

Weight of Statistical Evidence

Detection and Correction of Publication Bias

Servan Grüninger Zurich, November 29th 2019

EBPhD Admission Interview

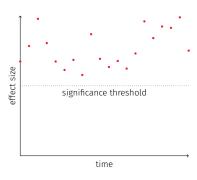
The Woozle effect



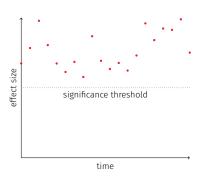
Pooh and Piglet tracking down the elusive Woozle (Image: Ernest H. Shepard)

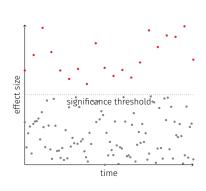
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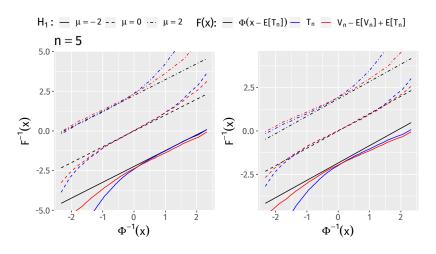
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$$V_n = h_n(T_n) = \sqrt{2n} \sinh^{-1}(T_n/\sqrt{2n})$$

Further improvement by finite sample correction:

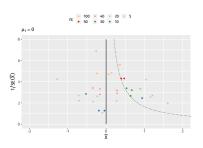
$$V_n^* = \frac{n - 1.7}{n - 1} \sqrt{2n} \sinh^{-1}(T_n / \sqrt{2n})$$

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Several methods assessed

• The funnel plot



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- · File drawer calculation



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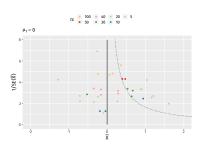


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- p-curve

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- Reweight by publication probability



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- Reweight by publication probability
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- Maximise truncated likelihood

Coming Full Circle



Fighting publication bias is like hunting the Woozle—all too often it forces you to go in circles