

# **Interfacing Probabilistic and Epistemic Update**

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## **Plan**

- (apologize for changing the topic)
- (apologize for not being a probability theorist)
- Try to see if we can import tools from Dynamic Epistemic Logic to Probability Theory.
- Specifically: use Baltag-Moss-Solecki style action models to represent probabilistic evidence.

## Epistemic and Probabilistic update

information state  $\xrightleftharpoons{\text{update}}$  new state

*epistemic update with  $\phi$*

eliminate non- $\phi$  worlds  
Set of worlds  $\xrightleftharpoons{\hspace{1.5cm}}$  Set of worlds

epistemic update generalizes to: *Bayesian conditioning*

Probability measure  $\xrightleftharpoons{\text{Condition on } \phi}$  Probability measure

## Action models

information state  $\xrightleftharpoons{\text{update}}$  new state

*epistemic update à la Baltag-Moss-Solecki:*

product with action model  
Set of worlds  $\xrightleftharpoons{\hspace{1cm}}$  Set of worlds

epistemic update generalizes to: ??

?? probabilistic action model  
Probability measure  $\xrightleftharpoons{\hspace{1cm}}$  Probability measure

## Running Example

Johnny has stolen a cookie.

You ask him: 'Did you steal the cookie?'

Johnny mumbles something — you heard 'no'; but you are not sure.

Anyway, he might be lying about it.

## Simplified Action models

(for the probability people: *Event Models*)

- An action model is a set of *actions*
- The agent is not sure which action from the set has happened
- Each action has a precondition
- An update of an information state with an action models consists of all pairs  $(w, a)$  such that the preconditions of  $a$  are fulfilled in

$w$

**information state**

$p$   
 $\neg p$

**action model**

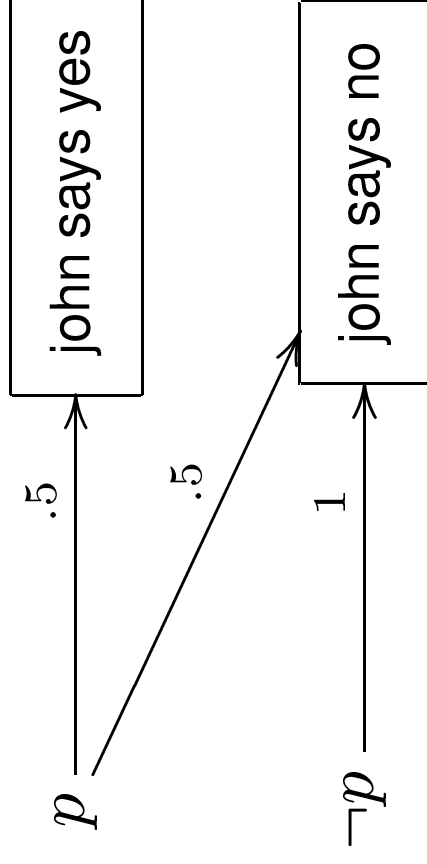
johnny says yes <sup>$p$</sup>   
johnny says no <sup>$\neg p$</sup>

