

# Belief bisimulation for hidden Markov models

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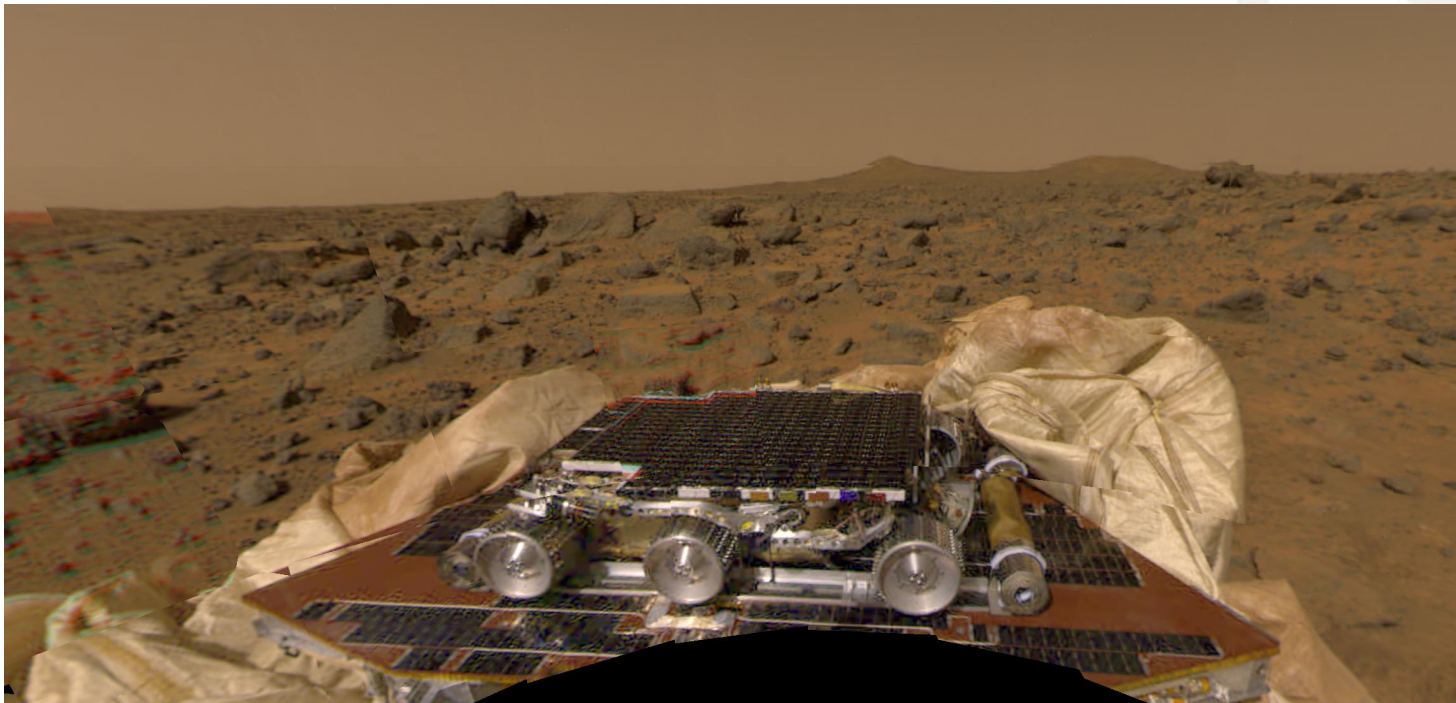
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## NASA's current Mars rover...

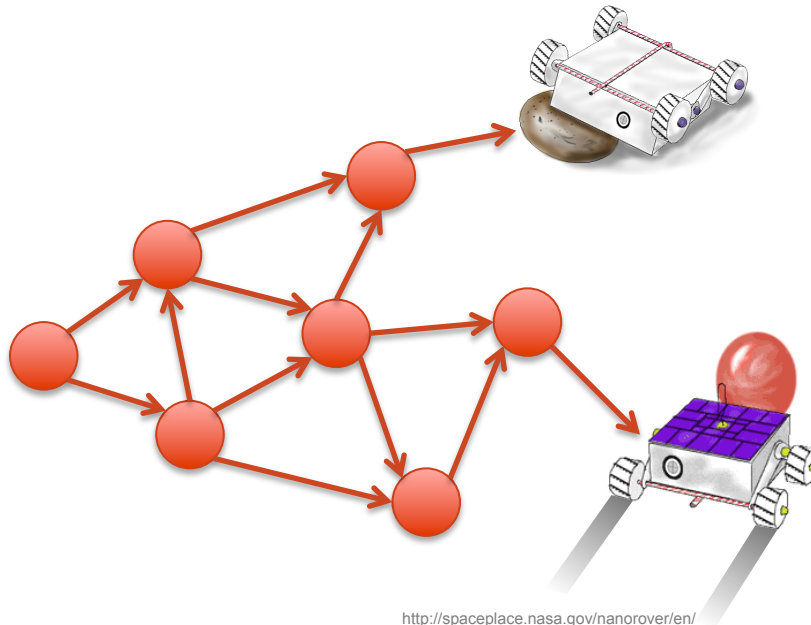
- How does a Mars rover find its position and orientation?
- Current: camera image sent to Earth, judged by humans



