## Generalized Probabilistic Modus Ponens<sup>1</sup>

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## Abstract

- Modus ponens (*from A and "if A then C" infer C*) is one of the most basic inference rules.
- The probabilistic modus ponens allows for managing uncertainty by transmitting assigned uncertainties from the premises to the conclusion (i.e., from P(A) and P(C|A) infer P(C)).
- We generalize the probabilistic modus ponens by replacing A by the conditional event A|H.
- The resulting inference rule involves iterated conditionals (formalized by conditional random quantities) and propagates previsions from the premises to the conclusion.
- Interestingly, the propagation rules for the lower and the upper bounds on the conclusion of the generalized probabilistic modus ponens coincide with the respective bounds on the conclusion of the (non-nested) probabilistic modus ponens.

## Modus ponens

Modus ponens	Instantiation
A	The son gets a B
If A, then C	If the son gets a B, then the mother is angry
С	The mother is angry

We recall that, given two logically independent events A and C, the set of all coherent assessment (x, y) on  $\{A, C|A\}$  is the unit square  $[0, 1]^2$ .

	Modus Ponens	Probabilistic Modus Ponens
(Categorical prem.)	Α	P(A) = x
(Conditional prem.)	If $A$ , then $C$	P(C A) = y
(Conclusion)	С	$xy \leq P(C) \leq xy + 1 - x$

That is, the set of all coherent assessment (x, y, z) on  $\{A, C | A, C\}$  is

$$\{(x, y, z) \in [0, 1]^3 : (x, y) \in [0, 1]^2, xy \le z \le xy + 1 - x\}$$

## From Modus ponens to Generalized modus ponens

	Modus ponens	Generalized modus ponens
(Categorical premise)	A	A H
(Conditional premise)	If A, then C	If $A H$ , then C
(Conclusion)	С	С



## Generalized Probabilistic MP

We will generalize the probabilistic modus ponens by replacing the categorical premise (i.e., A) and the antecedent of the conditional premise (i.e., A in "*if* A then C") by the conditional event A|H.

Generalized Modus Ponens	Generalized Probabilistic Modus Ponens
AlH	P(A H) = x
If $A H$ , then C	$\mathbb{P}(C (A H)) = y  (\text{What is } C (A H) ?)$
С	$P(C) \in [?,?]$

What does the conditional premise (i.e., an iterated conditional) mean and how can we assess its uncertainty? What is the set of all coherent assessment (x, y) on  $\{A|H, C|(A|H)\}$ ? What is the set of all coherent assessment (x, y, z) on  $\{A|H, C|(A|H), C\}$ ?

#### Cond. probabilities as probabilities of conditional event

By using the same symbols to denote events and their indicators, agreeing to the betting metaphor of the coherence framework, if you assess p = P(A|H), then coherence requires that  $p = \mathbb{P}(AH + p\overline{H})$ . Thus, we identify the conditional event A|H as the following random quantity (see, e.g., Lad, 1996; Gilio & Sanfilippo, 2014)

$$A|H = AH + p\overline{H} = \begin{cases} 1, & \text{if } AH \text{ is true,} \\ 0, & \text{if } \overline{A}H \text{ is true,} \\ p, & \text{if } \overline{H} \text{ is true,} \end{cases}$$

Given a random quantity X and an event H, we identify

$$X|H = XH + \mu \overline{H}$$
, where  $\mu = \mathbb{P}(X|H)$ .

## Conjoined conditionals

#### Definition (Conjunction)

Given any pair of conditional events A|H and B|K, with P(A|H) = x, P(B|K) = y, we define their *conjunction* as the following conditional random quantity (see, e.g., Gilio & Sanfilippo, 2013a, 2013b, 2014)

$$(A|H) \land (B|K) = \min\{A|H, B|K\} | (H \lor K) = \min\{AH + x\overline{H}, BK + y\overline{K}\} | (H \lor K).$$

Of course  $(A|H) \land (B|H) = (A \land B)|H$ . Notice that, if x = y = 1, then the conditional events  $A|H = AH + x\overline{H}, B|K = BK + y\overline{K}$  coincide with the material conditionals  $AH + \overline{H}, BK + \overline{K}$  and we recovery the quasi conjunction of Adams:

