

# Statistical philosophy and the language of significance

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U. Chicago, Oct 2018

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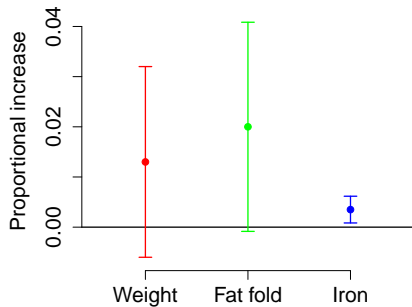
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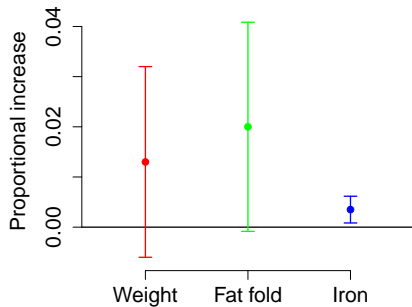
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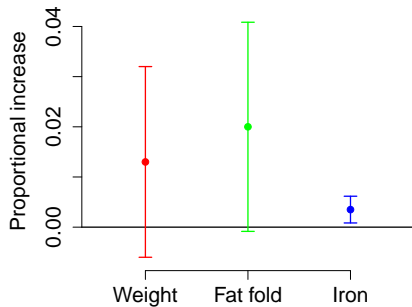
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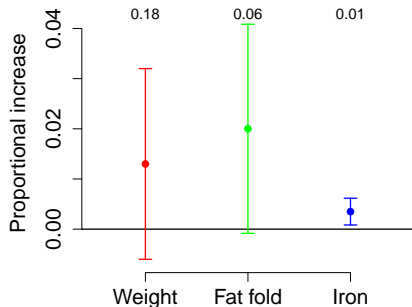
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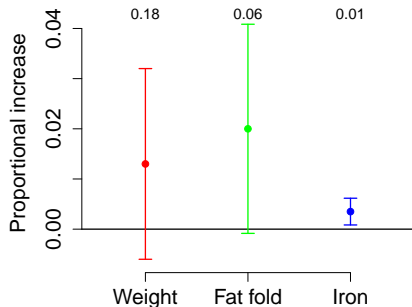
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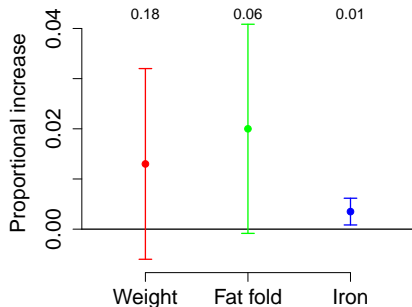
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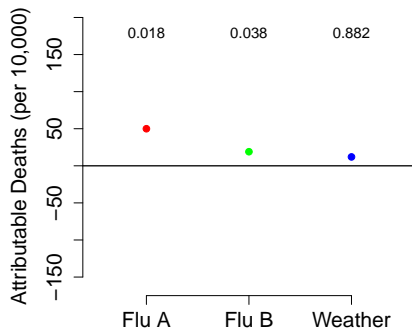
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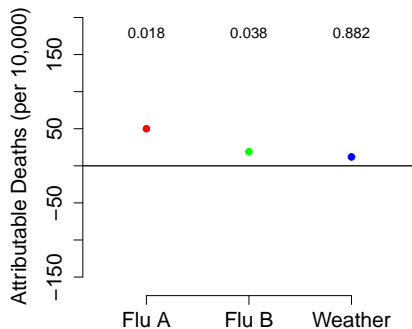