<http://mars.jpl.nasa.gov/MPF/ops/sol2.html>

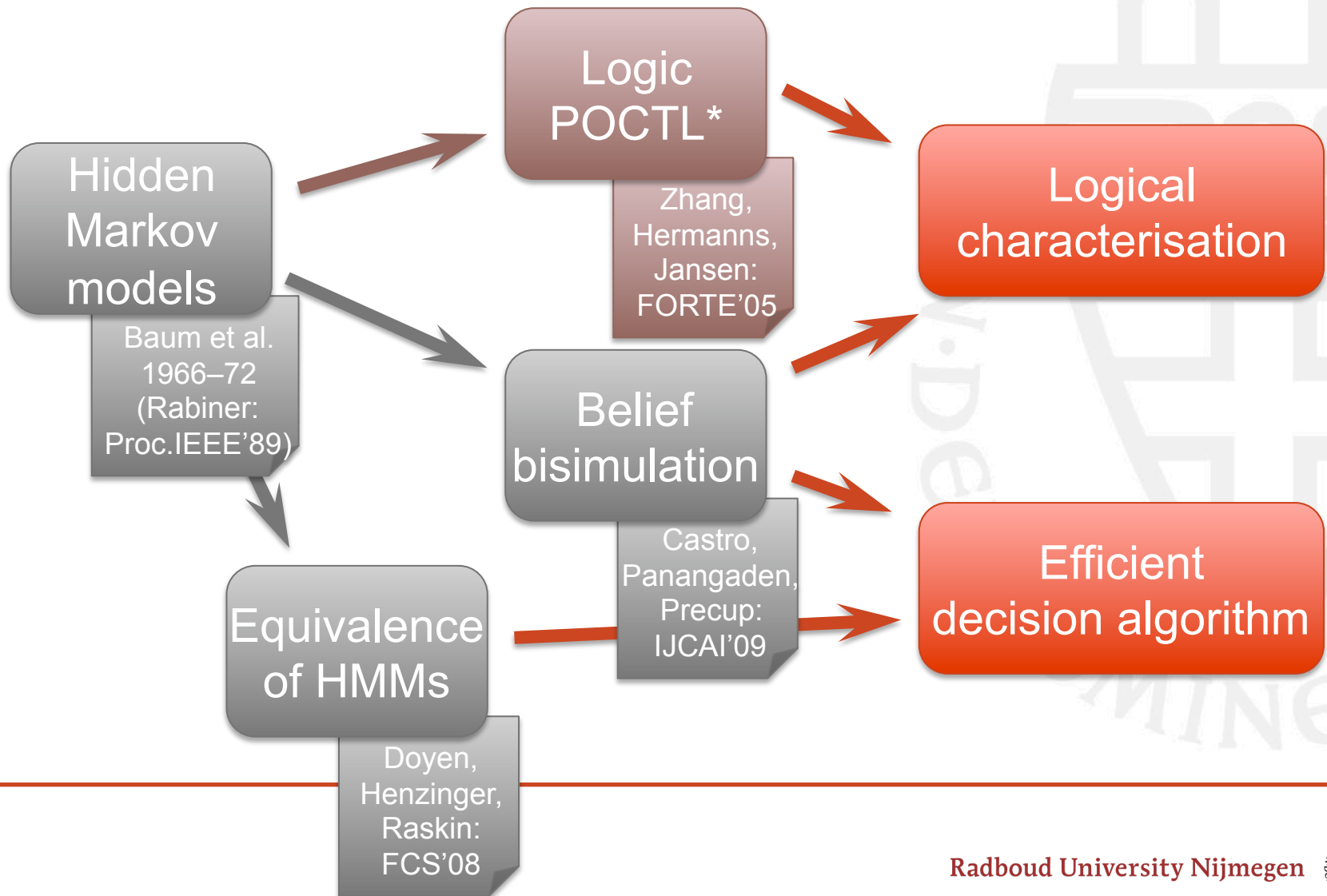
## NASA's next Mars rover...

- How does a Mars rover find its position and orientation?
- Current: camera image sent to Earth, judged by humans
- Future possibility: use sensors and infer state from measurements
- **Given a sequence of measurements, what is the current state?**