 $min\{AH + \overline{H}, BK + \overline{K}\}|(H \lor K) = QC(A|H, B|K)$ 

### Interpretation by the betting scheme

By assessing  $\mathbb{P}[(A|H) \land (B|K)] = z$ , you agree to pay z with the proviso that you will receive:

$$(A|H) \land (B|K) = \begin{cases} 1, & \text{if } AHBK \text{ is true,} \\ 0, & \text{if } \overline{A}H \text{ is true or } \overline{B}K \text{ is true,} \\ x = P(A|H), & \text{if } \overline{H}BK \text{ is true,} \\ y = P(B|K), & \text{if } \overline{K}AH \text{ is true,} \\ z = \mathbb{P}[(A|H) \land (B|K)], & \text{if } \overline{H}\overline{K} \text{ is true.} \end{cases}$$

In other words, you will receive:

- 1, if both conditional events are true;
- 0, if at least one of the conditional events is false;
- the probability of that conditional event which is void, if a conditional event is void and the other one is true;
- the quantity that you paid, if both conditional events are void.

Fréchet-Hoeffding bounds (Gilio & Sanfilippo, 2014): z is coherent iff:

$$max\{x+y-1,0\} \le z \le min\{x,y\}.$$
 (1)

## Quasi-Conjunction and Conjunction

Let A, H, B, K be log. ind. events, with  $H \neq \bot, K \neq \bot$ . Assuming that x = P(A|H) and y = P(B|K), we obtain

		A H	B K	$\mathcal{QC}(A H,B K)$	$A H \wedge B K$
$C_1 = AHBK$	$\Rightarrow$	1	1	1	1
$C_2 = AHB^c K$	$\Rightarrow$	1	0	0	0
$C_3 = A^c HBK$	$\Rightarrow$	0	1	0	0
$C_4 = A^c H B^c K$	$\Rightarrow$	0	0	0	0
$C_5 = H^c \mathbf{B} \mathbf{K}$	$\Rightarrow$	x	1	1	X
$C_6 = AHK^c$	$\Rightarrow$	1	У	1	У
$C_7 = \mathbf{A}^c \mathbf{H} \mathbf{K}^c$	$\Rightarrow$	0	У	0	0
$C_8 = H^c B^c K$	$\Rightarrow$	x	0	0	0
$C_9 = H^c K^c$	$\Rightarrow$	х	y	$\nu = \mathbb{P}(QC(A H,B K))$	$z = \mathbb{P}(A H \wedge B K)$

QC bounds:  $\nu$  is coherent iff  $\max\{x + y - 1, 0\} \le \nu \le S_0^H(x, y)$ where  $S_0^H(x, y) = \begin{cases} \frac{x - y - 2xy}{1 - xy}, & (x, y) \ne (1, 1), \\ 1, & (x, y) = (1, 1) \end{cases}$ 

## Iterated conditional and betting scheme

#### Definition

The iterated conditional (B|K)|(A|H) is the conditional random quantity

$$(B|K)|(A|H) = (B|K) \wedge (A|H) + \mu \cdot \overline{A}|H, \qquad (2)$$

where  $\mu = \mathbb{P}[(B|K)|(A|H)]$ . Notice that (2) is a generalization of  $A|H = A \land H + p \cdot \overline{H}$ , where p = P(A|H).

In the context of betting scheme  $\mu$  represents the amount you agree to pay with the proviso that you will receive the quantity

$$(B|K)|(A|H) = \begin{cases} 1, & \text{if } AHBK \text{ true,} \\ 0, & \text{if } AH\overline{B}K \text{ true,} \\ y, & \text{if } \overline{A}H\overline{K} \text{ true,} \\ \mu, & \text{if } \overline{A}H \text{ true,} \\ x + \mu(1-x), & \text{if } \overline{H}BK \text{ true,} \\ \mu(1-x), & \text{if } \overline{HB}K \text{ true,} \\ z + \mu(1-x), & \text{if } \overline{HK} \text{ true.} \end{cases}$$

Coherence requires that  $z + \mu(1-x) = \mu \in [0,1]$ .

#### Some properties

The product rule: (Gilio & Sanfilippo, 2014)

$$\mathbb{P}[(B|K) \wedge (A|H)] = \mathbb{P}[(B|K)|(A|H)] \cdot P(A|H).$$
(3)

Moreover, assuming x = P(A|H) > 0, one has:

$$\mathbb{P}[(B|K)|(A|H)] = \mu = \frac{\mathbb{P}[(B|K) \land (A|H)]}{P(A|H)} = \frac{z}{x}.$$

Decomposition formula

$$B|K = (A|H) \wedge (B|K) + (\overline{A}|H) \wedge (B|K).$$
(4)

