

Using Analogical Proportions for Explanations

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Explanation is an old topic in AI

- We expect from an “intelligence”, even an artificial one, that it *can explain its conclusions*
- The success of *expert systems*, based on *rules*, a little over 30 years ago, had led to work to develop systems capable of explaining their conclusions
- The success of learning methods based on *neural networks* has renewed interest, over the last past years, in explanation, by raising the problem of explaining the outcome of “black box” methods

Explanations

- Explanation in *neural networks* is often seen as a problem of **sensitivity** analysis,
In the *logical* view, we distinguish
abductive explanations for “why?” questions
contrastive explanations for “why not?” questions
- Both in expert systems and in machine learning, we have the knowledge about the process that led to the conclusion to be explained:
we know the set of rules used and the classification function
- Such knowledge is no longer necessary in the approach proposed here

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Abductive explanation

- \mathcal{A} a set of n attributes $i = 1, \dots, n$
 x_i a value of attribute i
 v_i a constant in \mathcal{D}_i , domain of attribute i
 $\mathcal{D} = \mathcal{D}_1 \times \dots \times \mathcal{D}_n$
 and cl a **classification function**
- Given $cl(v) = c_0$ for $v = (v_1, \dots, v_n)$, an **abductive explanation** (by prime implicant) consists of any **minimal** subset $\mathcal{X} \subseteq \mathcal{A}$ such that

$$\forall x \in \mathcal{D}. [\bigwedge_{i \in \mathcal{X}} (x_i = v_i)] \rightarrow (cl(x) = c_0)$$
- It is enough to fix the values x_i of attributes in \mathcal{X} to v_i for insuring that $cl(x) = c_0$

Contrastive explanation

- Given $cl(v) = c_0$, a *contrastive explanation* consists of any **minimal** subset $\mathcal{Y} \subseteq \mathcal{A}$ such that

$$\exists x \in \mathcal{D}. \left[\bigwedge_{j \in \mathcal{A} \setminus \mathcal{Y}} (x_j = v_j) \right] \wedge (cl(x) \neq c_0)$$

- One can find an x , outside c_0 , which coincides with v on a **maximal** subset of attributes, i.e., one can perform a minimal change on v so that x is no longer in c_0
- This corresponds to an answer to a question “Why not $cl(v) \neq c_0$?”, i.e., one identifies the attributes whose value should be changed for that

Boolean modeling

- analogical proportion :“ a is to b as c is to d ”
 “the calf is to the cow
 as the foal is to the mare”
- $a : b :: c : d =$
 $((a \wedge \neg b) \equiv (c \wedge \neg d)) \wedge ((\neg a \wedge b) \equiv (\neg c \wedge d))$
 $0 : 0 :: 0 : 0$
 $1 : 1 :: 1 : 1$
 $0 : 1 :: 0 : 1$
 • $1 : 0 :: 1 : 0$
 $0 : 0 :: 1 : 1$
 $1 : 1 :: 0 : 0$
- nominal values
 $(a, b, c, d) \in \{(g, g, g, g), (g, h, g, h), (g, g, h, h)\}$

Example and properties

- items a, b, c, d : vectors de values of n attributes
 $a : b :: c : d$ ssi $\forall i \in \{1, \dots, n\}, a_i : b_i :: c_i : d_i$

Table: AP: example with Boolean and nominal attributes

	<i>mammal</i>	<i>carnivore</i>	<i>young</i>	<i>adult</i>	<i>family</i>
calf	1	0	1	0	bovidae
cow	1	0	0	1	bovidae
foal	1	0	1	0	equidae
mare	1	0	0	1	equidae

- $a : b :: c : d \Rightarrow a : c :: b : d$ central permutation
- $a : b :: c : d \Rightarrow c : d :: a : b$ symmetry
- $a : b :: c : d$ et $c : d :: e : f \Rightarrow a : b :: e : f$ transitivity
- $a : b :: c : d \Rightarrow \neg a : \neg b :: \neg c : \neg d$

code independence

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code independence

A reading of data oriented towards explanation

	$\mathcal{A}_1 \dots \mathcal{A}_{i-1}$	$\mathcal{A}_i \dots \mathcal{A}_{j-1}$	$\mathcal{A}_j \dots \mathcal{A}_{k-1}$	$\mathcal{A}_k \dots \mathcal{A}_{r-1}$	$\mathcal{A}_r \dots \mathcal{A}_{s-1}$	$\mathcal{A}_s \dots \mathcal{A}_n$	\mathcal{C}
<i>a</i>	1	0	1	0	1	0	<i>p</i>
<i>b</i>	1	0	1	0	0	1	<i>q</i>
<i>c</i>	1	0	0	1	1	0	<i>p</i>
<i>d</i>	1	0	0	1	0	1	<i>q</i>

- ($p \neq q$) The change of value of \mathcal{C} from p to q between a and b and between c and d can only be explained by, giving the data, the change of values of attributes from \mathcal{A}_r to \mathcal{A}_n (which is the same for the pair (a, b) and pair (c, d))
- see these pairs as instances of a rule

expressing that the change on attributes from \mathcal{A}_r to \mathcal{A}_n determines the change for \mathcal{C} whatever the context

A reading of data oriented towards explanation

	$A_1 \dots A_{i-1}$	$A_j \dots A_{j-1}$	$A_j \dots A_{k-1}$	$A_k \dots A_{r-1}$	$A_r \dots A_{s-1}$	$A_s \dots A_n$	\mathcal{C}
a	1	0	1	0	1	0	p
b	1	0	1	0	0	1	q
c	1	0	0	1	1	0	p
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- ($p \neq q$) The **change** of value of \mathcal{C} from p to q between a and b and between c and d can only be explained by, giving the data, the **change** of values of attributes from A_r to A_n

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Illustrative example

case	situation	$c. - i.$	dec.	opt. 1	opt. 2
<i>a</i>	<i>sit</i> ₁	yes	δ	0	0
<i>b</i>	<i>sit</i> ₁	no	δ	1	0
<i>c</i>	<i>sit</i> ₂	yes	δ	0	1
<i>d</i>	<i>sit</i> ₂	no	δ	1	1

- decision: serve a coffee with or without sugar

(option 1), with or without milk (option 2) to a person

What to do in *sit*₂ if no *c. i.* ?