**result**

$(p, \text{says yes})$   
 $(p, \text{says no})$   
 $(\neg p, \text{says no})$

## Probabilistic action models

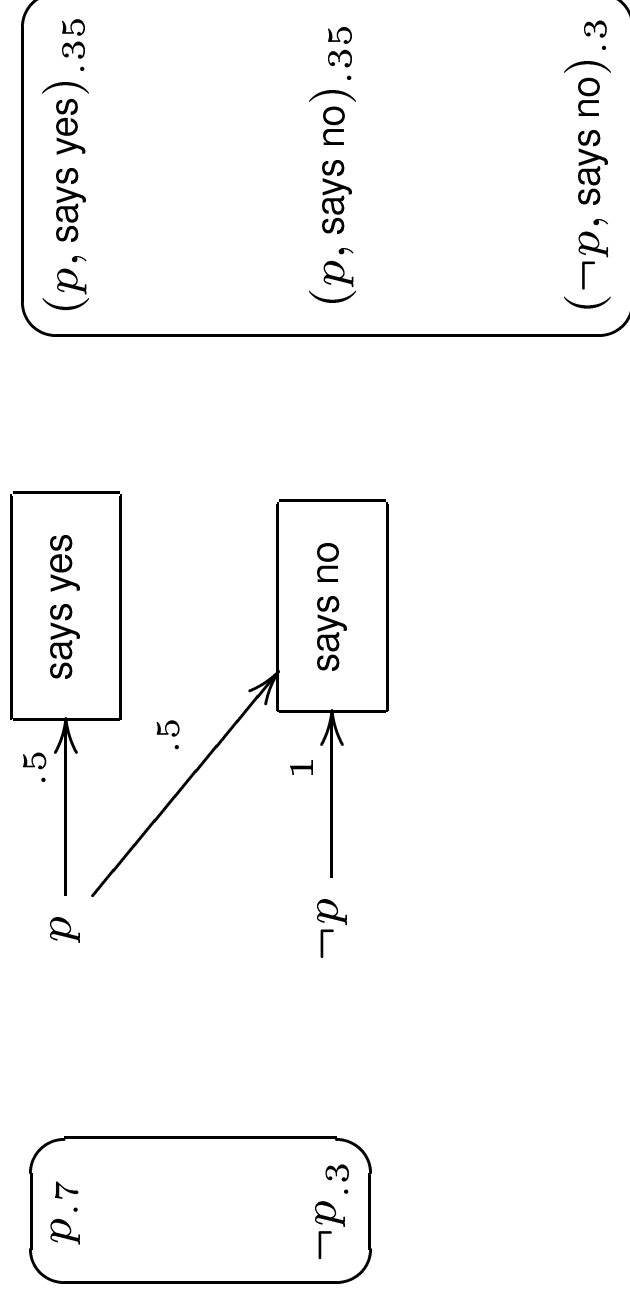
- Add (conditional) probabilities to the actions (van Benthem, JoLLI, 2003)





## Updates with Probabilistic Action Models

- Update as if probabilities represent Bayes' factors

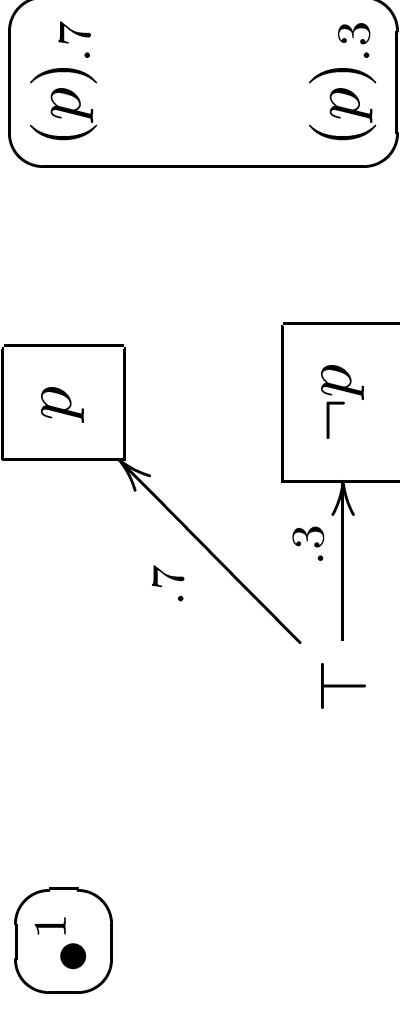


## **What Probabilistic Action models might give**

- Explicit representation of ‘internal structure’ of probabilistic evidence
- A logic of probabilistic update
- ‘Building rich models in a structured way’
- (Probabilistic updates for models with more agents)

