

Properties of attractiveness measures for data mining – a survey

Izabela Szczęch

Poznań University of Technology

Przegląd własności miar oceny reguł

IDSS Poznań, 08.12.2009

Introduction - 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 – attractiveness (interestingness) measures (e.g. support, confidence, gain)

Introduction - motivations

The choice of interestingness measure for a certain application is a difficult task

- each measure was proposed to capture different characteristics of rules
- 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

Introduction - motivations

Properties group the measures according to similarities in their characteristics



need to analyze which properties are most desirable

Presentation plan

- Desirable properties of objective attractiveness measures
 - property of Bayesian confirmation
 - property M
 - symmetry properties
 - property Ex₁ of preserving extremes
- Critical survey on other properties in the literature
- Summary

Notation

- 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; A - set of attributes
- $S = \langle U, C, D \rangle$ *decision table*, where C set of *condition attributes*, D – set of *decision attributes*, $C \cap D = \emptyset$
- Decision rule or association rule induced from S is a consequence relation: $\phi \rightarrow \psi$ read as **if** ϕ **then** ψ where ϕ and ψ are condition and conclusion formulas built from attribute-value pairs (q, v)
- If the division into independent and dependent attributes is fixed, then rules are regarded as decision rules, otherwise as association rules.

Notation

- $a = sup(\phi \rightarrow \psi)$ is the number of objects in *U* satisfying both the premise ϕ and the conclusion ψ of a rule $\phi \rightarrow \psi$
 - $b = sup(\neg \phi \rightarrow \psi),$ $c = sup(\phi \rightarrow \neg \psi),$ $d = sup(\neg \phi \rightarrow \neg \psi)$
- $a+c=sup(\phi)$, $a+b=sup(\psi)$, $b+d=sup(\neg \phi)$, $c+d=sup(\neg \psi)$, |U|=a+b+c+d
- A 2x2 contingency table

	ψ	$\neg \psi$	
φ	а	С	a+c
$\neg \phi$	b	d	b+d
	a+b	c+d	U

Property of Bayesian confirmation

• An attractiveness $c(\phi \rightarrow \psi)$ measure has the property of confirmation if it satisfies the following condition:

$$c(\phi \rightarrow \psi) \begin{cases} > 0 & if \ Pr(\psi|\phi) > Pr(\psi) \\ = 0 & if \ Pr(\psi|\phi) = Pr(\psi) \\ < 0 & if \ Pr(\psi|\phi) < Pr(\psi) \end{cases}$$
(BC)

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

Property of Bayesian confirmation

• Under "the closed world assumption" adopted in inductive reasoning, and because U is a finite set, it is legitimate to estimate probabilities in terms of frequencies, e.g. $Pr(\psi) = \frac{a+b}{|U|}$

$$c(\phi \rightarrow \psi) \begin{cases} > 0 & if \quad \frac{a}{a+c} > \frac{a+b}{|U|} \\ = 0 & if \quad \frac{a}{a+c} = \frac{a+b}{|U|} \\ < 0 & if \quad \frac{a}{a+c} < \frac{a+b}{|U|} \end{cases}$$
(BC)

where: $a=sup(\phi \rightarrow \psi)$, $b=sup(\neg \phi \rightarrow \psi)$, $c=sup(\phi \rightarrow \neg \psi)$, $d=sup(\neg \phi \rightarrow \neg \psi)$, |U|=a+b+c+d

Rival Bayesian confirmation measures

The condition

$$c\left(\phi \rightarrow \psi\right) \begin{cases} > 0 & if \quad \frac{a}{a+c} > \frac{a+b}{|U|} \\ = 0 & if \quad \frac{a}{a+c} = \frac{a+b}{|U|} \\ < 0 & if \quad \frac{a}{a+c} < \frac{a+b}{|U|} \end{cases}$$
(BC)

does not put any constraint on the value to be assigned to confirmatory arguments (as long as they are positive) or disconfirmatory arguments (as long as they are negative)

 There are many alternative, non-equivalent measures of Bayesian confirmation with different scales

Rival Bayesian confirmation measures

- Notation: $a = sup(\phi \rightarrow \psi)$, $b = sup(\neg \phi \rightarrow \psi)$, $c = sup(\phi \rightarrow \neg \psi)$, $d = sup(\neg \phi \rightarrow \neg \psi)$
- Among popular confirmation measures there are:

$$D(\phi \rightarrow \psi) = \frac{a}{a+c} - \frac{(a+b)}{a+b+c+d}$$
$$S(\phi \rightarrow \psi) = \frac{a}{a+c} - \frac{b}{b+d}$$
$$M(\phi \rightarrow \psi) = \frac{a}{a+b} - (a+c)$$
$$N(\phi \rightarrow \psi) = \frac{a}{a+b} - \frac{c}{c+d}$$
$$C(\phi \rightarrow \psi) = \frac{a - (a+c)(a+b)}{a+b+c+d}$$
$$R(\phi \rightarrow \psi) = \frac{a(a+b+c+d)}{(a+c)(a+b)} - 1$$

