NIST panel on quantifying the weight of evidence

Doug Armstrong South Dakota State University



South Dakota State University

- If one is subjective in choice of construction of LR/BF, it should be assumed he/she is coherent and follows the axioms of probability
- Assumptions are made with any model choice and must be valid and upheld
- If/when those model assumptions are found to be invalid, the model is wrong, and the subjective "justification" is wrong and the individual is incoherent
 - Some violations are more/less severe
- Some SLRs do not satisfy basic principles



-Principles of a well-constructed SLR

Not an exclusive list!

- Coherency Principle
 - In Bayesian decision theory, the coherency principle assumes degrees of belief obey to the axioms of probability and that consistent decisions can be made based on personal probabilities
- Sufficiency Principle
 - All Sufficient statistics based on data x for a given model $p_{\theta}(x)$ should lead to the same conclusion
 - Also provides that choice in score is a sufficient statistic
- Likelihood Principle
 - Two datasets that produce equal/proportional likelihoods should lead to the same conclusion



-An example of the Coherency principle

- Let H_A, H_B be two mutually exclusive propositions for a population of 2 sources A, B with a set of observations made on a trace e_u and control objects e_A, e_B. The LR should not be influenced by the order of proposition.
- Under coherency we get:

$$LR_{A,B} = \frac{\Pr(e_u|H_A)}{\Pr(e_u|H_B)} = \frac{1}{LR_{B,A}} = \frac{1}{\frac{\Pr(e_u|H_B)}{\Pr(e_u|H_B)}}$$



South Dakota State University

-An example of the Coherency principle

• However, with some SLRs:

$$LR_{H_A} = \frac{\Pr(e_u|H_A)}{\Pr(e_u|H_B)} \approx \frac{\Pr(\delta(e_u, e_A)|H_A)}{\Pr(\delta(e_u, e_A)|H_B)} \neq \frac{1}{\frac{\Pr(\delta(e_u, e_B)|H_B)}{\Pr(\delta(e_u, e_B)|H_A)}}$$

None of the SLRs we have studied uphold this principle!



-Lack of convergence



South Dakota State University

-SLRs do not consistently meet all principles

- By using some basic principles, we can objectify the selection of a score-based model
- At the end of the day, the courts need to trust that models used to estimate the LR/BF are valid!



Calibration of the LR

- It is acceptable to calibrate a LR, as long as the resulting calibrated-LR is not reported as a LR
 - When used in this manner, the LR behaves as a metric, and a system is optimized to reduce error rates.
 - A form of supervised training
- These are great in decision engines (speech for instance)



Calibration of the LR

- Some methods calibrate the LR via the posterior odds
 - These methods suggest prior odds to obtain optimized posteriors
- Calibrated LRs are not appropriate in court-room settings
 - If reported in court, they need to be made clear what they are
 - Any reported LR still needs to follow the first rule of LRs:
 - If LR > 1, support the prosecution (numerator) hypothesis
 - If LR < 1, support the defense (denominator) hypothesis



Future of the SLR

- We still want to use the data reduction techniques that make SLRs attractive in the first place.
- A method which takes into consideration the relationships and dependencies between the objects in the population of potential sources.
- A method which has similar theoretical properties as the LR



Future of the SLR

- A method has been suggested by Gantz & Saunders to estimate the numerator of a LR. The method makes use of kernel-based methods, prevalent in pattern recognition
- My current research is focused on extending this method to the denominator and suggesting a set of conditions necessary to ensure convergence to the ideal LR



Acknowledgement

 The opinions and conclusions or recommendations expressed in this presentation are those of the author and do not necessarily represent those of the Department of Justice. The research presented here was supported by the National Institute of Justice, Office of Justice Programs, US Department of Justice under Awards No. 2014-IJ-CX-K088 & 2015-R2-CX-0028.