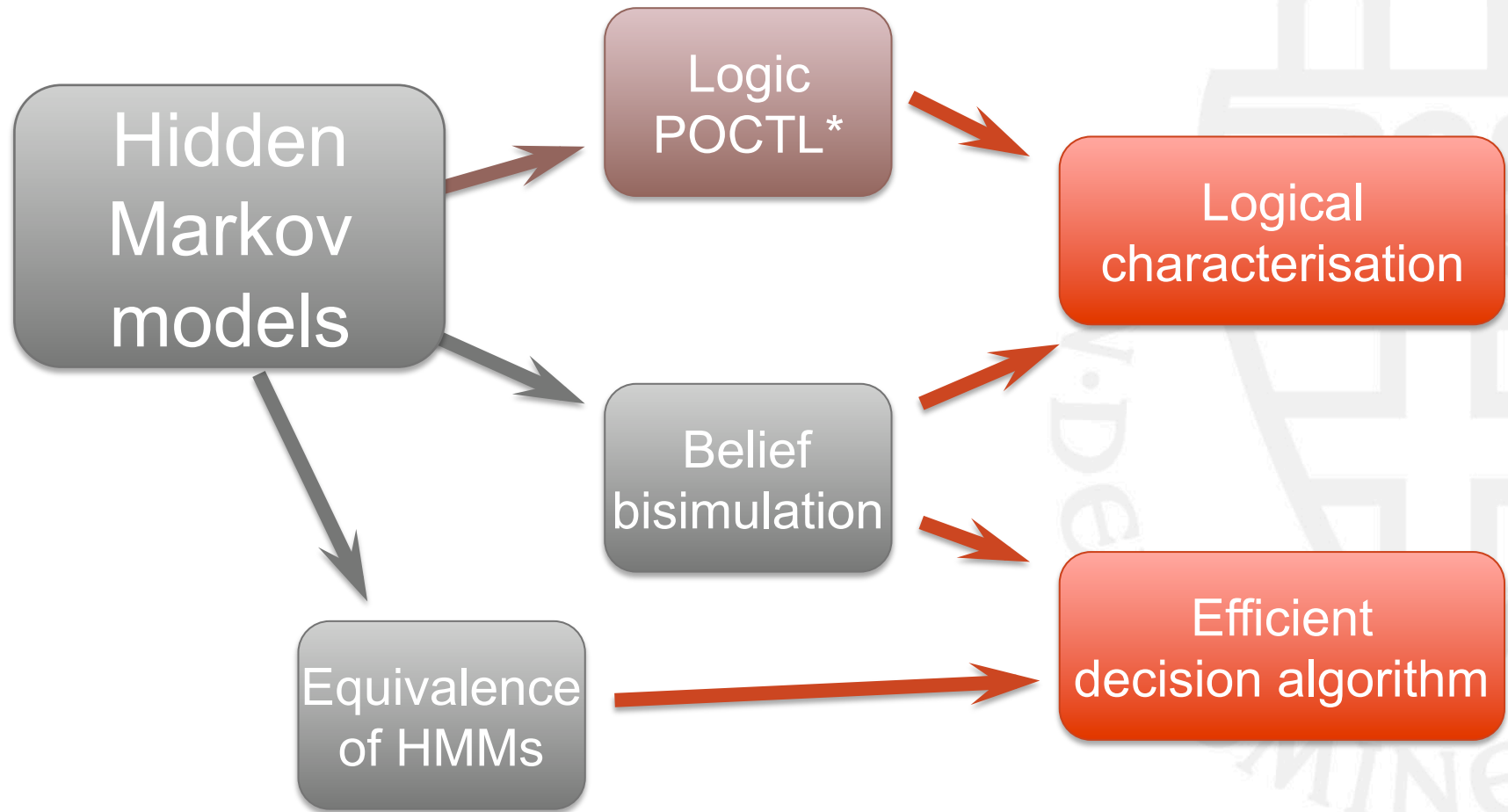


<http://spaceplace.nasa.gov/nanorover/en/>

## Context and Overview



## Context and Overview



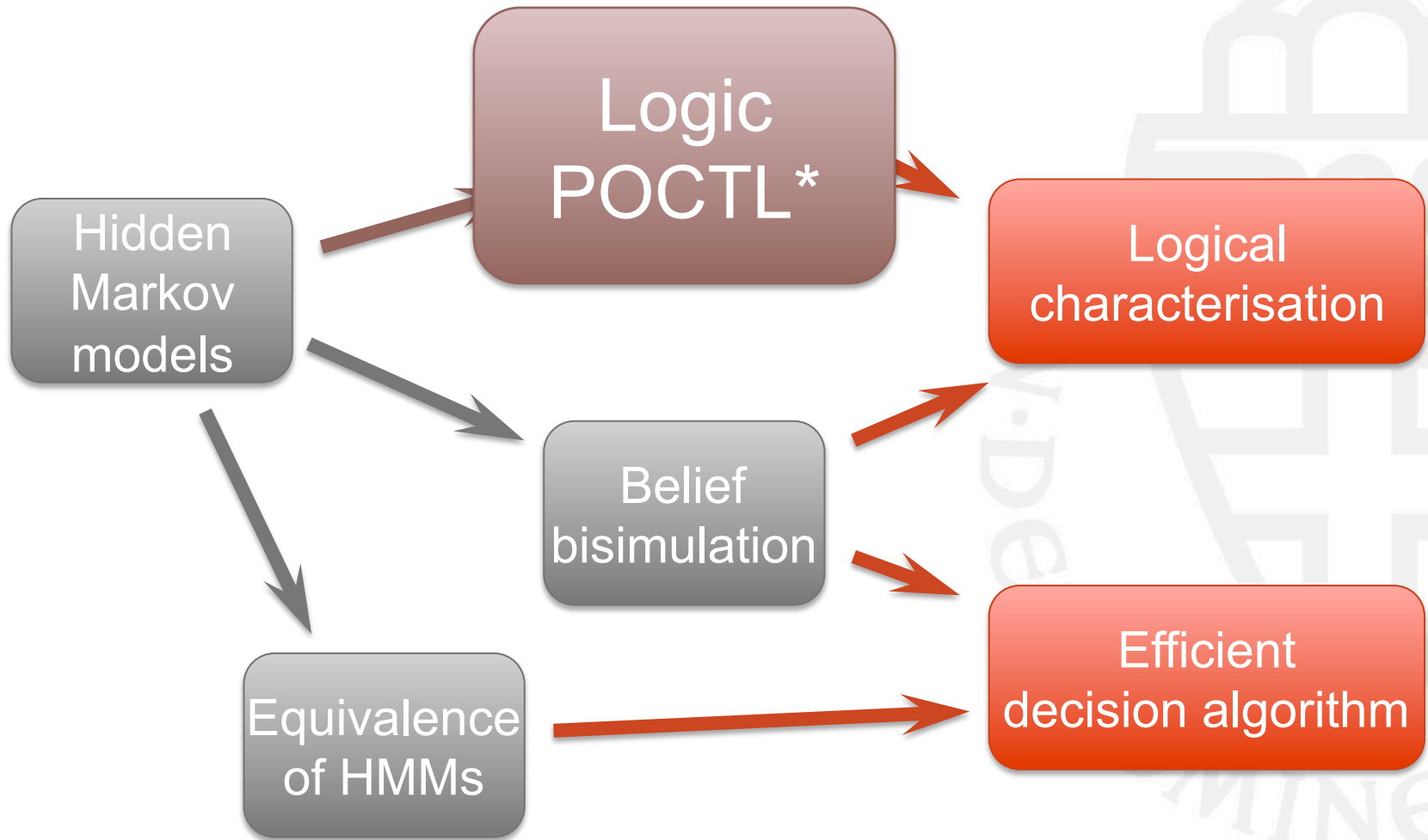
## What is a hidden Markov model?

- describes behaviour of a probabilistic process
- current state is hidden
- observations provide partial information on current state
- Definition:
  - set of states  $S$  (labelled with atomic propositions)
  - set of observations  $\Omega$
  - probabilistic transition relation  $P: S \times S \rightarrow [0,1]$  (labelled with observations)
- Belief state or information state:  
what is known about the current state, based on observations

## What is a HMM good for?

- speech recognition
  - What has been said, given a certain sequence of sounds?
- adaptive communication channel
  - What is the current reliability of the channel, given the recent pattern of breakdowns?
- biological sequence analysis
  - Is the probability of a match high enough, given the two sequences of acids/bases?
- card game
  - Which cards does the player still hold, given a certain play?
- Global question:  
**What state has a high/the highest probability, given a certain sequence of observations?**
- (Problem 2 in Rabiner '89)

## Context and Overview





## A logic for HMMs

- POCTL\* = probabilistic and observation-CTL\*
- extends CTL\* and PCTL
- allows to specify constraints on observations
- Example formulas:
  - The probability is at least 0.5 that after observations  $\omega_1$ ,  $\omega_2$  and  $\omega_3$ , the system is in state *turned*.

$$P_{\geq 0.5}(X_{\omega_1} X_{\omega_2} X_{\omega_3} \text{ turned})$$

- The probability to get observation sequence (u1 u2 u3) or (v1 v2) is smaller than 0.1.

$$P_{< 0.1}((X_{u_1} X_{u_2} X_{u_3} \text{ true}) \vee (X_{v_1} X_{v_2} \text{ true}))$$

- Model checking algorithm for these formulas exists  
[Zhang, Hermanns, Jansen: FORTE'05]

