

Vagueness at all orders

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Outline of the talk

Non-transitivity, Vagueness and Iterations

Logical ways around

Connection with Signal Detection Theory

Joint work with Paul Égré



“Vagueness, Uncertainty and Degrees of Clarity”
Synthese, 2009.

Non-transitivity and Vagueness

Intransitivity of indiscriminability

It may be the case that a sample A is indiscriminable in colour from a sample B, B is indiscriminable from C, and still A and C are not indiscriminable.

Vagueness may arise whenever there is non-transitivity

If A is categorized as 'red' and C as 'not-red', B seems destined to be a borderline case.

This holds of *direct discriminability*, one may put into question whether it applies to *indirect discriminability* as well.

Non-transitivity at work

- ▶ *Induction premiss in sorites paradox*
 $Px \wedge xRx' \rightarrow Px'$
- ▶ *Categorization effects in forced march*
I am more likely to categorize
some not so X thing as being X if,
starting from a clearly X thing,
I have been gradually led to do so.



Margins for errors (Williamson)

An operator \square obeys a margin for error iff,
If $w \models \square p$ and w is indiscriminable from w' then $w' \models p$.

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 \Box is 'The subject knows that' or 'it is clear that', etc.

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Williamson makes a further move and suggests to give a
semantics for \square by turning the *if* into an *iff*.

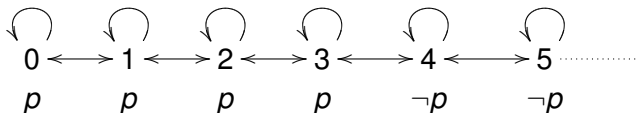
Margin for error semantics

A *fixed-margin* model is a 4-uplet $\langle W, d, \alpha, V \rangle$, where d is a metric over W , and α a real valued margin for error parameter.

$w \models \Box\phi$ iff for every w' such that $d(w, w') \leq \alpha$, $w' \models \phi$.

At a given world, the subject knows / it is clear that some proposition ϕ holds iff ϕ holds at every world included within the margin α from w .

A discrimination task

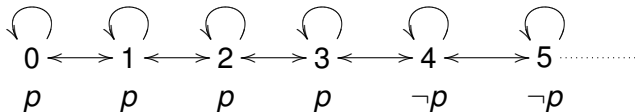


A series of pens, such that all and only pens that are less than 4 cm fit in a certain box.

A subject sees the pens and the box at a certain distance and is asked which pens will fit in the box.

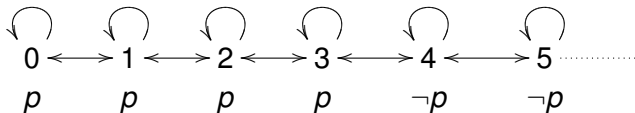
The subject can perceptually discriminate only between pens whose size differs by more than 1 cm. non-adjacent pens.

Inexact Knowledge



- ▶ $3 \models p$
- ▶ $3 \not\models \Box p$
- ▶ $3 \models \neg \Box p \wedge \neg \Box \neg p$
- ▶ $2 \models \Box p$

Vagueness spreads out



If \square is interpreted as 'the subject knows that' :
Introspection principles are not valid.

$$2 \models \square p \text{ but } 2 \not\models \square \square p$$

If \square is interpreted as 'it is clear that' :
we get higher-order vagueness.

$$2 \not\models \neg \square \square p \text{ et } 2 \not\models \square \neg \square p$$

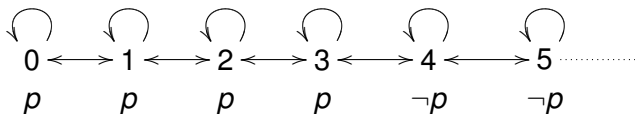
Resisting the Spread of Vagueness

- ▶ Iterations of Knowledge / Definiteness as ‘a process of gradual erosion’

vs

- ▶ KK as a rationality principle,
- ▶ Limited higher-order vagueness.

The Hidden Assumption



- ▶ 'Williamsonian' view

in order to determine whether $2 \models \Box\Box p$, alternatives to 3 which are *not* alternatives to 2 come into play.

- ▶ Supervenience view

in order to determine whether $2 \models (\Box\Diamond)p$, only alternatives to 2 should come into play.

