

## **Bayesian Confirmation Measures** and Their Properties – A New Perspective

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Nowe spojrzenie na Bayesowskie miary konfirmacji i ich właściwości

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## Rule induction

- Patterns in form of rules are induced from a data table
- $S = \langle U, A \rangle data \ table$ , where U and A are finite, non-empty sets U - universe of objects; A - set of attributes
- $S = \langle U, C, D \rangle$  *decision table*, where C set of *condition attributes*, D – set of *decision attributes*,  $C \cap D = \emptyset$
- Rule induced from S is a consequence relation:
   E → H read as if E then H where

*E* is condition (evidence or premise) and*H* is conclusion (hypothesis or decision)formula built from attribute-value pairs (q,v)

## Rule induction

U	Height	Hair	Eyes	Nationality	Support
1	tall	blond	blue	Swede	270
2	medium	dark	hazel	German	90
3	medium	blond	blue	Swede	90
4	tall	blond	blue	German	360
5	short	red	blue	German	45
6	medium	dark	hazel	Swede	45
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Characterization of nationalities

• E.g. decision rules induced from "characterization of nationalities":

- 1) **If** (*Height=tall*), **then** (*Nationality=Swede*)
- 2) If (Height=medium) & (Hair=dark), then (Nationality=German)

#### **Motivations**

#### The number of rules

induced from datasets is usually quite large

- overwhelming for human comprehension,
- many rules are irrelevant or obvious (low practical value)

rule evaluation – interestingness (attractiveness) measures (e.g. support, confidence, gain, rule interest, lift, measures of Bayesian confirmation)

• each measure was proposed to capture different characteristics of rules

• the number of proposed measures is very large

## Motivations

The choice of an interestingness measure for a certain application is a difficult problem

- there is no evidence which measure(s) is the best
- the users' expectations vary,
- the number of proposed measures is overwhelming

properties of interestingness measures, which reflect users' expectations towards the behavior of measures in particular situations



need to analyze measures with respect to their properties

### Presentation plan

- Property of confirmation and its different definitions
- Popular confirmation measures
- Properties of confirmation measures
  - Symmetry properties
  - Property of concordance
- Summary

## Notation

 Used notation corresponding to a 2x2 contingency table of rule's premise and conclusion

a=sup(H,E) is the number of objects in *U* satisfying both the premise *E* and the conclusion *H* of a rule  $E \rightarrow H_r$ 

	Σ
$c=sup(\neg H, E), F a c$	a+c
$d=sup(\neg H, \neg E),  \neg E \qquad b \qquad d$	b+d
$a+c=sup(E)$ , $\Sigma$ $a+b$ $c+d$	a+b+c+d=n

a+b=sup(H),...

 a, b, c and d can also be regarded as frequencies that can be used to estimate probabilities:

e.g., P(E) = (a+c)/n, P(H) = (a+b)/n, P(H|E) = a/(a+c).

Generally, measures possessing the property of confirmation (confirmation measures) are expected to obtain:

- values >0 when the premise of a rule confirms the conclusion,
- values = 0 when the rule's premise and conclusion are neutral to each other,
- values < 0 when the premise disconfirms the conclusion.</p>
- What does "premise confirms conclusion" mean?
- How to quantify such confirmation?

- Four definitions in the literature:
  - Bayesian confirmation
  - strong Bayesian confirmation:  $P(H|E) > P(H|\neg E)$
  - likelihoodist confirmation: P(E|H) > P(E)
  - strong likelihoodist confirmation:  $P(E|H) > P(E|\neg H)$
- An attractiveness measure c(H,E), has the property of Bayesian confirmation if is satisfies the following condition:

$$c(H,E) \begin{cases} > 0 \text{ if } P(H|E) > P(H) \\ = 0 \text{ if } P(H|E) = P(H) \\ < 0 \text{ if } P(H|E) < P(H) \end{cases}$$