## POCTL\* syntax

state formulas

$$\Phi ::= \text{true} \mid a \mid \neg\Phi \mid \Phi \wedge \Phi \mid \varepsilon$$

path formulas

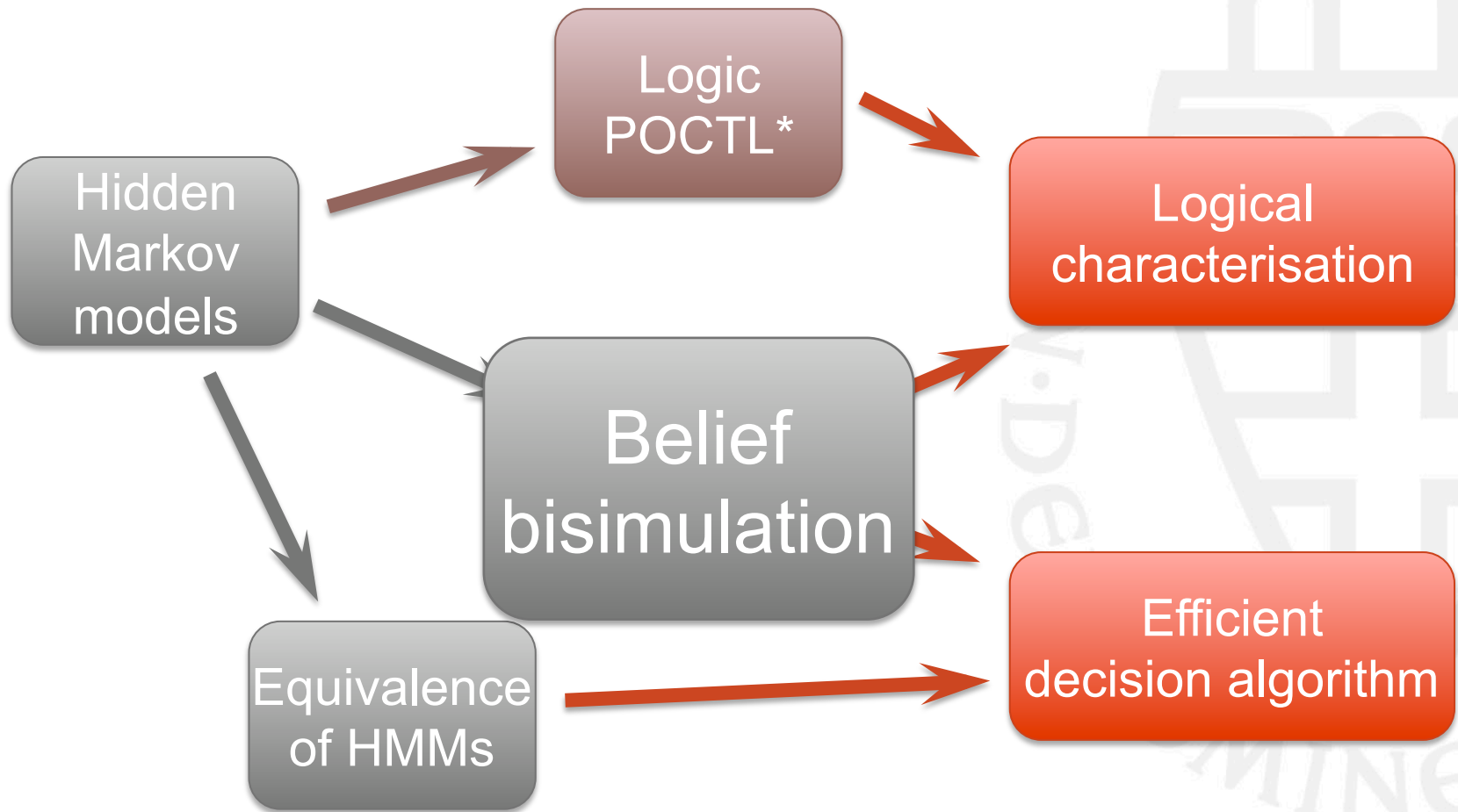
$$\varphi ::= \Phi \mid \neg\varphi \mid \varphi \wedge \varphi \mid X_Y \varphi \mid \varphi U^{\leq n} \varphi$$

belief state formulas

$$\varepsilon ::= \neg\varepsilon \mid \varepsilon \wedge \varepsilon \mid P_{\bowtie p}(\varphi)$$

Constraint on probability of  
paths with a given property

## Context and Overview



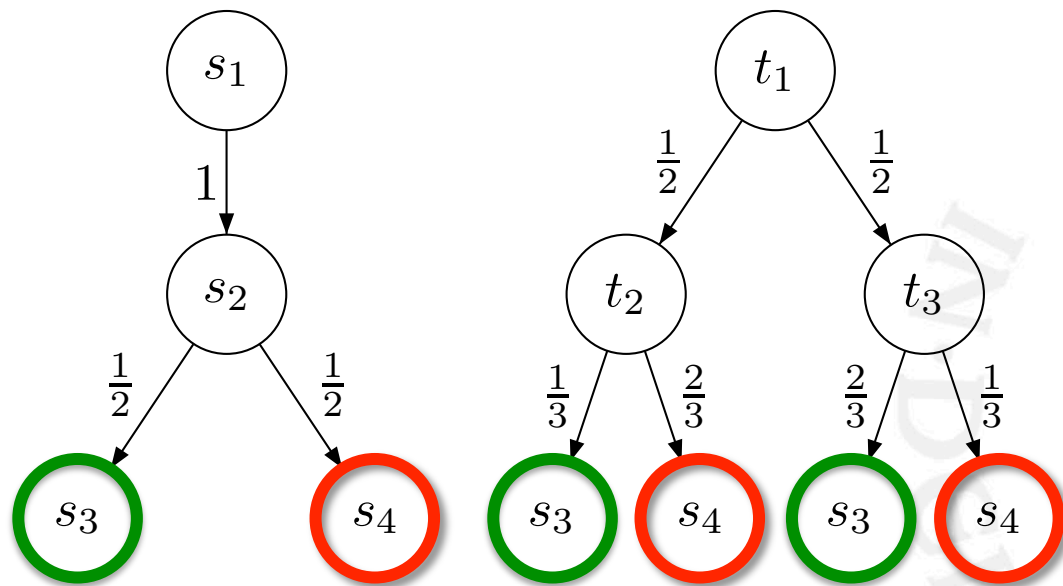
## What is bisimulation?