Centered Semantics

Truth is first defined for couple of worlds :

- (i) $\mathcal{M}, (w, w') \models_{CS} p$ iff $w' \in V(p)$
- (ii) $\mathcal{M}, (w, w') \models_{CS} \neg\phi$ iff $\mathcal{M}, (w, w') \not\models_{CS} \phi$
- (iii) $\mathcal{M}, (w, w') \models_{CS} (\phi \wedge \psi)$
iff $\mathcal{M}, (w, w') \models_{CS} \phi$ and $\mathcal{M}, (w, w') \models_{CS} \psi$
- (iv) $\mathcal{M}, (w, w') \models_{CS} \Box\phi$
iff for all w'' such that wRw'' , $\mathcal{M}, (w, w'') \models_{CS} \phi$

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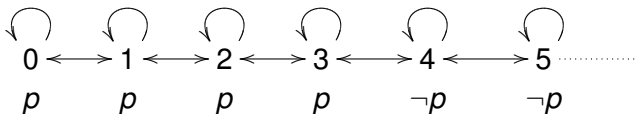
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iff for all w'' such that wRw'' , $\mathcal{M}, (w, w'') \models_{CS} \phi$

Then we set $\mathcal{M}, w \models_{CS} \phi$ iff $\mathcal{M}, (w, w) \models \phi$

Inexact Knowledge



- ▶ $(3, 3) \not\models_{CS} \Box p$
 because $(3, 4) \not\models_{CS} p$
- ▶ $(2, 2) \models_{CS} \Box p$
 because $(2, 3) \models_{CS} p$ and so do $(2, 2)$ and $(2, 1)$
- ▶ $(2, 2) \models_{CS} \Box \Box p$ because $(2, 3) \models_{CS} \Box p$ and so do $(2, 2)$ and $(2, 1)$

Axioms for CS

Fact

CS validity is axiomatized by K45

Fact

CS validity over reflexive frames is axiomatized by S5

Knowledge vs Clarity

Iteration patterns might differ for \square as ‘The subject knows that’ and \square as ‘It is clear that’.

- ▶ accept KK
- ▶ reject higher-order vagueness *beyond some level*

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Generalizing *CS* yields a family of resource-sensitive logics, *TS*(n) for n *Token Semantics*, where n bounds the number of allowed context shifts. In *TS*(n), iterations of \square or \diamond beyond n are trivialized.

- ▶ *CS* is *TS*(1),
- ▶ *ML* is *TS*(ω),
- ▶ A logic of clarity might be provided by *TS*(n) for $1 < n < \omega$

Halpern's solution

We have changed the semantics. An alternative would be to change the models.

Halpern (2004) : Non-transitivity is the product of two more fundamental transitive 'similarity' relations.

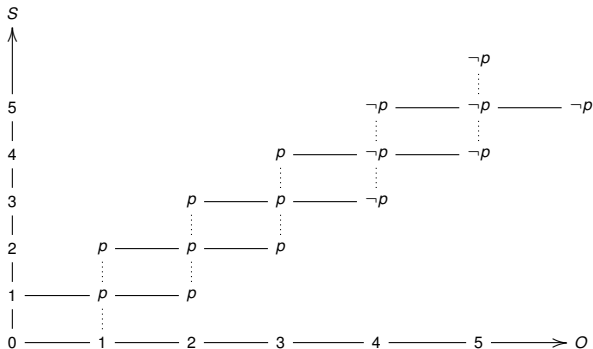
- ▶ a bidimensional modeling : external states and inner states,
- ▶ a relation \sim_s describes the agent's uncertainty about the target value, given her subjective state.
- ▶ a relation \sim_o which describes the fluctuation of the agent's estimate around the target value.

A Halpern model is a structure $\langle W, \sim_s, \sim_o, V \rangle$, with $W \subseteq O \times S$ where S is a set of subjective states and O a set of objective states.

and

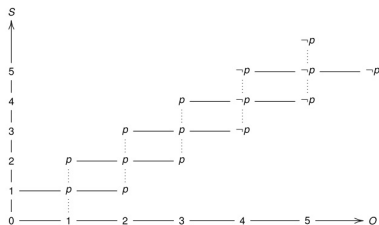
- ▶ \sim_s is an equivalence relation btw pairs with the same subjective index,
and interprets an operator D 'it is definitely the case that'.
- ▶ \sim_o is an equivalence relation btw pairs with the same objective index,
and interprets an operator R 'the agent reports that'.