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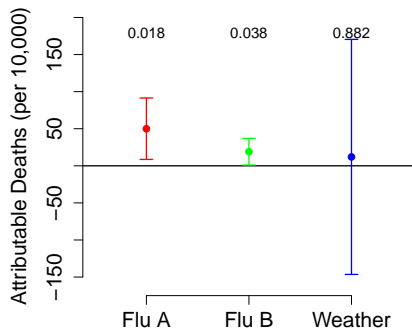
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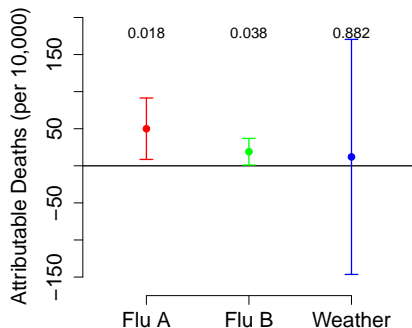
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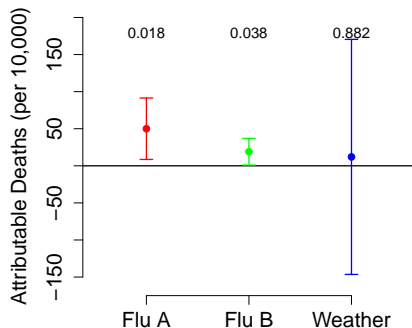
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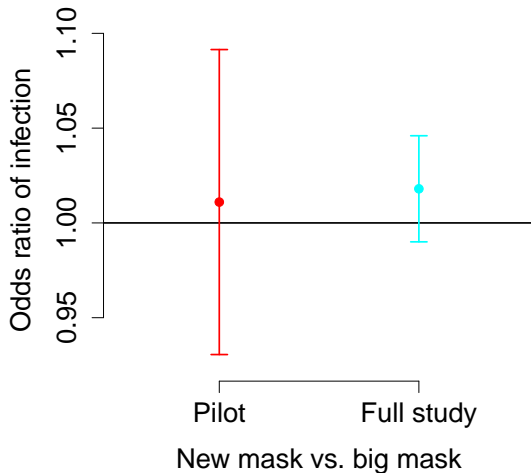
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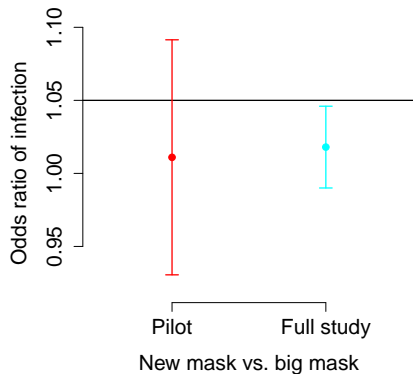
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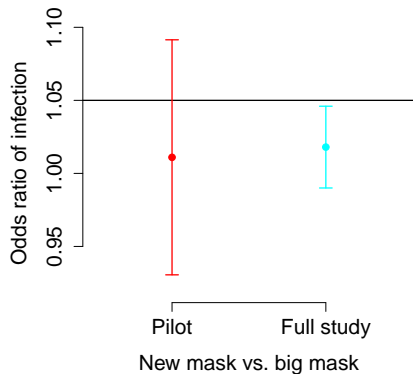


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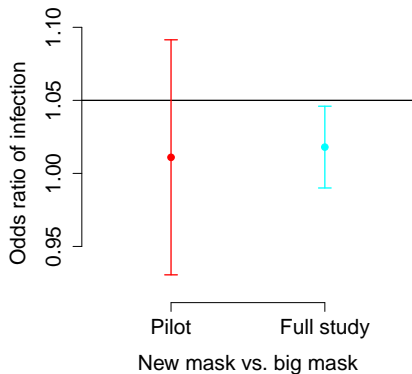
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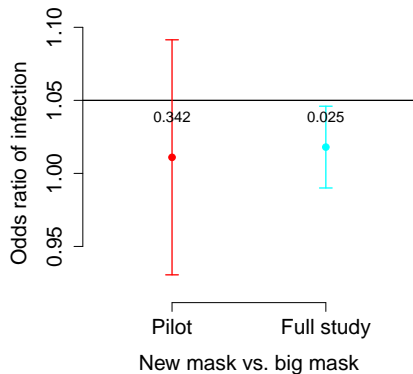
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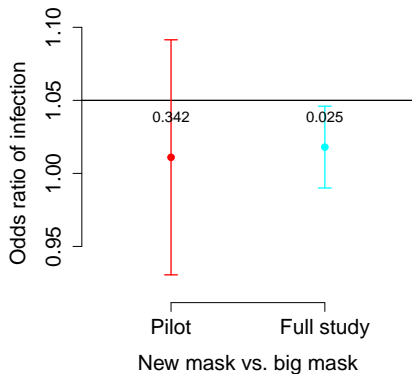
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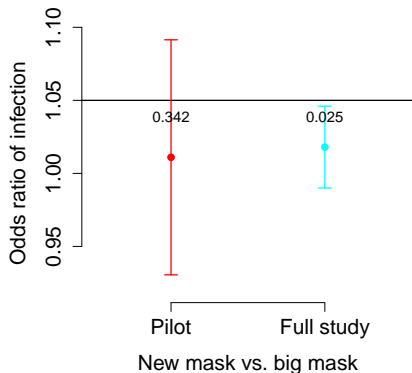