- Identify states that show equivalent behaviour
- Goal: smaller model  $\rightarrow$  simpler model checking
- standard definition for Markov chains, extended to HMM:
- Equivalence relation  $R \subseteq S \times S$   
is a bisimulation if
  - it respects the **labelling**:  
for all atomic propositions  $a$ ,  $L(s, a) = L(t, a)$
  - it respects the **observations**:  
for all  $\omega \in \Omega$ ,  $\text{Prob}_s(\omega) = \text{Prob}_t(\omega)$
  - it respects the **transitions** (conditional on  $\omega$ ):  
for all  $\omega \in \Omega$  and equivalence classes  $C \in S/R$ ,  $\text{Prob}(C \mid s \xrightarrow{\omega}) R \text{Prob}(C \mid t \xrightarrow{\omega})$for all  $(s, t) \in R$
- notation:  $s \sim t$

## strong belief bisimulation

- Problem of (standard) bisimulation:  
relation between states, but states are invisible
- Solution:  
define relation between belief states
- Equivalence relation  $R \subseteq \mathcal{B} \times \mathcal{B}$   
is a strong belief bisimulation if
  - it respects the **labelling**:  
for all atomic propositions  $a$ ,  $L(b, a) = L(c, a)$
  - it respects the **observations**:  
for all  $\omega \in \Omega$ ,  $\text{Prob}_b(\omega) = \text{Prob}_c(\omega)$
  - it respects the **transitions** (conditional on  $\omega$ ):  
for all  $\omega \in \Omega$ ,  $\text{Prob}(* \mid b \xrightarrow{\omega}) R \text{Prob}(* \mid c \xrightarrow{\omega})$for all  $(b, c) \in R$
- notation:  $b \sim_{\text{sb}} c$

## bisimulation vs. belief bisimulation



$$s_1 \not\sim t_1$$

$$\{s_1 \mapsto 1\} \sim_{sb} \{t_1 \mapsto 1\}$$

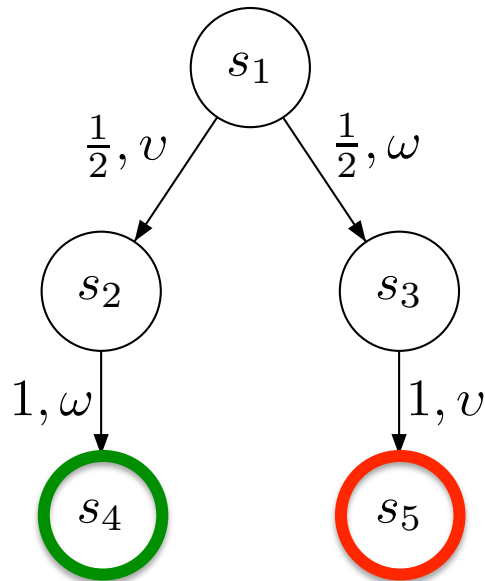
## weak belief bisimulation

- Equivalence relation  $R \subseteq \mathcal{B} \times \mathcal{B}$  is a weak belief bisimulation if
  - it respects the **labelling**:  
for all atomic propositions  $a$ ,  $L(b, a) = L(c, a)$
  - it respects the **observations**:  
for all  $\omega \in \Omega$ ,  $\text{Prob}_b(\omega) = \text{Prob}_c(\omega)$
  - it respects the **transitions**:  
for all equivalence classes  $B \in \mathcal{B}/R$ ,  
 $\text{Prob}(B \mid b \rightarrow) = \text{Prob}(B \mid c \rightarrow)$   
for all  $(b, c) \in R$
- notation:  $b \sim_{\text{wb}} c$

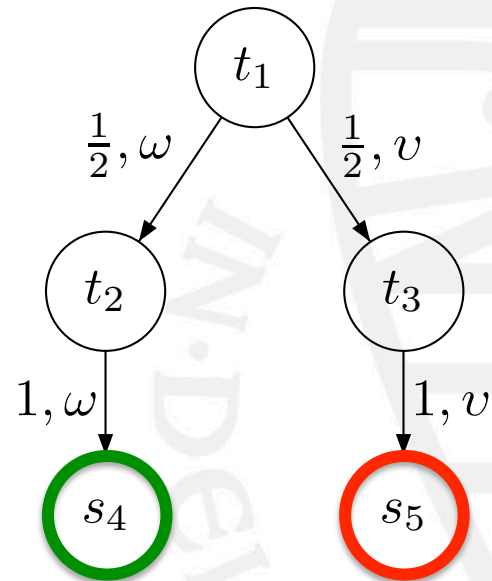
in strong belief bisimulation:

it respects the transitions (conditional on  $\omega$ ):  
for all  $\omega \in \Omega$ ,  $\text{Prob}(* \mid b \xrightarrow{\omega}) R \text{Prob}(* \mid c \xrightarrow{\omega})$