By the linearity of prevision, and by the product rule, we obtain

 $P(B|K) = \mathbb{P}[(B|K)|(A|H)]P(A|H) + \mathbb{P}[(B|K)|(\overline{A}|H)]P(\overline{A}|H).$ 

#### A particular case

If  $K = \Omega$  (by replacing *B* with *C*), then we obtain from the decomposition formula:

$$C = (A|H) \wedge C + (\overline{A}|H) \wedge C, \qquad (5)$$

$$P(C) = \mathbb{P}[C|(A|H)]P(A|H) + \mathbb{P}[C|(\overline{A}|H)]P(\overline{A}|H)$$
(6)

By applying Definition 2, with  $K = \Omega$  and by replacing B with C, we obtain  $C|(A|H) = C \land (A|H) + \mu \overline{A}|H$ . That is,

$$C|(A|H) = \begin{cases} 1, & \text{if } AHC \text{ true,} \\ 0, & \text{if } \overline{AHC} \text{ true,} \\ \mu, & \text{if } \overline{AH} \text{ true,} \\ x + \mu(1-x), & \text{if } \overline{HC} \text{ true,} \\ \mu(1-x), & \text{if } \overline{HC} \text{ true,} \end{cases}$$

Notice that  $C|(A|H) \neq C|AH$ .

#### Coherent sets and coherent extensions

#### Theorem (2)

Let three logically independent events A, C, H be given, with  $A \neq \bot$ ,  $H \neq \bot$ . The set of all coherent assessments  $\mathcal{M} = (x, y, z)$  on  $\mathcal{F} = \{A|H, C|(A|H), C|(\overline{A}|H)\}$  is the unit cube  $[0,1]^3$ .

That is, assuming logical independence, every point  $(x, y, z) \in [0, 1]^3$  is a coherent assessment on  $\{A|H, C|(A|H), C|(\overline{A}|H)\}$ .

#### Theorem (3)

Given any coherent assessment (x, y) on  $\{A|H, C|(A|H)\}$ , with A, C, Hlogically independent, but  $A \neq \bot$  and  $H \neq \bot$ , the extension z = P(C) is coherent if and only if  $z \in [z', z'']$ , where

$$z' = xy$$
 and  $z'' = xy + 1 - x$ . (7)

That is, we generalized the probabilistic *modus ponens* to the case where the first premise A is replaced by the conditional event A|H.

## Proof

From P(A|H) = x and  $\mathbb{P}[C|(A|H)] = y$  infer  $xy \le P(C) \le xy + 1 - x$ From (6), by the linearity of prevision, and the product rule, we obtain

$$z = P(C) = \underbrace{P(A|H)}_{x} \underbrace{\mathbb{P}[C|(A|H)]}_{y} + \underbrace{P(\overline{A}|H)}_{1-x} \underbrace{\mathbb{P}[C|(\overline{A}|H)]}_{t \in [0,1]}$$

From Theorem 3, given any coherent assessment (x, y) on  $\{A|H, C|(A|H)\}$ , the extension  $t = \mathbb{P}[C|(\overline{A}|H)]$  on  $C|(\overline{A}|H)$  is coherent for every  $t \in [0,1]$ . As z = xy + (1-x)t, it follows that

$$\underbrace{xy}_{if \ t=0} \leq P(C) \leq \underbrace{xy + (1-x)}_{if \ t=1}$$

Prevision entailment:

 $P(A|H) = 1, \& \mathbb{P}[C|(A|H)] = 1 \Longrightarrow P(C) = 1$ 

## Conclusion

- We generalized the probabilistic modus ponens in terms of conditional random quantities in the setting of coherence.
- Specifically, we replaced the categorical premise A and the antecedent A of the conditional premise C|A by the conditional event A|H.
- We proved a generalized decomposition formula for conditional events and we gave some results.
- We propagated the previsions from the premises of the generalized probabilistic modus ponens to the conclusion.
- We have shown that the lower and the upper bounds on the conclusion of the generalized probabilistic modus ponens coincide with the respective bounds on the conclusion for the (non-nested) probabilistic modus ponens.

### Further work

- We will study other instantiations to obtain further generalizations of modus ponens, e.g., by also replacing the consequent C of the conditional premise C|A and the conclusion C by a conditional event C|K: from {A|H, (C|K)|(A|H)} infer C|K.
- We will focus on similar generalizations (also involving imprecision) of other argument forms like the probabilistic modus tollens.
- We will study similar generalization of inference rules in System P where a conditional will be replaced by an iterated one.

# Thank you for your attention!

Link to the full paper

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