

## Visual-based Detection of Properties of Confirmation Measures

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#### Presentation plan

- Rule induction and interestingness measures
- Confirmation measures
- Properties of confirmation measures
  - Property of monotonicity M
  - Property of maximality/minimality
  - Property of hypothesis symmetry
- Visualization of measures
  - The experimental dataset
  - Visualization technique
  - Visual-based detection of properties
- Conclusions



#### 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

Chai	Characterization of nationalities					
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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• 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)

If	Evidence	then	Hypothesis
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 $E \rightarrow H$ 

#### Interestingness measures

#### The number of rules

induced from data sets 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

In this work we focus on a group of measures called **measures of confirmation** 

#### Notation

• Used notation corresponding to a 2x2 contingency table of rule's premise and conclusion. For a rule  $E \rightarrow H$ :

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

$b=sup(H, \neg E),$		Н	¬ H	Σ
$c=sup(\neg H, E),$	Е	а	С	a+c
$d=sup(\neg H, \neg E),$	¬ E	Ь	d	b+d
a+c=sup(E).	Σ	a+b	c+d	a+b+c+d=n

a+b=sup(H),...  $a, b, c, d \ge 0$ 

*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).

### Notation

Height	Hair	Eves	Nationality							
tall	blond	blue	Swede		$\neg E$	$\neg H$			Η	$\neg H$
medium	dark	hazel	German		$\neg E$	Н		E	1	0
medium	blond	blue	Swede		$\neg E$	$\neg H$		$\neg F$	2	3
tall	blond	blue	German		$\neg E$	Н		• —		
short	red	blue	German		Ε	Н		a =	sup(E	,H)
medium	dark	hazel	Swede		$\neg E$	$\neg H$		b = c = c	sup(–	(E,H)
		•		-			-	d = d	sup(L sun(_	,_,,, ,FH)

if	(Hair = red) & (Eyes = blue)	then	(Nationality = German)
if	Evidence	then	Hypothesis

 The contingency table is a form used to calculate the value of interestingness measures (e.g. confirmation measures)

#### Confirmation measures

An interestingness measure c(H,E) has the property of confirmation (i.e. is a confirmation measure) if is satisfies the following condition:

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

- Measures of confirmation quantify the strength of confirmation that premise *E* gives to conclusion *H*
- "*H* is verified more often, when *E* is verified,
  rather than when *E* is not verified"

#### Selected confirmation measures

There are many alternative, non-equivalent measures of confirmation

 $\mathbf{M}(H,E) = \frac{a}{a+b} - \frac{a+c}{|U|}$ (Mortimer 1988)  $S(H,E) = \frac{a}{a+c} - \frac{b}{b+d}$ (Christensen 1999)  $F(H,E) = \frac{ad - bc}{ad + bc + 2ac}$ (Kemeny and Oppenheim 1952)  $FS(H,E) = \frac{1}{2}(F(H,E) + S(H,E))$ (Glass 2013)  $\phi(H,E) = \frac{ad-bc}{\sqrt{(a+c)(a+b)(b+d)(c+d)}}$ (Mortimer 1988)  $F\phi(H,E) = \frac{1}{2}(F(H,E) + \phi(H,E))$ (Glass 2013)

- The values of all of the above measures range from -1 to +1,
- otherwise they are undefined, e.g. when a+b=0 measure M(H,E) is NaN.

## Properties of confirmation measures

The choice of a confirmation measure for a certain application is a difficult problem

- •the number of proposed measures is overwhelming
- there is no evidence which measure is the best
- the users' expectations vary

properties of confirmation measures, which reflect users' expectations towards the behaviour of measures in particular situations



need to analyze measures with respect to their properties

Motivation for this work: **Detect properties of measures** and compare measures easily through their visualizations

#### Property of monotonicity M

	Н	¬ H
E	а	С
¬ E	b	d

• Desirable property of c(H,E) = f(a,b,c,d) : monotonicity (M)\*

*f* should be non-decreasing with respect to *a* and *d* and non-increasing with respect to *b* and *c* 

- Interpretation of (M):  $(E \rightarrow H \equiv if x is a raven, then x is black)$ 
  - a) the more black ravens we observe, the more credible becomes  $E \rightarrow H$
  - b) the more black non-ravens we observe, the less credible becomes  $E \rightarrow H$
  - c) the more non-black ravens we observe, the less credible becomes  $E \rightarrow H$
  - d) the more non-black non-ravens we observe, the more credible becomes  $E \rightarrow H$

\*S.Greco, Z.Pawlak, R.Słowiński: Can Bayesian confirmation measures be useful for rough set decision rules? *Engineering Applications of Artificial Intelligence*, 17 (2004) no.4, 345-361

## Property of maximality/minimality

	Н	¬ H
E	а	С
¬ E	b	d

Desirable property of c(H,E): maximality/minimality\*

c(H,E) is maximal if and only if  $P(E, \neg H) = P(\neg E, H) = 0$  and

c(H,E) is minimal if and only if  $P(E,H) = P(\neg E, \neg H) = 0$ .