## strong vs. weak belief bisimulation



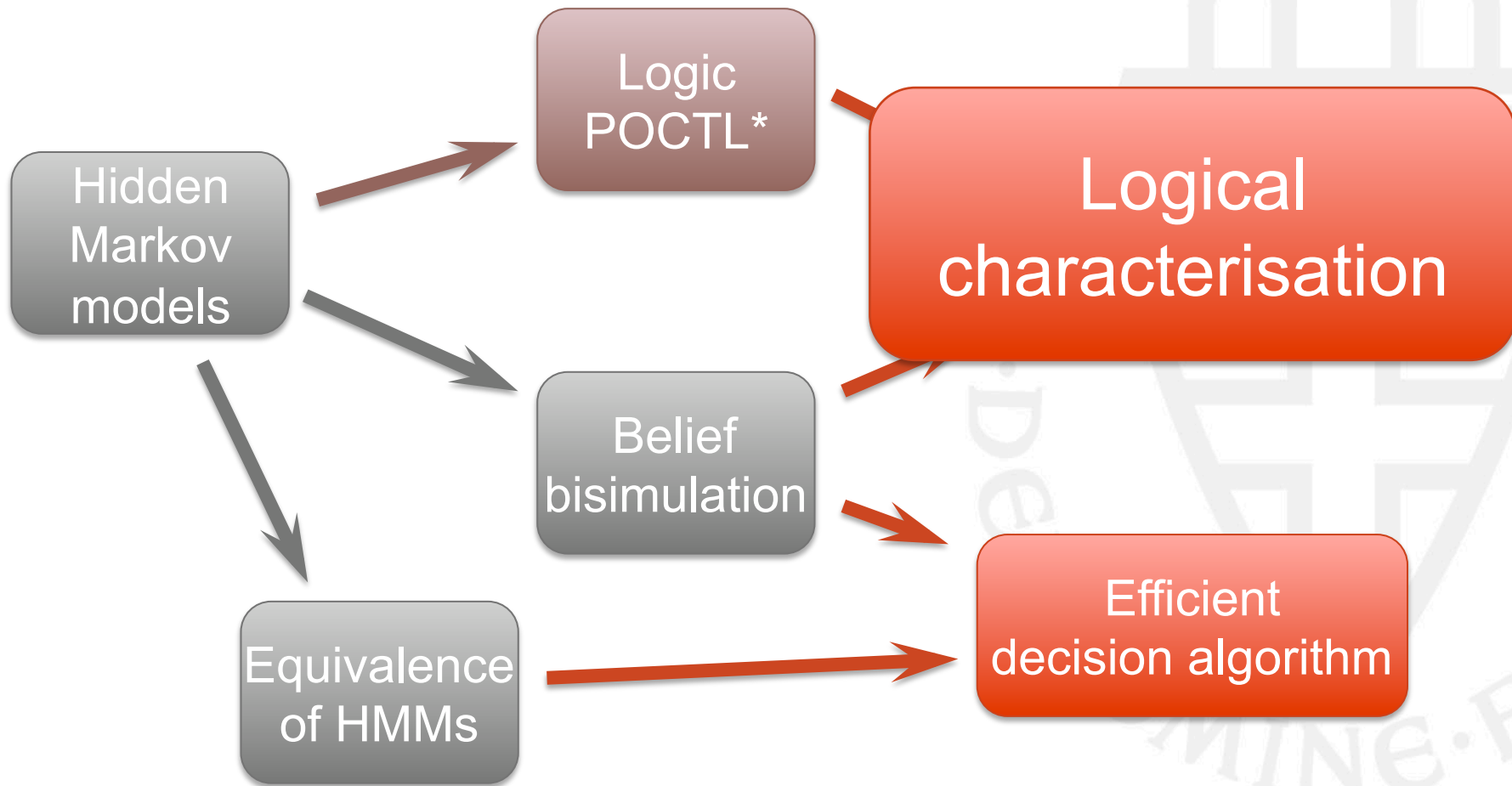
$$\{s_1 \mapsto 1\} \not\sim_{sb} \{t_1 \mapsto 1\}$$



$$\{s_1 \mapsto 1\} \sim_{wb} \{t_1 \mapsto 1\}$$



## Context and Overview



## logical characterisation

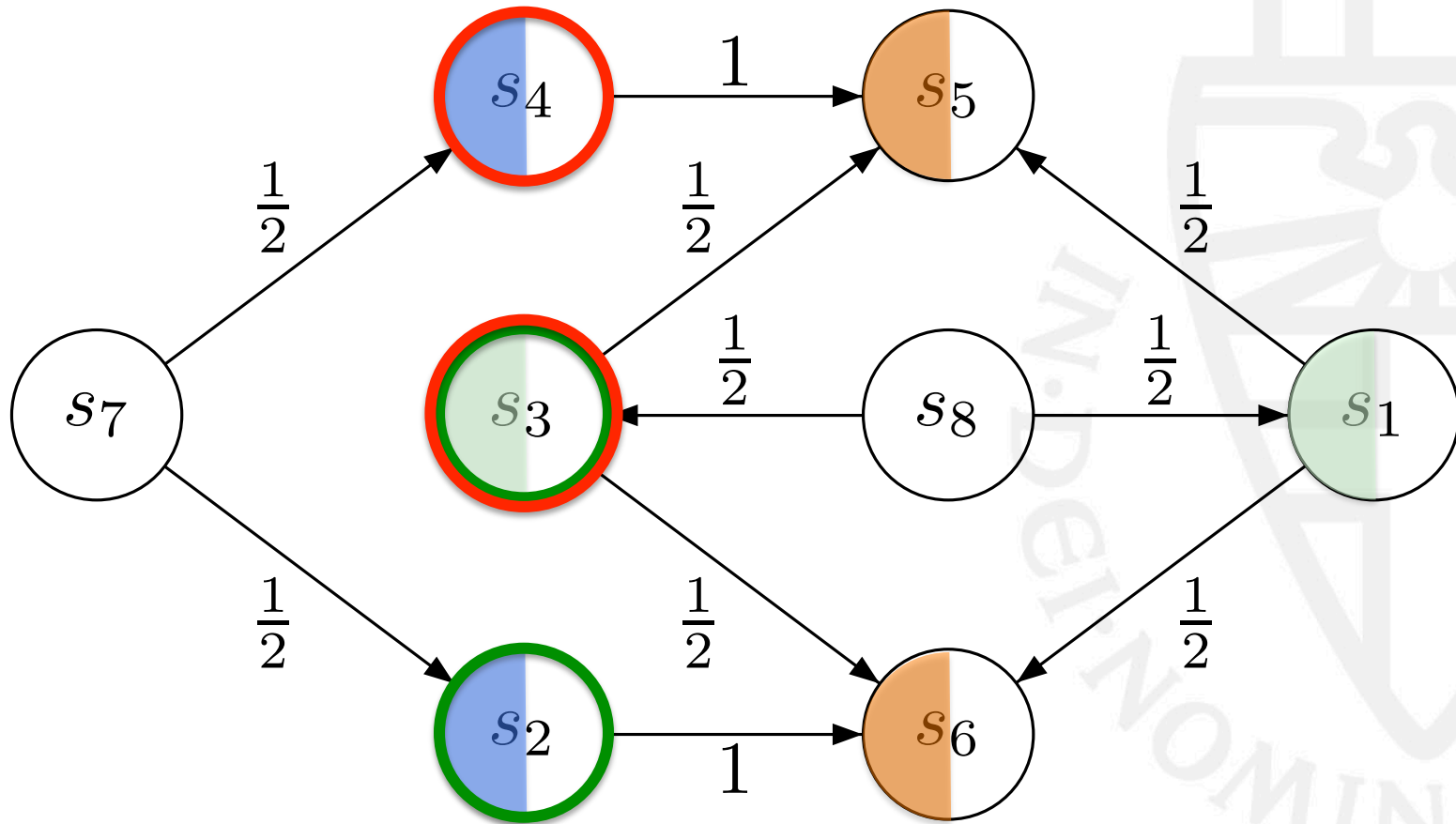
Goal of logical characterisation:

Find a logic with the property

“Two states are bisimilar ( $\sim$ ,  $\sim_{sb}$ ,  $\sim_{wb}$ ) iff they satisfy the same formulas.”