Back to the discrimination task

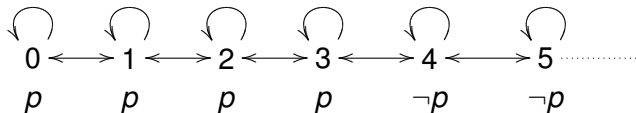


- ▶ $(3, 3) \vdash \neg Dp$
- ▶ $(2, 2) \vDash Dp$ and $(2, 2) \vDash DDp$,
- ▶ $(2, 2) \vDash \neg DRDRp$.

Halpern semantics and Centered Semantics



is obtained from



by dividing the structure into spheres of radius 1.

Layering

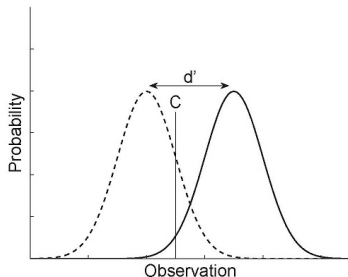
Given a Kripke model $M = \langle W, R, V \rangle$, let $L(M) = \langle W', R', V' \rangle$ be the *layering* of M defined by :

- ▶ $W' = \{(w, w') \in W \times W / w' R w \vee w' = w\}$
- ▶ $(w, w') R'(u, u')$ iff $w' = u'$ and $w' R u$; and finally,
 $(w, w') \in V'(p)$ iff $w \in V(p)$

Fact

$M, w \models_{\text{CS}} \phi \iff L(M), (w, w) \models \phi$ (for $\Box = D$).

Signal Detection Theory



Signal Detection Theory

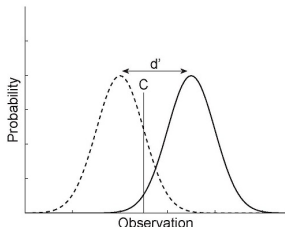
A statistical model, which is commonly used in psychophysics to account for sensory processes, and more generally to account for decisions in uncertainty.

A typical SDT plot

- ▶ dashed curve represents $P(x|n)$.
- ▶ continuous curve represents $P(x|s)$.

There is uncertainty when the probability distributions overlap. The smaller d' , the more uncertainty there is.

C represents some decision criterion.
eg : categorize as s when $\frac{P(x|s)}{P(x|n)} > 1$.



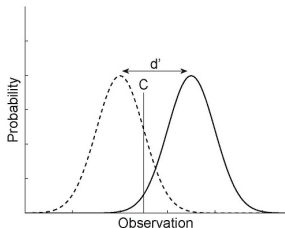
Prior and posterior probabilities

Prior probabilities of the form $P(x|A)$ constitute a quantitative version of \sim_o .

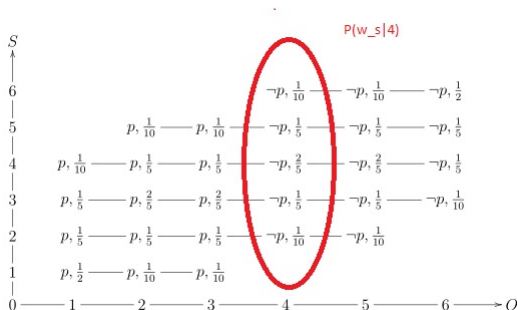
Posterior probabilities of the form $P(A|x)$ constitute a quantitative version of \sim_s .

Posterior probabilities can be computed from prior probabilities and probabilities for the objective states.

$$P(s|x) = \frac{P(s) \times P(x|s)}{P(s) \times P(x|s) + P(n) \times P(x|n)}$$

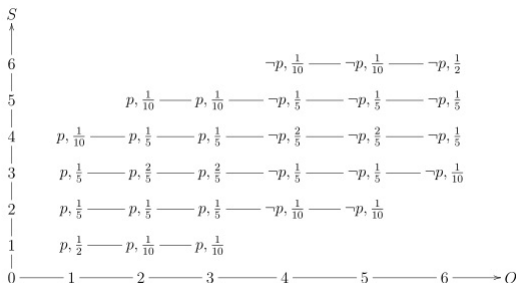


Back to the discrimination task

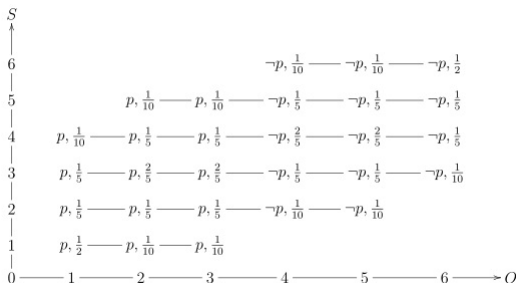


where the number in cell (x, y) is $P(x|y)$.