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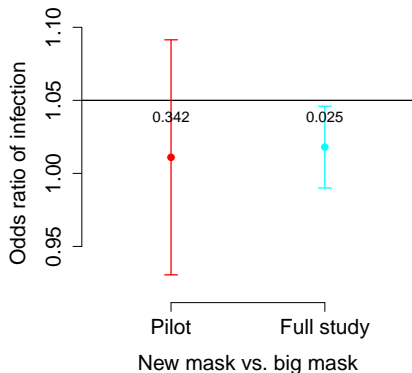
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## Null effects of boot camps and short-format training for PhD students in life sciences

David F. Feldon<sup>a,1</sup>, Soojeong Jeong<sup>a</sup>, James Peugh<sup>b</sup>, Josipa Roksa<sup>c,d</sup>, Cathy Maahs-Fladung<sup>a</sup>, Alok Shenoy<sup>a</sup>, and Michael Oliva<sup>a</sup>

<sup>a</sup>Department of Instructional Technology & Learning Sciences, Utah State University, Logan, UT 84322-2830; <sup>b</sup>Department of Pediatrics, Cincinnati Children's Hospital Medical Center, Cincinnati, OH 45229-3026; <sup>c</sup>Department of Sociology, University of Virginia, Charlottesville, VA 22904; and <sup>d</sup>Curry School of Education, University of Virginia, Charlottesville, VA 22904

Edited by Dale Purves, Duke University, Durham, NC, and approved July 28, 2017 (received for review April 6, 2017)

Many PhD programs incorporate boot camps and summer bridge programs to accelerate the development of doctoral students' research skills and acculturation into their respective disciplines. These brief, high-intensity experiences span no more than several weeks

and what they may have learned from a given experience is notoriously inaccurate (6–10).

Extensive evidence suggests that effective instruction or practice should be spaced out over an extended period to support mean-



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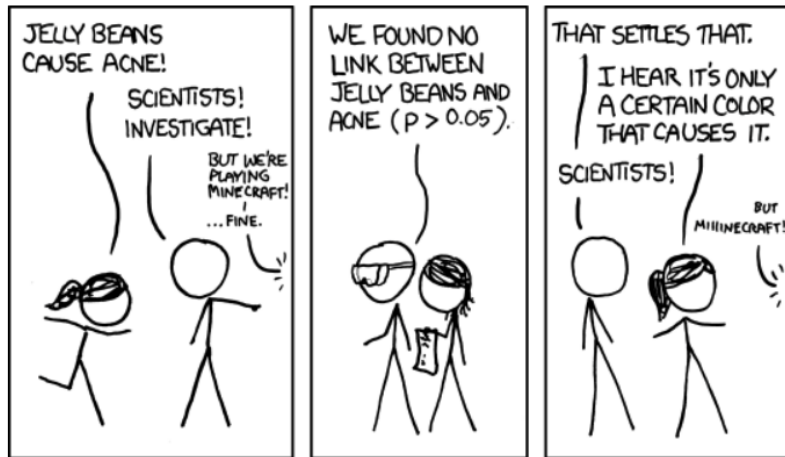
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# Confusion





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*synonyms:* importance, import, consequence, seriousness, gravity, weight, magnitude, momentousness; *formal* moment  
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