Relation	Characterising logic
$\sim$	POCTL*
$\sim_{sb}$	SBBL*
$\sim_{wb}$	WBBL*

POCTL\* is too strong for  $\sim_{sb}$



## POCTL\* is too strong for $\sim_{sb}$

- |                                |                              |  |
|--------------------------------|------------------------------|--|
| • $\bullet \models$            | • $\nmodels$                 | $P_{\geq 0.5}(P_{\geq 1}(X a_3))$                            |
| • $\bullet \nmodels$           | • $\models$                  | $P_{\geq 0.5}(a_1 \wedge a_2)$                               |
| • $\{s_7 \mapsto 1\} \nmodels$ | $\{s_8 \mapsto 1\} \models$  | $P_{\geq 0.5}(X (a_1 \wedge a_2))$                           |
| • $\{s_7 \mapsto 1\} \models$  | $\{s_8 \mapsto 1\} \nmodels$ | $P_{=0}((X a_1) U^{\leq \infty} a_2)$                        |
| • $\{s_7 \mapsto 1\} \models$  | $\{s_8 \mapsto 1\} \nmodels$ | $P_{=0}(\neg a_1 U^{\leq \infty} (a_2 U^{\leq \infty} a_3))$ |

## SBBL\*

state formulas

$$\Phi ::= \text{true} \mid a \mid \neg\Phi \mid \Phi \wedge \Phi \mid \varepsilon$$

path formulas

$$\varphi ::= \Phi \mid \neg\varphi \mid \varphi \wedge \varphi \mid X_Y \varphi \mid \varphi U^{\leq n} \varphi$$

belief state formulas

$$\varepsilon ::= \neg\varepsilon \mid \varepsilon \wedge \varepsilon \mid P_{\bowtie p}(\varphi) \mid P_{\bowtie p}(\Phi U^{\leq n} \Phi)$$

**SBBL\* characterises strong belief bisimilarity.**

## WBBL\*

state formulas

$$\Phi ::= \text{true} \mid a \mid \neg\Phi$$

path formulas

$$\varphi ::= \Phi \mid X_{\text{Y}} \varphi \mid X_{\text{Y}} \text{true}$$

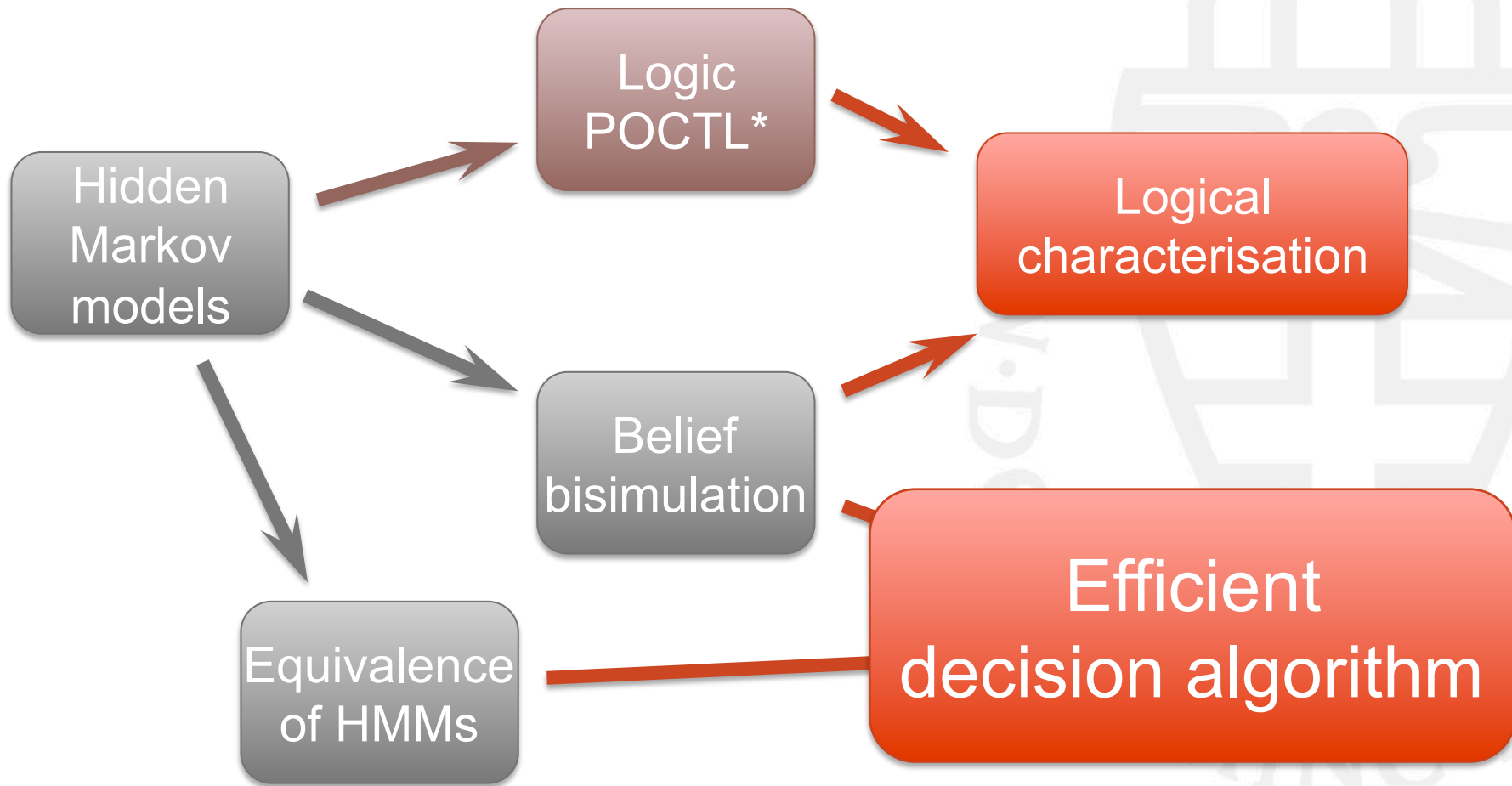
Example formulas  
are not in WBBL\*

belief state formulas

$$\varepsilon ::= \neg\varepsilon \mid \varepsilon \wedge \varepsilon \mid P_{\bowtie p}(\varphi) \mid P_{\bowtie p}(\Phi \cup^{\leq n} \Phi)$$

