

Title: Workshop: Hypothesis Evaluation Using the Bayes Factor

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Abstract:

Since Cohen's (1994) paper "the earth is round,  $p < .05$ " there is increasing awareness that the null-hypothesis, e.g.,  $H_0: m_1 = m_2 = m_3$ , where the  $m$ 's denote the means in three groups, only rarely represents the expectations that researchers have. Informative hypotheses (Gu et al., 2018, Hoijtink et al., 2019) use equality and inequality constraints to formally represent researcher's expectation. Two (hypothetical) examples of such hypotheses are:  $H_1: m_1 > m_2 > m_3$  and  $H_2: m_1 - m_2 > m_2 - m_3$ . Since both  $H_1$  and  $H_2$  may be wrong, it is customary to add  $H_u: m_1, m_2, m_3$  to the set of hypotheses of interest. In  $H_u$  there are no restrictions on the parameters of interest. Only if  $H_1$  and  $H_2$  are better than  $H_u$  they may be valuable.

Additionally, in the last years there is increasing for alternatives for null-hypothesis significance testing. One such alternative, (informative) hypothesis evaluation using the Bayes factor, will be introduced. The Bayes factor quantifies the support in the data for a pair of hypotheses based on the fit and the complexity of the hypotheses. Loosely formulated, if, for example estimates of the three means in  $H_1$  are, 2, 5, and 7, respectively, then the fit of  $H_1$  is rather bad. It can also be seen that  $H_1$  is more specific than  $H_2$  (and therefore less complex) because it imposes more constraints on the three means. If, for example,  $BF_{12} = 5$  and  $BF_{1u} = 10$ , this means that the support in the data for  $H_1$  is 5 times larger than the support for  $H_2$  and 10 times larger than for  $H_u$ . This would imply that, currently,  $H_1$  is the best available description of the population of interest.

In the workshop it will be elaborated what the Bayes factor is, how it can be applied and should be interpreted. There will be attention for Bayesian updating (an alternative for power analysis), Bayesian (conditional) error probabilities, limitations of the approach, and the statistical underpinnings of the software with which the Bayes factor can be computed.

Assignment:

Participants can prepare by reading the tutorial about Bayesian hypothesis evaluation (Hoijtink et al, 2019, downloading the course materials, and installing R, R-studio, and Bain on their laptop. Course materials retrievable from <https://informative-hypotheses.sites.uu.nl/software/bain/> at the bottom of the page under SMiP-Mannheim 2/3 December 2021 (course materials will be updated mid November 2021).

Credits: 2 workshop days

## References:

Cohen, J. (1994). The earth is round,  $p < .05$ . *American Psychologist*, 49, 997-1003.

Gu, X., Mulder, J., and Hoijtink, H. (2018). Approximate adjusted fractional Bayes factors: A general method for testing informative hypotheses. *British Journal of Mathematical and Statistical Psychology*, 71, 229-261. DOI: 10.1111/bmsp.12110

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Hoijtink, H., Mulder, J., van Lissa, C., and Gu, X. (2019). A tutorial on testing hypotheses using the Bayes factor. *Psychological Methods*, 24, 539-556. DOI: 10.1037/met0000201

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