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April 6, 2018



#### PhD-student Machine Learning Group



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#### Projects

- Bayesian inference under model misspecification (SafeBayes)
- Foundations of Bayesianism
- Hypothesis testing (Frequentist, Bayesian, new methods)









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#### Frequentist vs. Bayesian probability

Different views on probability:

- ► Frequentists: frequency in a long run
- Bayesians: degree of belief

- Null Hypothesis  $\mathcal{H}_0$ :  $X^n \sim N(0, \sigma)$
- Alternative Hypothesis  $\mathcal{H}_1: X^n \sim N(\theta, \sigma), \ \theta \neq 0$

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  Limitations: see De Heide and Grünwald (2018)
- Test martingales and S-tests (current work)

**Optional stopping**: 'peeking at the results so far to decide whether or not to gather more data'

 $\rightarrow$  reproducibility crisis in life and behavioural sciences (far too many false positives)

- ▶ Data  $X^n = X_1, X_2, \dots, X_n$ ;  $X_i \in \mathcal{X}$
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#### Frequentist optional stopping

A sequence of hypothesis tests  $T_n : \mathcal{X}^n \to \{0, 1\}$  with significance level  $\alpha$  is said to be *robust under frequentist optional stopping* if

$$\mathbf{P}_{H_0}(\exists n: T_n(X^n)=1) \leq \alpha.$$

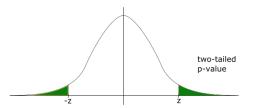
Example:

Data  $X^n \overset{\text{i.i.d.}}{\sim} N(\theta, 1)$ 

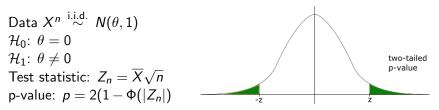
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Test statistic:  $Z_n = \overline{X}\sqrt{n}$   
p-value:  $p = 2(1 - \Phi(|Z_n|))$ 

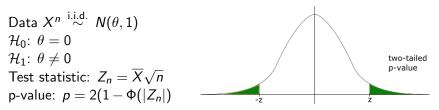


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Stopping rule: Continue until  $|Z_n| > k$ , then stop (with  $\alpha = 0.05, k = 1.96$ ).

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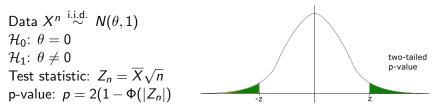


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This gives us:  $Z_n > \lambda \sqrt{2 \log \log n}$  i.o. w.p. 1, for  $\lambda < 1$ .

#### Frequentist Optional Stopping and Bayes Factors

Can Bayes Factor hypothesis tests handle frequentist optional stopping? (Bayesian statisticians suggest they do...)

- Not always (see Sanborn and Hills (2014); De Heide and Grünwald (2018) )
- Yes, if  $\mathcal{H}_0$  is simple
- Yes, with composite H<sub>0</sub> and certain group invariance structure (see Hendriksen, De Heide and Grünwald (2018))
  Example: Bayesian t-test

# Test Martingales (1)

The Test Martingale (Vovk et al., 2011):

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# Test Martingales (2)

Example:  $\mathcal{H}_0$ : the coin is fair, i.e. the outcomes  $X_1, X_2, \ldots$  are i.i.d. Bernoulli(0.5) Start with capital  $K_0 = 1$ At each round *i* 

- We invest a fraction q<sub>i</sub> of our capital on the outcome {X<sub>i</sub> = 0} and 1 − q<sub>i</sub> on {X<sub>i</sub> = 1}
- The outcome is revealed
- We get a pay-off: twice our stakes on the correct outcome, nothing on the other

After N rounds:  $K_N = 2^N \prod_{t=1}^N q(X_N | X^{N-1}).$ 

### Test Martingales (3)

- ▶ If  $\mathcal{H}_0$  is true, the bet is *fair*, i.e.  $\mathbf{E}[K_N] \leq K_0$
- If  $K_N$  is large, it is an indication that  $\mathcal{H}_0$  is not true.
- Test martingales can handle frequentist optional stopping.

Test martingales can handle frequentist optional stopping and have additional advantages over p-values (interpretability)

Problem: how to construct test martingales for composite  $\mathcal{H}_0$ .

Current work: S-value

#### References

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- A. Hendriksen, R. de Heide, and P. Grünwald. Folklore results on optional stopping of Bayesian methods made precise. *work in progress*, 2018.
- A. Sanborn and T. Hills. The frequentist implications of optional stopping on Bayesian hypothesis tests. *Psychonomic Bulletin & Review*, 21(2):283–300, 2014.
- V. Vovk, N. Vereshchagin, A. Shen, and G. Shafer. Test martingales, Bayes factors and p-values. *Statistical Science*, 26 (1):84–101, 2011.