- Bayesian approach is related to the idea that the *E* confirms *H*, if *H* is more frequent with *E* rather than with ¬*E* (perspective of rule's conclusion)
- Bayesian confirmation: P(H|E) > P(H)
  - *H* is satisfied more often when *E* is satisfied (then, this frequency is P(*H*|*E*)), rather than generically (P(*H*)) Assumtion: P(*E*)≠0
- strong Bayesian confirmation:  $P(H|E) > P(H|\neg E)$ 
  - *H* is satisfied more often, when *E* is satisfied, rather than when *not E* is satisfied Assumtion: P(*E*)≠0, P(¬*E*)≠0

- Likelihoodist approach is based on the idea that *E* confirms *H*, if *E* is more frequent with *H* rather than with ¬*H* (perspective of rule's premise)
  - likelihoodist confirmation: P(E|H)>P(E)
  - strong likelihoodist confirmation:  $P(E|H) > P(E|\neg H)$

## Logical equivalance of four definitions of confirmation

- Bayesian confirmation: a/(a+c) > (a+b)/n
- strong Bayesian confirmation: a/(a+c) > b/(b+d)
- likelihoodist confirmation: a/(a+b)>(a+c)/n
- strong likelihoodist confirmation: a/(a+b) > c/(c+d)
- Obviously, the above definitions differ.
  - What is the relationship between them?
  - Do they "switch" (between +, zero and —) at the same times?
- All four definitions boil down to one general, always-defined formulation:

$$c(H,E) \begin{cases} > 0 \ if \ ad - bc > 0 \\ = 0 \ if \ ad - bc = 0 \\ < 0 \ if \ ad - bc < 0 \end{cases}$$

Advantage: *ad-bc* is never undefined, no denominator

## Popular confirmation measures

$$D(H, E) = P(H | E) - P(H) = \frac{a}{a+c} - \frac{a+b}{n} = \frac{ad-bc}{n(a+c)}$$

$$M(H, E) = P(E | H) - P(E) = \frac{a}{a+b} - \frac{a+c}{n} = \frac{ad-bc}{n(a+b)}$$

$$S(H, E) = P(H | E) - P(H | \neg E) = \frac{a}{a+c} - \frac{b}{b+d} = \frac{ad-bc}{(a+c)(b+d)}$$

$$N(H, E) = P(E | H) - P(E | \neg H) = \frac{a}{a+b} - \frac{c}{c+d} = \frac{ad-bc}{(a+b)(c+d)}$$

$$C(H, E) = P(E \land H) - P(E)P(H) = \frac{a}{n} - \frac{(a+c)(a+b)}{n^2} = \frac{ad-bc}{n^2}$$

$$F(H, E) = \frac{P(E | H) - P(E | \neg H)}{P(E | H) + P(E | \neg H)} = \frac{\frac{a}{a+b} - \frac{c}{c+d}}{\frac{a}{a+b} + \frac{c}{c+d}} = \frac{ad-bc}{ad+bc+2ac}$$

Popular confirmation measures

$$D(H,E) = P(H | E) - P(H) = \frac{a}{a+c} - \frac{a+b}{n} = \frac{ad-bc}{n(a+c)}$$

- Notice, that measure D(H,E) is undefined whenever a+c=0, i.e. when a=c=0 (we exclude degenerated cases when n=0).
- Exemplary dataset with 6545 different contingency tables
   (combinations of *a*, *b*, *c* and *d*) contained 33 cases when *a*=*c*=0.
- Solution: use *ad-bc* >0 definition of confirmation
  - whenever ad=bc, i.e. also when a=c=0 assume that D(H,E)=0.