(Carnap 1950/1962)

(Christensen 1999)

(Mortimer 1988)

(Nozick 1981)

(Carnap 1950/1962)

(Finch 1960)

Property M

- Property M (Greco, Pawlak, Słowiński 2004*)
- An attractiveness measure I(a, b, c, d) has the property M if it is a function
- 1. non-decreasing with respect to a and
- 2. non-increasing with respect to *b* and
- 3. non-increasing with respect to *c* and
- 4. non-decreasing with respect to *d*.

where: $a = sup(\phi \rightarrow \psi)$, $b = sup(\neg \phi \rightarrow \psi)$, $c = sup(\phi \rightarrow \neg \psi)$, $d = sup(\neg \phi \rightarrow \neg \psi)$

* Greco, S., Pawlak, Z., Słowiński, R., 2004. Can Bayesian confirmation measures be useful for rough set decision rules? Engineering Applications of Artificial Intelligence, 17: 345-361.

Interpretation of the property M

• E.g. consider rule $\phi \rightarrow \psi$:

if x is a raven then x is black

non-decreasing with respect to a

the more *black ravens* (positive examples) we observe,

the more credible becomes the rule

- non-increasing with respect to b
- non-increasing with respect to c

the more *non-black ravens* (counter examples) we observe, the less credible becomes the rule

non-decreasing with respect to d

Interpretation of the property M

- A positive value of a confirmation measure means that the rule's conclusion ψ is satisfied more frequently when the premise φ is satisfied rather than when φ is not satisfied.
 From this viewpoint we can justify the relationship between *I* and *b*, *d*:
 - objects ¬φ∧ψ (i.e. objects represented by *b*) reflect the latter situation and therefore decrease the value of confirmation.
 Thus, measure *I* should be non-increasing with respect to *b*.
 - objects ¬φ∧¬ψ (i.e. objects represented by *d*) decrease the frequency of ψ in the situations where φ is not satisfied, and therefore should increase the value of confirmation.
 Thus, measure *I* should be non-decreasing with respect to *d*

Properties of symmetry

- Properties of symmetry (Carnap 1962*, Eells & Fitelson 2002**):
 - Evidence symmetry (ES): $I(\phi \rightarrow \psi) = -I(\neg \phi \rightarrow \psi)$
 - Inversion symmetry (IS): $I(\phi \rightarrow \psi) = I(\psi \rightarrow \phi)$
 - Hypothesis symmetry (HS): $I(\phi \rightarrow \psi) = -I(\phi \rightarrow \neg \psi)$
 - Total symmetry (TS): $I(\phi \rightarrow \psi) = -I(\neg \phi \rightarrow \neg \psi)$
- Only hypothesis symmetry (HS) is desirable

HS: the impact of ϕ on ψ should be of the same strength, but of the opposite sign, as the impact of ϕ on $\neg \psi$

* 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)



Total Symmetry (TS)



Property of preserving extremes (Ex₁)

- Crupi, Tentori and Gonzalez 2007* have considered the confirmation measures from the viewpoint of classical deductive logic introducing function *v* such that for any argument (φ, ψ):
 - v assigns it the same positive value (e.g., 1)
 iff φ entails ψ, i.e. φ ↦ ψ,
 - an equivalent value of opposite sign (e.g., **-1**) iff ϕ entails the negation of ψ , i.e. $\phi \mapsto \neg \psi$, and
 - value 0, otherwise.

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

Property of preserving extremes (Ex₁)

The relationship between the logical implication or refutation of ψ by φ, and the conditional probability of ψ subject to φ requires that any Bayesian confirmation measure c(φ→ψ) agrees with v(φ,ψ) in the following sense:

(Ex₁): *if*
$$v(\phi_1 \rightarrow \psi_1) > v(\phi_2 \rightarrow \psi_2)$$
, then $c(\phi_1 \rightarrow \psi_1) > c(\phi_2 \rightarrow \psi_2)$.

1	0
1	-1
0	-1

Property of preserving extremes (Ex₁)

(Ex₁): if $v(\phi_1, \psi_1) > v(\phi_2, \psi_2)$, then $c(\phi_1 \rightarrow \psi_1) > c(\phi_2 \rightarrow \psi_2)$.