Deriving conditional probabilities for objective states



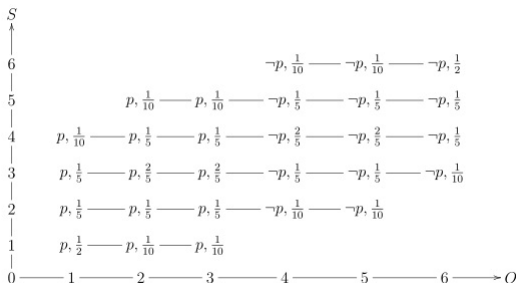
Deriving conditional probabilities for objective states



Posterior probabilities :

$$P(w_o|w_s) = \frac{P(w_s|w_o)}{\sum_y P(w_s|y)}$$

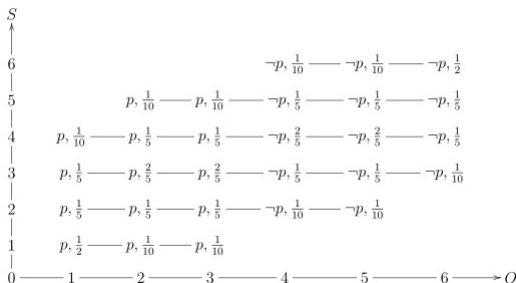
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Posterior probabilities :

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→ for every $w \in S$, we get

a probability measure Π_w on $\{(y, w_s) / y \in O\}$,

From quantities to qualities

How can we define $(w, w') \models \Box\phi$ from $\Pi_{w'}$?

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(well-known) problem :

If we want $\Box p \wedge \Box q \rightarrow \Box(p \wedge q)$

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Idea (Skyrms, Leitgeb, etc.) :

Require that $\Pi_{w'}(\phi)$ be robustly high.

Skyrm's solution

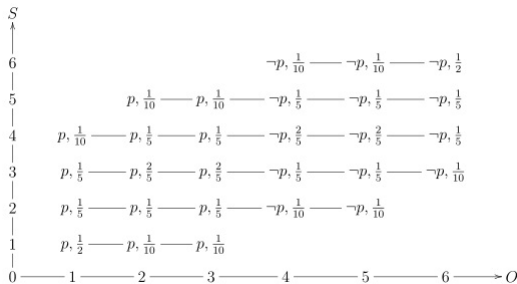
Definition (resistance)

Let the resistance $R(\phi)$ of ϕ be $Min(Pr(\phi|\psi))$ where ψ is compatible with ϕ .

Theorem (Skyrms, 1980)

The set of formulas whose resistance is greater than α for $\alpha > 0.5$ is coherent and closed under conjunction.

Back to the discrimination task



Take $\alpha = 0,5$,

- ▶ $(2,3) \models \square p$
- ▶ $(4,4) \not\models \square p$

Iterations with probabilities

In order to say whether $(w, w') \models \Box\Box\phi$
one need to determine whether $(x, w') \models \Box\phi$
for all x such that $P_{w'}(\{x\}) \neq 0$

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as before, they are several options

- ▶ similar to Centered Semantics : supervenience
- ▶ similar to Williamson's : context-shift
- ▶ similar to Token Semantics : limited context-shift

Iterations with probabilities (cont.)

How to determine whether $(x, w') \models \Box\phi$
for all x such that $P_{w'}(\{x\}) \neq 0$?

- ▶ supervenience :
To check whether $(x, w') \models \Box\phi$
keep using $P_{w'}$.
- ▶ context-shift
Whether $(x, w') \models \Box\phi$
will depend on P_y for the y s.t. $P(y|x) \neq 0$

- ▶ context-shift up to some point

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Whether $(x, w') \models \Box\phi$
will depend on P_y for the y s.t. $P(y|x) \neq 0$
but depend how ?

- ▶ require $\Box\phi$ wrt all such P_y
 - ▶ require $\Box\phi$ with a (stably) high probability
- ▶ context-shift up to some point

Conclusion

- ▶ Indiscriminability may be at the heart of vagueness, but one cannot jump from facts about indiscriminability to semantic facts about knowledge / clarity,
- ▶ non-transitivity need not imply failure of KK / higher-order vagueness, and non-transitivity may be factored as a product of transitive relations,
- ▶ probabilistic models in psychophysics provide a natural background for a quantitative approach to the analysis of (perceptual) belief / clarity.