## Popular confirmation measures

$$Z(H,E) = \begin{cases} 1 - \frac{P(\neg H \mid E)}{P(\neg H)} = \frac{ad - bc}{(a+c)(c+d)} \text{ in case of confirmation} \\ \frac{P(H \mid E)}{P(H)} - 1 = \frac{ad - bc}{(a+c)(a+b)} \text{ in case of disconfirmation} \end{cases}$$

$$A(H,E) = \begin{cases} \frac{P(E \mid H) - P(E)}{1 - P(E)} = \frac{ad - bc}{(a+b)(b+d)} & \text{ in case of confirmation} \\ \frac{P(H) - P(H \mid \neg E)}{1 - P(H)} = \frac{ad - bc}{(b+d)(c+d)} & \text{ in case of disconfirmation} \end{cases}$$

#### Derived confirmation measures

$$c_{1}(H,E) = \begin{cases} \alpha + \beta A(H,E) & \text{in case of confirmation when } c = 0 \\ \alpha Z(H,E) & \text{in case of confirmation when } c > 0 \\ \alpha Z(H,E) & \text{in case of disconfirmation when } a > 0 \\ -\alpha + \beta A(H,E) & \text{in case of disconfirmation when } a = 0 \end{cases}$$

0

$$c_{2}(H,E) = \begin{cases} \alpha + \beta Z(H,E) \text{ in case of confirmation when } b = 0\\ \alpha A(H,E) \text{ in case of confirmation when } b > 0\\ \alpha A(H,E) \text{ in case of disconfirmation when } d > 0\\ -\alpha + \beta Z(H,E) \text{ in case of disconfirmation when } d = 0 \end{cases}$$

 $c_{3}(H,E) = \begin{cases} A(H,E)Z(H,E) & \text{in case of confirmation} \\ -A(H,E)Z(H,E) & \text{in case of disconfirmation} \end{cases}$ 

 $c_4(H,E) = \begin{cases} \min(A(H,E),Z(H,E)) \text{ in case of confirmation} \\ \max(A(H,E),Z(H,E)) \text{ in case of disconfirmation} \end{cases}$ 

#### Notation – reminder

- Caution! in the following *c* stands for:
  - c(H,E) a confirmation measures (general)
  - c<sub>1</sub>(H,E), c<sub>2</sub>(H,E), c<sub>3</sub>(H,E), c<sub>4</sub>(H,E) particular confirmation measures
  - **c** one of the *a*, *b*, *c*, *d* frequencies in the contingency table

## Symmetry properties

#### Symmetry properties

 Symmetry properties are formed by applying the negation operator to the rule's premise/conclusion, or both, as well as switching the position of the premise and the conclusion.

• Example:

???  $c(H,E) = c(\neg H, E)$  c(H,E) = c(E, H)  $c(H,E) = c(\neg E, \neg H)$ 

## Symmetry properties – Carnap, Eells & Fitelson

- Carnap, Eells and Fitelson have analyzed confirmation measures from the viewpoint of four properties of symmetry
  - evidence symmetry ES:  $c(H, E) = -c(H, \neg E)$
  - hypothesis symmetry HS:  $c(H, E) = -c(\neg H, E)$
  - inversion(commutativity) symmetry IS: c(H, E) = c(E, H)
  - evidence-hypothesis (total) symmetry EHS:  $c(H, E) = c(\neg H, \neg E)$
- Their conclusion: only hypothesis symmetry HS is a desirable property

Carnap, R., 1962. Logical Foundations of Probability, Univ. of Chicago Press, Chicago.

Eells, E., Fitelson, B., 2002. Symmetries and asymmetries in evidential support. Philosophical Studies, 107 (2): 129-142.

## *Hypothesis Symmetry* (HS)



## *Evidence Symmetry* (ES)



## Inversion Symmetry (IS)



## Evidence-hypothesis Symmetry (EHS)



#### Symmetry properties - Crupi et al.

Recently, Crupi, Tentori and Gonzalez propose to analyze a confirmation measure c(H, E) with respect to the following symmetries

 $ES(H, E): c(H, E) = -c(H, \neg E)$   $EIS(H, E): c(H, E) = -c(\neg E, H)$ 
 $HS(H, E): c(H, E) = -c(\neg H, E)$   $HIS(H, E): c(H, E) = -c(E, \neg H)$  