Four desirable properties

- Desirable properties of objective attractiveness measures
 - property of Bayesian confirmation
 - property M
 - property of hypothesis symmetry
 - property Ex₁ of preserving extremes

 property of evidence symmetry, inversion symmetry and total symmetry are **undesirable**

 Piatetsky-Shapiro* proposed three principles that shoud be obeyed by any objective measure, F:

(P₁) F=0 if ϕ and ψ are statistically independent, i.e. $Pr(\phi \land \psi) = Pr(\phi) Pr(\psi)$,

- (P₂) *F* monotonically increases with $Pr(\phi \land \psi)$ when $Pr(\phi)$, and $Pr(\psi)$ remain the same,
- (P₃) *F* monotonically decreases with $Pr(\phi)$ (or $Pr(\psi)$) when $Pr(\phi \land \psi)$ and $Pr(\psi)$ (or $Pr(\phi)$) remain the same.

* Piatetsky-Shapiro, G., 1991. Discovery, analysis, and presentation of strong rules. Chapter 12, in: Knowledge Discovery in Databases, AAAI/MIT Press.

- P₁: F=0 if φ and ψ are statistically independent,
 i.e. Pr(φ∧ψ)=Pr(φ) Pr(ψ)
- P₁ agrees with the "middle" condition of property of confirmation

$$c\left(\phi \rightarrow \psi\right) \begin{cases} > 0 & if \quad \frac{a}{a+c} > \frac{a+b}{|U|} \\ = 0 & if \quad \frac{a}{a+c} = \frac{a+b}{|U|} \\ < 0 & if \quad \frac{a}{a+c} < \frac{a+b}{|U|} \end{cases}$$
(BC)

Notation:

$$a = sup(\phi \rightarrow \psi)$$

 $b = sup(\neg \phi \rightarrow \psi)$
 $c = sup(\phi \rightarrow \neg \psi)$
 $d = sup(\neg \phi \rightarrow \neg \psi)$

$$Pr(\phi \land \psi) = \frac{a}{|U|}, Pr(\phi) = \frac{a+c}{|U|}, Pr(\psi) = \frac{a+b}{|U|}$$
$$Pr(\phi \land \psi) = Pr(\phi)Pr(\psi) \rightarrow aU^{2} = |U|(a+c)(a+b)$$
$$\frac{a}{a+c} = \frac{a+b}{|U|}$$

- P_2 : *F* monotonically increases with $Pr(\phi \land \psi)$
 - when Pr(φ) remains the same
 - and when $Pr(\psi)$ remains the same

Notation: $a = sup(\phi \rightarrow \psi)$ $b = sup(\neg \phi \rightarrow \psi)$ $C = sup(\phi \rightarrow \neg \psi)$ $d = sup(\neg \phi \rightarrow \neg \psi)$

Pr($\phi \land \psi$) increases while Pr(ϕ) remains unchanged when some observations ($\phi \land \neg \psi$) change into ($\phi \land \psi$), i.e. when *c* decreases and *a* increases.

Conclusion: P₂ agrees with III and I condition of property M

Pr($\phi \land \psi$) increases while Pr(ψ) remains unchanged when some observations ($\neg \phi \land \psi$) change into ($\phi \land \psi$), i.e. when *b* decreases and *a* increases.

Conclusion: P₂ agrees with II and I condition of property M

- P_3 : *F* monotonically decreases with $Pr(\phi)$ when $Pr(\phi \land \psi)$ and $Pr(\psi)$ remain the same,
- P_3 : *F* monotonically decreases with $Pr(\psi)$ when $Pr(\phi \land \psi)$ and $Pr(\phi)$ remain the same,

Notation: $a = sup(\phi \rightarrow \psi)$ $b = sup(\neg \phi \rightarrow \psi)$ $c = sup(\phi \rightarrow \neg \psi)$ $d = sup(\neg \phi \rightarrow \neg \psi)$

Pr(ϕ) increases while Pr($\phi \land \psi$) and Pr(ψ) remain unchanged when the number of observations ($\phi \land \neg \psi$) increases, i.e. when *c* increases.

Conclusion: P₃ agrees with III condition of property M

Pr(ψ) increases while Pr($\phi \land \psi$) and Pr(ϕ) remain unchanged when the number of observations ($\neg \phi \land \psi$) increases, i.e. when *b* increases.

Conclusion: P₂ agrees with II condition of property M

 Tan et al.* proposed properties based on operations for 2x2 contingency tables:

	Ψ	$\neg \psi$	
φ	а	С	a+c
φ	b	d	b+d
	a+b	c+d	U

 (O_1) F should be symmetric under variable permutation,

 (O_2) *F* should be the same when we scale any row or column by a positive factor,

(O₃) *F* should become -F if either the rows or columns are permuted, i.e. swapping either the rows or columns in the contingency table makes interestingness values change their signs,

(O₄) *F* should have no relationship with the count of the records that do not contain ϕ and ψ .

