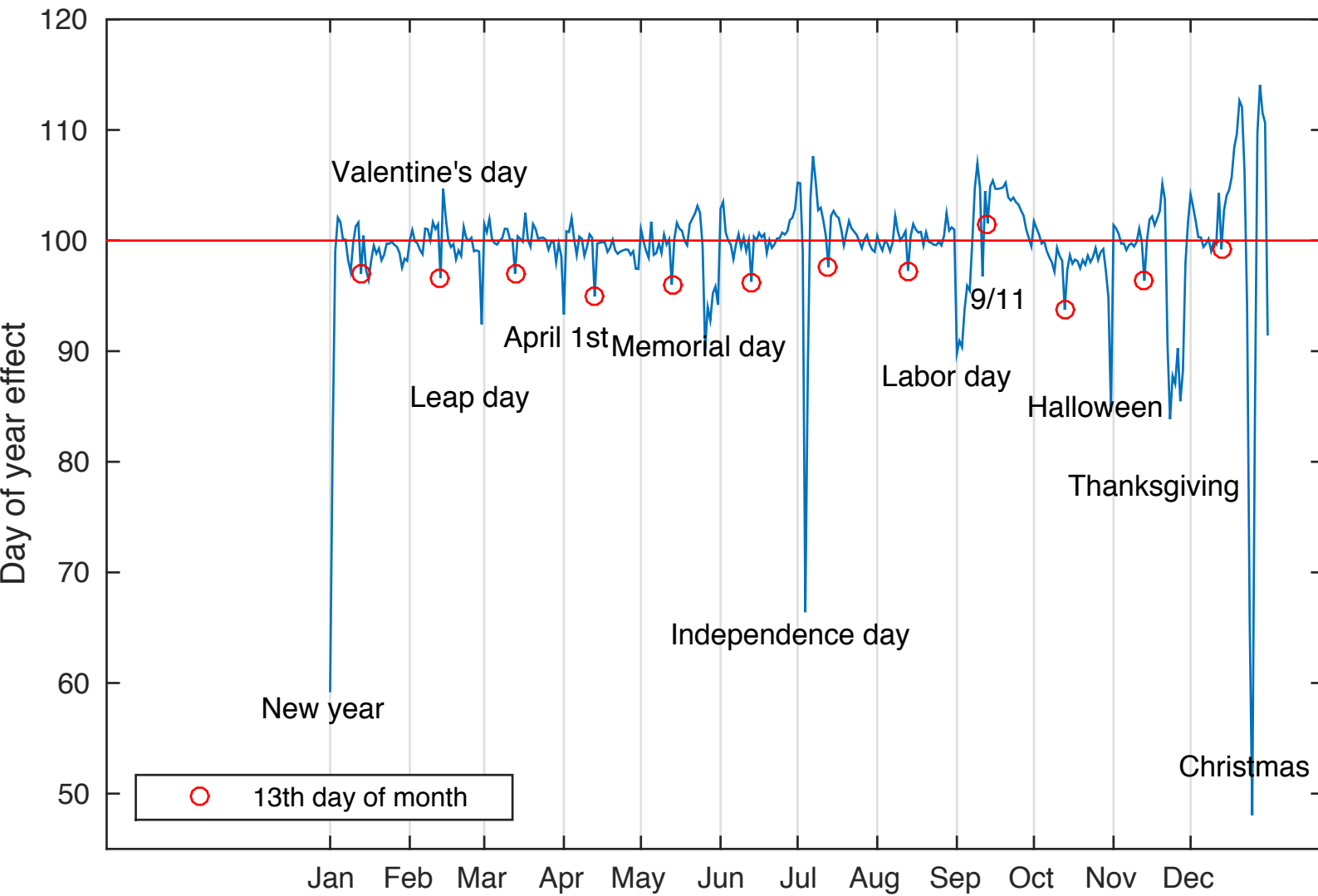
Source: National Vital Statistics System natality data, as provided by Google BigQuery. Graph by Chris Mulligan (chmullig.com)