 IS(H, E): c(H, E) = c(E, H)  $EHIS(H, E): c(H, E) = c(\neg E, \neg H)$ 
 $EHS(H, E): c(H, E) = c(\neg H, \neg E)$ 

- Crupi et al. claim that the analysis should be conducted separately for:
  - the case of confirmation (i.e. when P(H|E) > P(H)), and
  - for the case of disconfirmation (i.e. when P(H|E) < P(H))
- Such approach results in 14 symmetry properties

Crupi, V., Tentori, K., Gonzalez, M. ,2007. On Bayesian measures of evidential support: Theoretical and empirical issues, *Philosophy of Science*, vol. 74, 229-252.

### Crupi et al. symmetries – inversion symmetry

- Crupi et al. concur with the results of Eells and Fitelson regarding the inversion symmetry only in case of confirmation
- Crupi et al. claim that IS is desirable in case of disconfirmation
- Let us consider a rule:

#### if the drawn card is an Ace, then it is a face

- the strength with which an Ace disconfirms face is the same as the strength with which the face disconfirms an Ace,
  i.e. c(H, E) = c(E, H)
- Conclusions of Crupi et al.:
  - in case of confirmation only the HS, HIS and EHIS are the desirable properties
  - in case of disconfirmation only HS, EIS and IS properties are the desirable properties

## Symmetries for Bayesian confirmation - doubts

- The propositions of Eells and Fitelson as well as Crupi et al. are dedicated for the definition of the Bayesian confirmation: P(H|E)>P(H)
- Their reasoning is based on assumption that:
  - the highest confirmation should occur
     in case of entailment (E|=H ⇔ P(H|E)=1 ⇔ c=0)
  - the highest disconfirmation should occur in case of refutation  $(E|=\neg H \Leftrightarrow P(H|E)=0 \Leftrightarrow a=0)$
- Such reasoning boils down to verification whether P(H|E) is 1 or 0
- However the definition of Bayesian confirmation also takes into account P(H).
- Confirmation measures should somehow express:
   what is the "gain" for H from knowing that E occured.
   We want to know if passing from P(H) to P(H|E) is profitable or not.

## Symmetries for Bayesian confirmation - doubts

- We want to know if passing from P(H) to P(H|E) is profitable or not.
- The biggest profits when
  - P(H) is minimal and
  - $P(H|E)=a/(a+c)=1 \Leftrightarrow c=0.$
- Practical problem: determination when P(H) = (a+b)/n is minimal
  - we want to have a case of confirmation (ad>bc), so at least  $a\neq 0$
  - $P(H) \rightarrow 0$  when  $n \rightarrow \infty$ ,

but we have a closed world of a decision table

Solution: use the definition of strong Bayesian confirmation  $P(H|E) > P(H|\neg E)$ 

### Strong Bayesian confirmation

- The definition of strong Bayesian confirmation  $P(H|E) > P(H|\neg E)$ 
  - Confirmation measures should express:
     what is the "gain" (profit ) for H from passing from ¬E to E
  - The biggest profits when:
    - $P(H|\neg E) = 0$  (i.e.,  $b/(b+d)=0 \Leftrightarrow b=0$ ) and
    - P(H|E)=1 (i.e.,  $a/(a+c)=1 \Leftrightarrow c=0$ ).

- A confirmation measure should give an account of the credibility that it is more probable to have the conclusion when the premise is present, rather than when the premise is absent
- Both conditional probabilities P(H|E) and  $P(H|\neg E)$  should be considered both in case of confirmation and disconfirmation
- There is no need to treat case of confirmation and disconfirmation separately

• ES:  $c(H, E) = -c(H, \neg E)$  is desirable for strong Bayesian confirmation  $(P(H|E) > P(H|\neg E))$ 

•  $E \rightarrow H$ : if the drawn card is the **7**, then the card is **black** 

■ Let us observe, that for c(H, E) we have that  $P(H|\neg E) = b/(b+d) = 25/51 = 0.49$  and