* Tan, P.-N., Kumar, V., Srivastava, J., 2002. Selecting the right interestingness measure for association patterns. In: Proc. of the 8th international Conf. on Knowledge Discovery and Data Mining (KDD 2002). Edmonton, Canada, pp.32-41.

O₁: F should be symmetric under variable permutation,

i.e. rules $\phi \rightarrow \psi$ and $\psi \rightarrow \phi$ should have the same interestingness value.

$$F(\phi \rightarrow \psi) = F(\psi \rightarrow \phi)$$

O₁ is another formulation of inversion symmetry and should be considered as an **undesirable** property

 $c(Jack \rightarrow face) \neq c(face \rightarrow Jack)$

O₂: F should be the same when we scale any row or column by a positive factor

Scaling of rows or columns effects the values of a, b, c or d. Any change of those values should be reflected by a measure. O₂ is **undesirable**.

O₃: *F* should become –*F* if either the rows or columns are permuted,
 i.e. swapping either the rows or columns in the contingency table makes interestingness values change their signs.

$$F(\phi \rightarrow \psi) = -F(\phi \rightarrow \neg \psi) = -F(\neg \phi \rightarrow \psi)$$

• $F(\phi \rightarrow \psi) = -F(\phi \rightarrow \neg \psi)$ is a hypothesis symmetry (**desirable**)

 $c(Jack \rightarrow face) = - c(Jack \rightarrow \neg face)$

100% = -(-100%)

• $F(\phi \rightarrow \psi) = -F(\neg \phi \rightarrow \psi)$ is an evidence symmetry (**undesirable**)

$$c(Jack \rightarrow face) \neq - c(\neg Jack \rightarrow face)$$

 $100\% \neq -(52-4)/_{12} * 100\%$

• O_4 : *F* should have no relationship with the count of the records that do not contain ϕ and ψ .

The number of observations that do not contain ϕ or ψ effects whether ψ is satisfied more frequently when the premise ϕ is satisfied rather than when ϕ is not satisfied. Thus, the count of records that do not contain ϕ and ψ should have a relationship with *F*.

O₄ is **undesirable**.

Properties of Lenca et al.

Lenca et al.* proposed five properties to evaluate association rules:

 (Q_1) F is constant if there is no counterexamples to the rule,

(Q₂) *F* decreases with $Pr(\phi \neg \psi)$ in a linear, concave or convex fashion around 0+,

(Q₃) *F* increases as the total number of records increases assuming that $Pr(\phi)$, $Pr(\psi)$ and $Pr(\phi \land \psi)$ are held constant,

 (Q_4) The threshold is easy to fix,

 (Q_5) The semantics of the measure are easy to express.

* Lenca, Ph., Meyer, P., Vaillant, B., Lallich, S., 2008. On selecting interestingness measures for association rules: User oriented description and multiple criteria decision aid. European Journal of Operational Research, Vol. 184, Issue 2, 610-626

Properties of Lenca et al.

Q₁: *F* is constant if there is no counterexamples to the rule
 i.e. rules with a confidence of 1 should have the same interestingness value, regardless of the support.

It is desirable that the value of F is not only constant but maximal, which agrees with property Ex_1 .

■ Q_2 : *F* decreases with Pr($\phi \land \neg \psi$) in a linear, concave or convex fashion around 0+

 $Q_{\rm 2}$ agrees with III condition of property M

* Lenca, Ph., Meyer, P., Vaillant, B., Lallich, S., 2008. On selecting interestingness measures for association rules: User oriented description and multiple criteria decision aid. European Journal of Operational Research, Vol. 184, Issue 2, 610-626

Properties of Lenca et al.

• Q_3 : *F* increases as the total number of records increases assuming that Pr (ϕ), Pr(ψ) and Pr($\phi \land \psi$) are held constant,

|U| increases while $Pr(\phi)$, $Pr(\psi)$ and $Pr(\phi \land \psi)$ remain unchanged when new observations $(\neg \phi \land \neg \psi)$ are added to the dataset, i.e. *d* increases.

Conclusion: Q₃ agrees with IV condition of property M

Q₄: The threshold is easy to fix,

 Q_5 : The semantics of the measure are easy to express.

 Q_4 and Q_5 are subjective properties.

* Lenca, Ph., Meyer, P., Vaillant, B., Lallich, S., 2008. On selecting interestingness measures for association rules: User oriented description and multiple criteria decision aid. European Journal of Operational Research, Vol. 184, Issue 2, 610-626

Summary

- Desirable properties of objective attractiveness measures
 - property of Bayesian confirmation
 - property M
 - property of hypothesis symmetry
 - property Ex₁ of preserving extremes
- Sets of properties proposed by Piatetsky-Shapiro, Tan et al., and Lenca et al. has been presented and commented showing which of them are desirable and which are concordant with the above properties.

Thank you!