Interpretation of maximality/minimality:

a measure obtains its maximum iff c=b=0 and its minimum iff a=d=0.

\*Glass, D.H.: Confirmation measures of association rule interestingness, Knowledge-Based Systems 44, (2013) 65–77

## Property of hypothesis symmetry HS

Desirable property of c(H,E): hypothesis symmetry (HS)\*

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

Interpretation of (HS):  $(E \rightarrow H = if x is a square, then x is rectangle)$ 

the strength with which

the premise (*x* is a square) confirms the conclusion (*x* is rectangle)

is the same as the strength with which

the premise disconfirms the negated conclsuion (*x* is not a rectangle).

\*Carnap, R.: Logical Foundations of Probability, second ed. University of Chicago Press, Chicago (1962)

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

Visualization of measures

#### The experimental dataset

- Given n > 0 (the total number of observations), a synthetic data set is generated as the set of all possible contingency tables satisfying
  a + b + c + d = n
- The set is thus exhaustive and non-redundant (i.e. it contains exactly one copy of each contingency table satisfying the above condition)

Height	Hair	Eyes	Nationality
tall	blond	blue	Swede
medium	dark	hazel	German
medium	blond	blue	Swede
tall	blond	blue	German
short	red	blue	German
medium	dark	hazel	Swede

H ¬H E a C → E b d

а	b	С	d
0	0	0	6
0	0	1	5
0	0	2	4
0	0	3	3
0	0	4	2
0	0	5	1
0	0	6	0
0	0	5	1
0	1	0	5
0	1	1	4
0	1	2	3
6	0	0	0

### Visualization technique - barycentric coordinates

- Our synthetic data set comprises t rows and 4 columns: a, b, c and d; t=(n+1)(n+2)(n+3)/6
- In general, four independent columns correspond to four degrees of freedom, visualization of such data in the form of a scatter-plot would formally require four dimensions.
- Owing to the condition a + b + c + d = n however, the number of degrees of freedom is reduced to three, so it is possible to visualize such data in three dimensions (3D) using tetrahedron-based barycentric coordinates
- The 3D view of the tetrahedron, proposed in the paper, has its four vertices *A*, *B*, *C* and *D* coinciding with points of the following [*x*, *y*, *z*] coordinates: *A*: [1,1,1]
  *C*: [-1,-1,1]
  *B*: [-1,1,-1]
  *D*: [1,-1,-1]

#### Visualization technique - barycentric coordinates



- the vertex A corresponds to the (single) contingency table satisfying a=n and b=c=d=0,
- the edge AB corresponds to the (multiple) contingency tables satisfying a+b=n and c=d=0,
- the face ABC corresponds to the (multiple) contingency tables satisfying a+b+c=n and d=0, etc.

#### Visualization technique – colour map

- Because the individual points of the tetrahedron may be displayed in colour, it is possible to visualize a function f(a,b,c,d) of the four arguments (e.g. any interestingness measure)
- It is assumed that the value set of this function is a real interval [r,s], with r < s, so that its values may be rendered using a pre-defined colour map
- For all the analysed confirmation measures the standard colour map ranges from -1 to +1:





 Non-numeric values, i.e. +∞, NaN and -∞, if generated by a particular function, may be rendered as colours not occurring in the map.

## Visualization technique – exemplary external visualizations



## Visualization technique – exemplary internal visualizations



Visual-based detection of properties

## Property of monotonicity M

	Н	¬ H
E	а	С
¬ E	b	d

• Desirable property of c(H,E) = f(a,b,c,d) : monotonicity (M)

*f* should be non-decreasing with respect to *a* and *d* and non-increasing with respect to *b* and *c* 

- Visual-based detection:
  - the "non-decreasing with a and d" condition should be reflected in the visualization as colours changing towards dark brown (increase of confirmation) around vertices A and D and
  - the "non-increasing with b and c" condition should be reflected in the visualization as colours changing towards dark blue (increase of disconfirmation) around vertices B and C
  - a thorough analysis with respect to property M requires an insight into the tetrahedron as potential counterexamples to this property may be located inside the shape

## Measure M(H,E) – counterexamples to property M



Clearly, measure M(H,E) does not satisfy property M, as in the visualization the colour changes from dark brown at vertex D to pale green at vertex A, violating the demands the of the non-decrease with a.