**WBBL\* characterises weak belief bisimilarity.**

## Context and Overview



## Deciding belief bisimilarity

- problem:  $\mathcal{B}$  is uncountable  
→ no complete description of  $\mathcal{B}/\sim_{sb}$  or  $\mathcal{B}/\sim_{wb}$
- solution:
  - equation system over variables  
 $b_1 := b(s_1), b_2 := b(s_2), \dots, b_n := b(s_n),$   
 $c_1 := c(s_1), c_2 := c(s_2), \dots, c_n := c(s_n)$
  - equality holds  $\Leftrightarrow b \sim_{sb} c$



## Iterative generation of equation system

- basis of iteration:  
belief states  $b$  and  $c$  should have same probability of label or observation

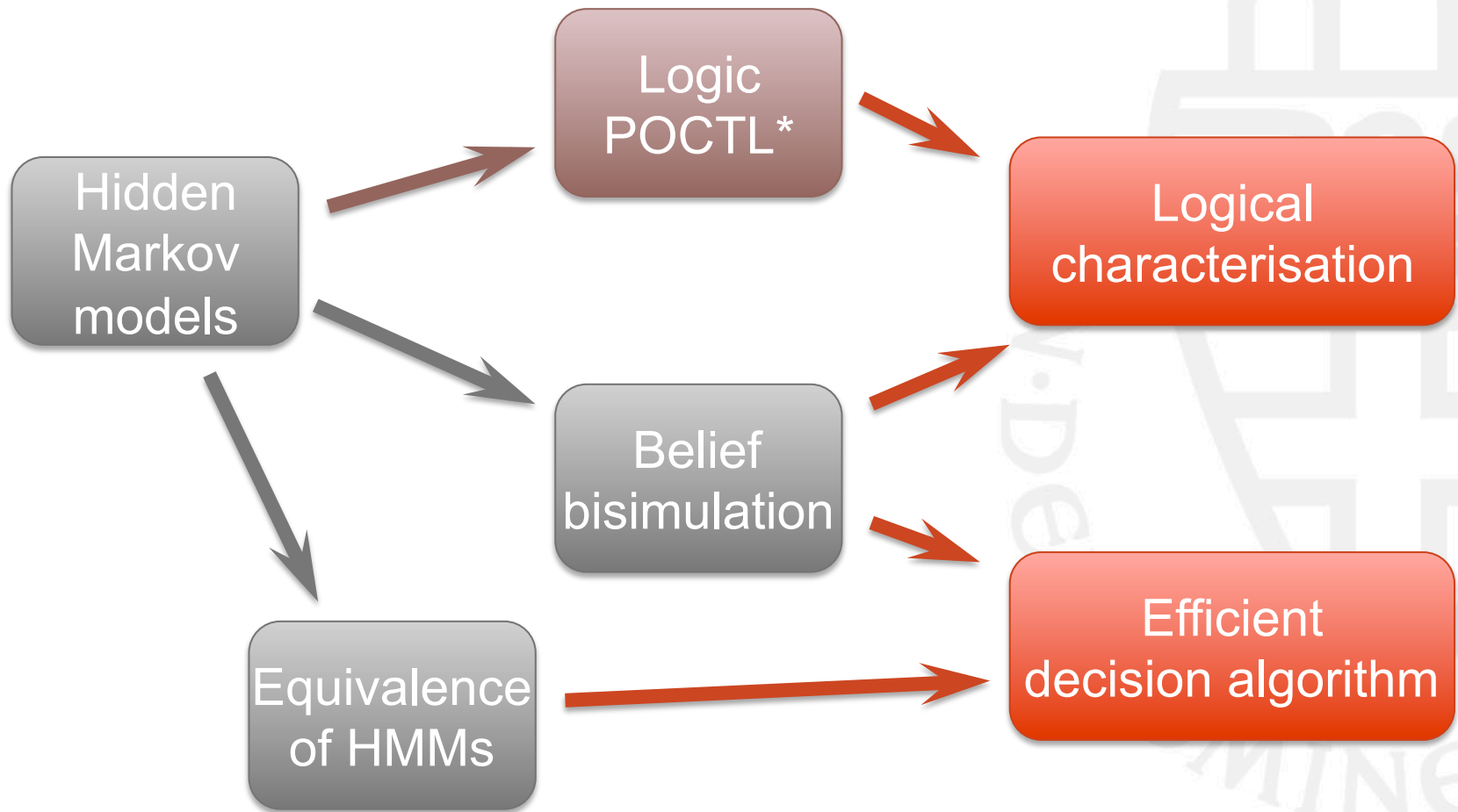
$$\sum_{s_i \models a} b_i = L(b, a) = L(c, a) = \sum_{s_i \models a} c_i$$

- induction step:
  - + regard one more transition
  - + bring into upper triangular form
  - + delete linearly dependent equations
- equation system with  $2|S|$  variables
  - special form: coefficient for  $c_i$  is always negative of coefficient for  $b_i$ 
    - at most  $|S|$  linearly independent equations

## time complexity

- bringing to upper triangular form is most costly step
- weak belief bisimilarity:
  - generate  $\leq |S| + |AP| + |\Omega|$  equations with  $2|S|$  unknowns
  - time complexity is  $\in O(|S|^3)$
- strong belief bisimilarity:
  - generate  $\leq \max \{|AP| + |\Omega|, |S| \cdot |\Omega|\}$  equations with  $2|S|$  unknowns
  - time complexity is  $\in O(|S|^3|\Omega|)$
- better than  $\in O(|S|^4)$   
[Doyen, Henzinger, Raskin: FCS'08]

## Wrapping up



## Conclusions

- WBBL\* is too weak to describe standard property on HMMs:  
“What state has a high/the highest probability,  
given a certain sequence of observations?”
- Weak belief bisimulation does not preserve this property
- Improved time complexity from  $O(|S|^4)$  to  $O(|S|^3)$