- **question** “why milk and sugar for *d*?”

answer (for milk) “because we are in *sit*₂ (not in *sit*₁)”

“because there is no *c. i.*” for sugar

question “why no milk for *b*?”,

answer “because we are in *sit*₁ (not in *sit*₂)”

Illustrative example

case	situation	$c. - i.$	dec.	opt. 1	opt. 2
a	sit_1	yes	δ	0	0
b	sit_1	no	δ	1	0
c	sit_2	yes	δ	0	1
d	sit_2	no	δ	1	1

- decision: serve a coffee with or without sugar (option 1), with or without milk (option 2) to a person

What to do in sit_2 if no $c. i.$?

- **question** “why milk and sugar for d ?”

answer (for milk) “because we are in sit_2 (not in sit_1)”

“because there is no $c. i.$ ” for sugar

question “why no milk for b ?”,

answer “because we are in sit_1 (not in sit_2)”

Illustrative example

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question “why no milk for b ?”,

answer “because we are in sit_1 (not in sit_2)”

Analogy and contrastive explanations

case	context	change	class
<i>a</i>	<i>sit</i> ₁	<i>yes</i>	<i>p</i>
<i>b</i>	<i>sit</i> ₁	<i>no</i>	<i>q</i>
<i>c</i>	<i>sit</i> ₂	<i>yes</i>	<i>p</i>
<i>d</i>	<i>sit</i> ₂	<i>no</i>	<i>q</i>

Table: Schematic situation of analogical explanation

- The answer to the question “why *d* is not in class *p*?” relies in the values taken *d* for the attributes in *change*. When *c* is a close neighbor of *d*, the number of attributes in *change* is **small**. We are close to a **contrastive explanation** :

$$\exists x = c \in \mathcal{S}. [\bigwedge_{j \in A \setminus \text{change}} (x_j = c_j = d_j)] \wedge (cl(x) \neq q)$$

- contrastive explanation

$$\exists x \in \mathcal{D}. [\text{Disagree}(x, v) = \mathcal{Y} \wedge (cl(x) \neq c_0)]$$

Analogy and abductive explanation

- The explanation is richer here, one knows at least another pair (here (a, b)) that corresponds to another *contexte* where the same change of attribute values leads to the same change of classe, which suggests the possibility of **rules** $\forall sit$,

$$(contexte = sit) \wedge (chang. = oui) \rightarrow cl((sit, non)) = p$$

$$(contexte = sit) \wedge (chang. = non) \rightarrow cl((sit, non)) = q$$

The rules enable a reading of the Table with an **abductive** explanation flavor, which says why the item is in class p (or in class q).

abductive explanation

$$\forall x. [(Acc.(x, v) = \mathcal{X}) \rightarrow (cl(x) = c_0)]$$

Confidence in explanations

<i>case</i>	<i>context</i>	<i>change</i>	<i>class</i>
\vec{a}	<i>sit</i> ₁	<i>yes</i>	<i>p</i>
\vec{b}	<i>sit</i> ₁	<i>no</i>	<i>q</i>
\vec{c}	<i>sit</i> ₂	<i>yes</i>	<i>p</i>
\vec{d}	<i>sit</i> ₂	<i>no</i>	<i>q</i>
\vec{a}'	<i>sit'</i>	<i>yes</i>	<i>p</i>
\vec{b}'	<i>sit'</i>	<i>no</i>	<i>p</i>

- BUT **exception** if $\exists (\vec{a}', \vec{b}')$ s. t. $\vec{a}' = (\textit{sit}', \textit{yes})$, $\vec{b}' = (\textit{sit}', \textit{no})$ with $cl(\vec{a}') = cl(\vec{b}') = p$
- So we may calculate the **confidence** and **support** of the rule associated with pairs (a, b) and (c, d) in the data set

Concluding remarks - 1

- Explanatory use of analogical proportions in learning Hüllermeier (2020)
- Analogical proportions have great explanatory potential **from data**
- “*why*” and “*why not*” questions can be answered
- has been *implemented*
 - interesting to *precompile* the data set in **pairs** by identifying where items are *equal* and where and how they *differ* **to facilitate an analogical analysis of the data**
 - start by determining the **relevant** attributes,
- *confidence, support* of rules associated with pairs

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Concluding remarks - 2

- apply to preferences learning

From $a : b :: c : d$ and “ a is preferred to b ”

analogical inference concludes “ c is preferred to d ”

Analogical explanation *would also apply*

- A 2nd kind of analogical proportion where a and c on the one hand and b and d on the other hand belong to **2 different universes**:
 “*this drug is to colds what aspirin is to headache*”
 (it is quite effective and cheap)
- Analogical proportions have an explanatory value
 “*Star Wars (1977) is to Raiders of the Lost Ark (1981) as Return of the Jedi (1983) is to Indiana Jones and the Last Crusade (1989)*”

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