### Measure S(H,E) – no observable counterexamples to M



There are no observable counterexamples to property M in the external visualizations of measure S(H,E) which, together with additional analysis of the shape's inside, determines the possession of the property M by S(H,E).

## Property of maximality/minimality

	Н	¬ H
E	а	С
¬ E	b	d

Desirable property of c(H,E): maximality/minimality\*

c(H,E) is maximal if and only if b=c=0 and

c(H,E) is minimal if and only if a=d=0.

- Visual-based detection:
  - the dark brown (dark blue) colour must be found on the AD (BC) edge of the tetrahedron and cannot be found anywhere else

Let us observe that the AD (BC) edge contains all points for which b=c=0 (a=d=0), i.e., the points most distant from the vertices B and C (A and D)

 a thorough analysis with respect to maximality/minimality requires an insight into the tetrahedron as potential counterexamples to this property may be located inside the shape

## Measure F(H,E) – counterexamples to property max/min



Visual-based detection of maximality/minimality property reveals that measure F(H,E) does not satisfy this property. It is, among others, due to the fact that the points with maximal values of F(H,E) cover the whole *ABD* face (i.e., there are too many of them).

#### Measure FS(H,E) – no observable counterexamples to max/min



There are no observable counterexamples to property max/min in the external visualizations of measure FS(H,E) which, together with additional analysis of the shape's inside, determines the possession of the maximality/minimality property by FS(H,E).

Property of hypothesis symmetry HS

$$E \rightarrow H \qquad E \rightarrow \neg H$$

Desirable property of c(H,E): hypothesis symmetry (HS)

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

- Visual-based detection:
  - $c(H,E)=f(a,b,c,d) = -c(\neg H,E) = -f(a', b', c', d') = -f(c,d,a,b)$ , reflecting the exchange of columns in the contingency tables (a=c', b=d', c=a' d=b')
  - two views must have the same gradient profile (i.e., the left view must be just like the right one, provided the colour map is reversed)
  - if the "recoloured" views are not the same, then the visualized measure does not possess the hypothesis symmetry
  - a thorough analysis with respect to HS requires an insight into the tetrahedron as potential counterexamples to this property may be located inside the shape

### Measure M(H,E) – counterexamples to property HS



Clearly, measure M(H,E) does not satisfy property HS since e.g., the BCD face has a gradient profile that is characterized by straight lines, while the DAB face has a profile that is characterized by curved lines.

### Measure $F\Phi(H,E)$ – no observable counterexamples to max/min



There are no observable counterexamples to property HS in the external visualizations of measure  $F\Phi(H,E)$  which, together with additional analysis of the shape's inside, determines the possession of the property by  $F\Phi(H,E)$ .

#### Conclusions

- Determination of properties possessed by measures is an active research area
- The proposed visualization allows us to promptly detect distinct properties of the measures and compare them, increasing the general comprehension of the measures and helping the users choose one for their particular application

#### Conclusions

- Our proposition starts with constructing a synthetic, exhaustive and non-redundant set of contingency tables, which are commonly used to calculate the values of measures
- Using such a dataset, a 3-dimensional tetrahedron is built
- The position of points in the shape translates to corresponding contingency tables and the colour of the points represents values of the visualized measure
- Such visual-based approach is advantageous, especially when time constraints impede conducting in-depth, theoretical analyses of large numbers of such measures (e.g., generated automatically)
- Clearly, the analyses can be generalized to a wider range of measures or properties

# Thank you!

## Regular views of confirmation measures: F(H,E)





#### Visualization technique – exemplary external visualizations

- The visualization of a 'solid' tetrahedron shows only extreme values of the arguments of the visualized function (external view):
  - 2D view
  - parallelogram visualization of the net of the tetrahedron,
    i.e. a set of planar triangles, which when folded along selected edges, become the faces

## Visualization technique – external visualizations for C(H,E)



#### Visualization technique – exemplary internal visualizations

- If areas located strictly inside the tetrahedron have to be additionally visualized, various internal views can be generated:
  - 2D view
  - parallelogram visualization of the net of the tetrahedron,
    i.e. a set of planar triangles, which when folded along selected edges, become the faces

## Visualization technique – internal visualizations for C(H,E)

2D view



## Visualization technique – internal visualizations for C(H,E)

parallelograms