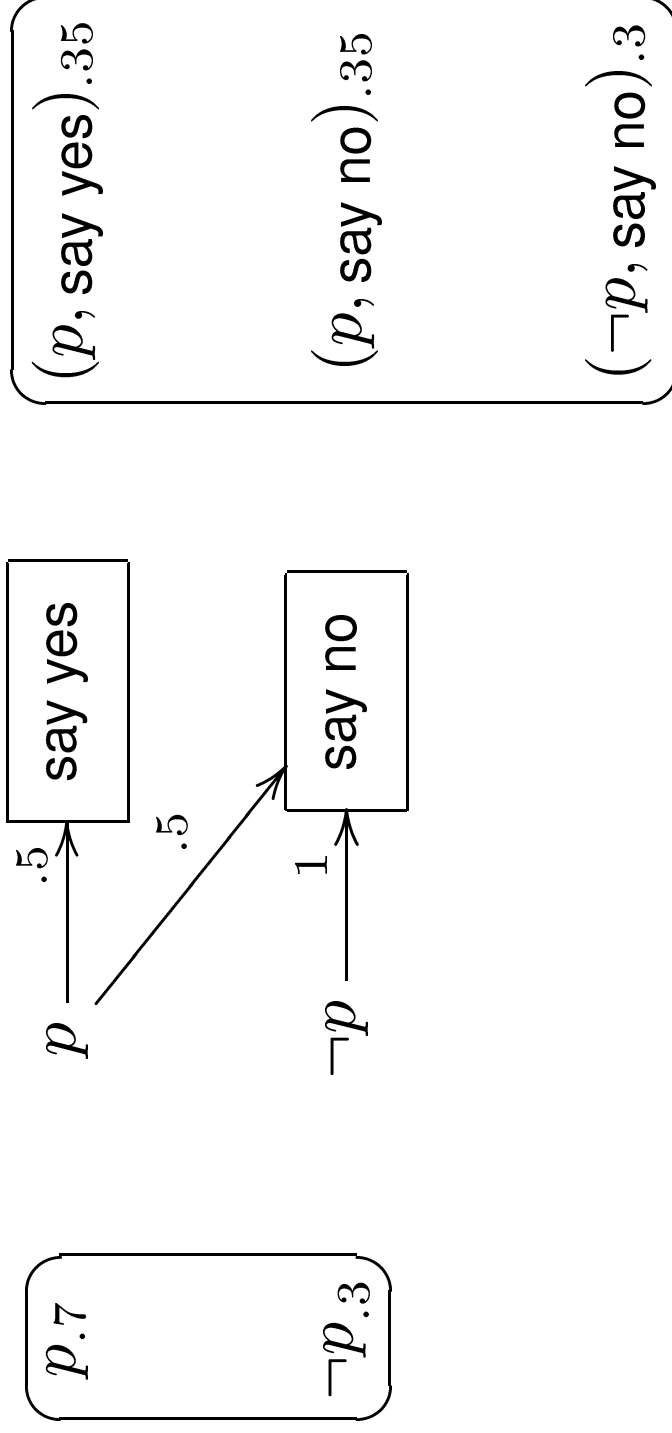
## Johnny and the cookie

You learn that Johnny might have stolen a cookie (probability .7)



# Johnny and the cookie

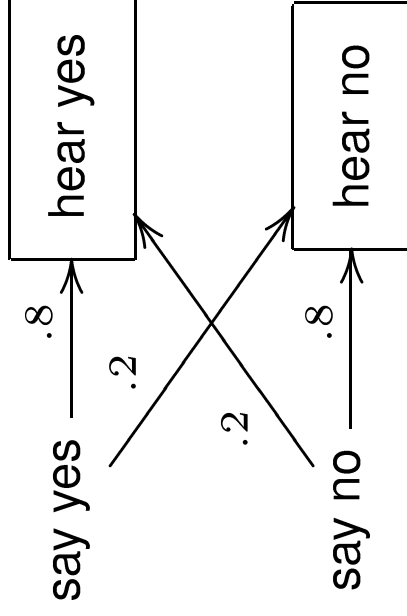
You learn that Johnny is a liar



# Johnny and the cookie

You learn that Johnny mumbles

$(p, \text{say yes}) .35$   
 $(p, \text{say no}) .35$   
 $(\neg p, \text{say no}) .3$

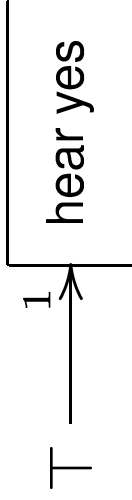


$(p, \text{say yes, hear yes}) .28$   
 $(p, \text{say yes, hear no}) .07$   
 $(p, \text{say no, hear yes}) .07$   
 $(p, \text{say no, hear no}) .28$   
 $(\neg p, \text{say no, hear yes}) .06$   
 $(\neg p, \text{say no, hear no}) .24$

# Johnny and the cookie

You hear Johnny say 'yes'

$(p, \text{ say yes, hear yes})$	.28
$(p, \text{ say yes, hear no})$	.07
$(p, \text{ say no, hear yes})$	.07
$(p, \text{ say no, hear no})$	.28
$(\neg p, \text{ say no, hear yes})$	.06
$(\neg p, \text{ say no, hear no})$	.24



$(p, \text{ say yes, hear yes})$	$\frac{28}{41}$
$(p, \text{ say no, hear yes})$	$\frac{7}{41}$
$(\neg p, \text{ say no, hear yes})$	$\frac{6}{41}$

## What we like about this

- Modular
  - Scenarios can be decomposed into more or less independent action models
  - Action models can be composed to make more complex ones (and the order in which we do this does not matter)
- Structured
  - Create a *logic of probabilistic evidence*
  - $P(p \mid \phi) = \alpha \cdot P(q \mid \phi) = \beta \rightarrow P(p \wedge q \mid \phi) = \alpha \cdot \beta$

## **What we would like to know**

- Does this make sense to a probability theorist?
- Does this seem useful to a probability theorist?