 $\rightarrow$  P(H|E)=a/(a+c)=1, which gives us a 49% increase of confirmation

• On the other hand, for 
$$c(H, \neg E)$$
 we get:

$$P(H| \neg \neg E) = P(H|E) = 1$$
 and

- P( $H|\neg E$ )=0.49, which results in 49% decrease
- Thus, clearly the confirmation of a rule  $E \rightarrow H$  should be of the same value but of the opposite sign as the confirmation of a  $\neg E \rightarrow H$  rule

Η

a=1

b=25

F

 $\neg F$ 

 $\neg H$ 

c=0

*d*=26

• ES:  $c(H, E) = -c(H, \neg E)$  is desirable

Let us examine both sides of this equation using an exemplary scenario where the values of contingency table of *E* and *H* are:

	Н	¬ H
Ε	a=100	c=0
¬ E	b=10	d=40

• Let us observe, that for 
$$c(H, E)$$
 we have that

P( $H|\neg E$ ) = b/(b+d)=0.20 and P(H|E)=a/(a+c)=1, which gives us a 80% increase of confirmation

• On the other hand, for  $c(H, \neg E)$  we get:

P(H|E)=1 and

 $P(H|\neg E)=0.20$ , which results in 80% decrease

Thus, clearly the confirmation of a rule  $E \rightarrow H$  should be of the same value but of the opposite sign as the confirmation of a  $\neg E \rightarrow H$  rule

ES	YES for any (H,E) $c(H, E) = -c(H, \neg E)$
HS	YES for any (H,E) $c(H, E) = -c(\neg H, E)$
EIS	NO for some (H,E) $c(H, E) \neq -c(\neg E, H)$
HIS	NO for some (H,E) $c(H, E) \neq -c(E, \neg H)$
IS	NO for some (H,E) $c(H, E) \neq c(E, H)$
EHS	YES for any (H,E) $c(H, E) = c(\neg H, \neg E)$
EHIS	NO for some (H,E) $c(H, E) \neq c(\neg E, \neg H)$

## Symmetries for different definitions of confirmation

- Systematic approach to symmetry properties in the context of different definitions of confirmation
- Let us focus only on the probabilities involved in different definitions of confirmation:
  - a) P(H|E) and P(H) for Bayesian confirmation
  - b) P(H|E) and  $P(H|\neg E)$  for strong Bayesian confirmation
  - c) P(E|H) and P(E) for likelihoodist confirmation
  - d) P(E|H) and  $P(E|\neg H)$  for strong likelihoodist confirmation

Symmetries for different definitions of confirmation

- From basic probability theory:
  - a) for Bayesian confirmation P(H|E) > P(H):  $P(\neg H|E) = 1 - P(H|E)$  and
    - $P(\neg H)=1-P(H);$
  - hypothesis symmetry: c(H,E)=-c(¬ H,E)
     because P(¬H|E)<P(¬ H) is equivalent to P(H|E)> P(H)

Symmetries for different definitions of confirmation

- From basic probability theory:
  - b) for strong Bayesian confirmation P(H|E) > P(H|-E):  $P(\neg H|E) = 1 - P(H|E)$  and  $P(\neg H| \neg E) = 1 - P(H| \neg E);$
  - hypothesis symmetry:  $c(H,E) = -c(\neg H,E)$ because  $P(\neg H|E) < P(\neg H| \neg E)$ is equivalent to  $P(H|E) > P(H|\neg E)$
  - evidence symmetry: c(H,E)=-c(H, ¬E)
     because P(H| ¬E)<P(H| E)</li>
     is equivalent to P(H|E)< P(H|¬E)</li>
  - evidence-hypothesis symmetry:



## Symmetries for different definitions of confirmation - summary

Definition of confirmation	Desirable symmetry
Bayesian confirmation	HS
strong Bayesian confirmation	ES, HS, EHS
likelihoodist confirmation	ES
strong likelihoodist confirmation	ES, HS, EHS

## Property of concordance

# Thank you!