### Relative Number of Births



### Relative Number of Births





○ 13th day of month



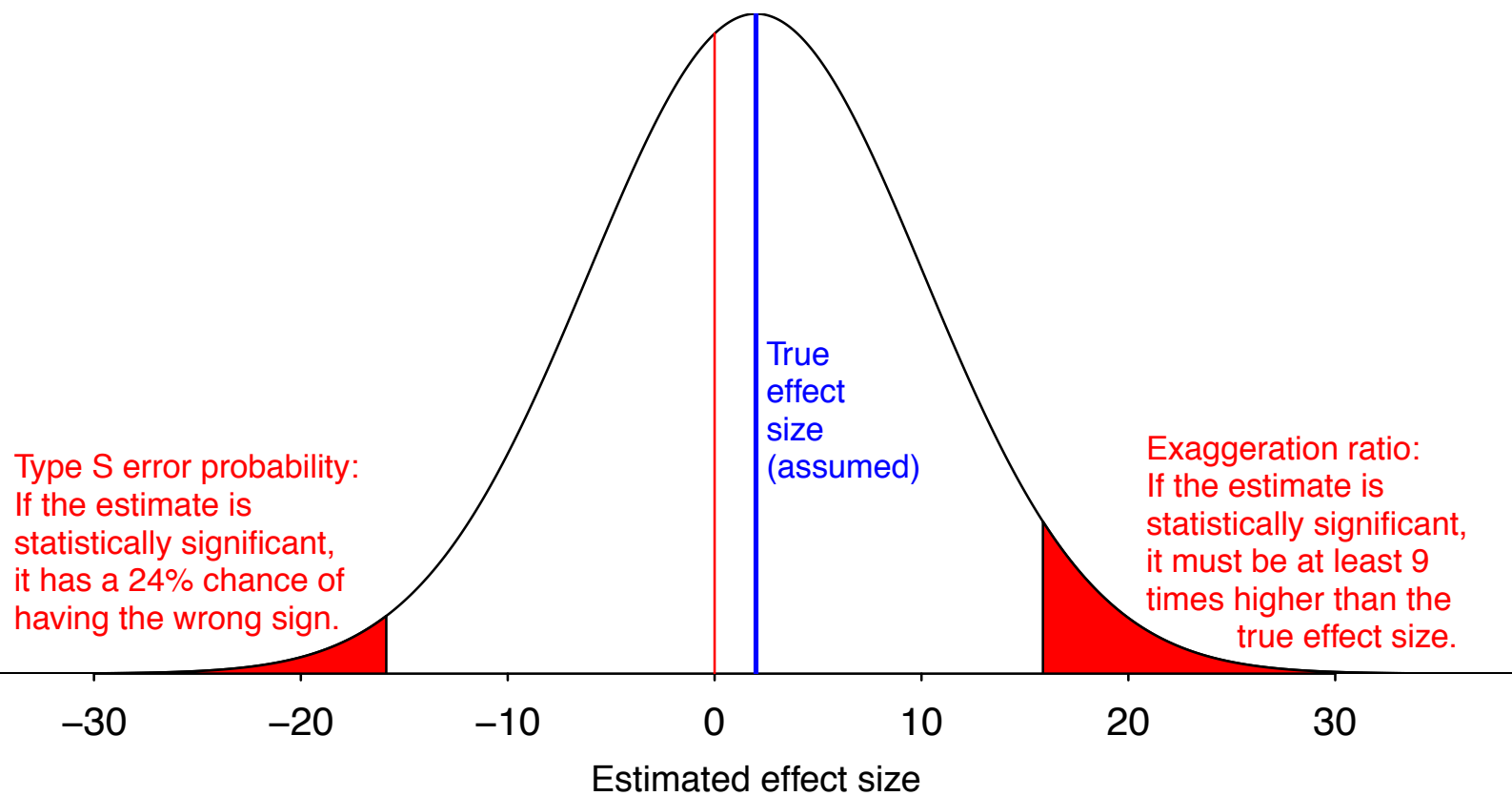
# Methodology of Scientific Research Programmes



How can things go wrong?

The statistical significance filter and the resulting feedback loop

**This is what "power = 0.06" looks like.  
Get used to it.**



Policy!

## Labor Market Returns to Early Childhood Stimulation: a 20-year Followup to an Experimental Intervention in Jamaica

Paul Gertler, James Heckman, Rodrigo Pinto, Arianna Zanolini, Christel Vermeersch, Susan Walker, Susan M. Chang, Sally Grantham-McGregor

We find large effects on the earnings of participants from a randomized intervention that gave psychosocial stimulation to stunted Jamaican toddlers living in poverty. The intervention consisted of one-hour weekly visits from community Jamaican health workers over a 2-year period that taught parenting skills and encouraged mothers to interact and play with their children in ways that would develop their children's cognitive and personality skills. We re-interviewed the study participants 20 years after the intervention. Stimulation increased the average earnings of participants by 42 percent. Treatment group earnings caught up to the earnings of a matched non-stunted comparison group. These findings show that psychosocial stimulation early in childhood in disadvantaged settings can have substantial effects on labor market outcomes and reduce later life inequality.

# Bad Bayes



# Bad Bayes

► Model:



## Bad Bayes

- ▶ Model:
  - ▶  $y|\theta \sim N(\theta, 1)$



## Bad Bayes

- ▶ Model:

- ▶  $y|\theta \sim N(\theta, 1)$
- ▶  $p(\theta) \propto 1$











## Bad Bayes

- ▶ Model:
  - ▶  $y|\theta \sim N(\theta, 1)$
  - ▶  $p(\theta) \propto 1$
- ▶ Data:
  - ▶  $y = 1$
- ▶ Inference:
  - ▶  $\theta|y \sim N(y, 1)$



## Bad Bayes

- ▶ Model:
  - ▶  $y|\theta \sim N(\theta, 1)$
  - ▶  $p(\theta) \propto 1$
- ▶ Data:
  - ▶  $y = 1$
- ▶ Inference:
  - ▶  $\theta|y \sim N(y, 1)$
  - ▶  $\Pr(\theta > 0|y) = .84$



## Bad Bayes

- ▶ Model:
  - ▶  $y|\theta \sim N(\theta, 1)$
  - ▶  $p(\theta) \propto 1$
- ▶ Data:
  - ▶  $y = 1$
- ▶ Inference:
  - ▶  $\theta|y \sim N(y, 1)$
  - ▶  $\Pr(\theta > 0|y) = .84$
- ▶ Wanna bet??



## Hypothesis testing is a bad idea:

Through a series of examples, we consider problems with classical hypothesis testing, whether performed using classical p-values or confidence intervals, Bayes factors, or Bayesian inference using noninformative priors. We locate the problem not in the use of any particular statistical method but rather with larger problems of deterministic thinking and a misguided version of Popperianism in which the rejection of a straw-man null hypothesis is taken as confirmation of a preferred alternative. We suggest solutions involving multilevel modeling and informative Bayesian inference.