

# MuBDDy Children.

## Symbolic Model Checking DEL.

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# Outline

Coffee Machines

Binary Decision Diagrams

Knowledge Structures

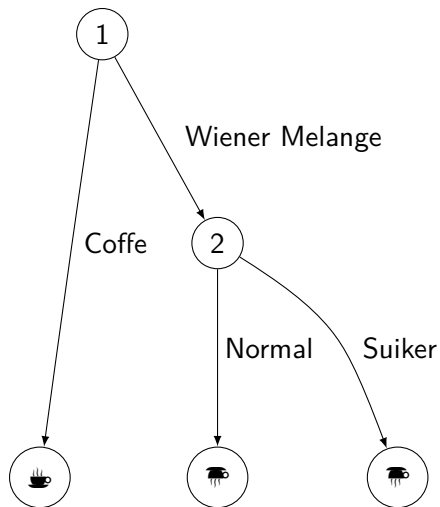
Muddy Children

Product Update

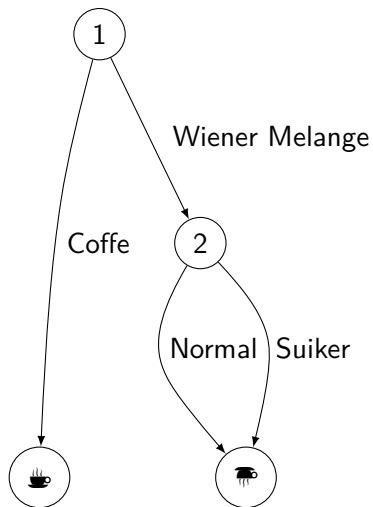
Open Questions

## Coffee Machines

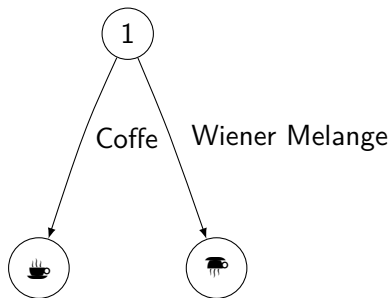
Coffee machines are annoying.



# Coffee machines are still annoying.



Coffee machines do not have to be annoying.



## Binary Decision Diagrams

# Binary Decision Diagrams

## Definition

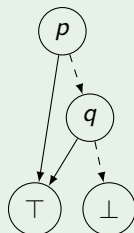
A Binary Decision Diagram [Bry86] for the variables  $V$  is a directed acyclic graph where non-terminal nodes are from  $V$  with two outgoing edges and terminal nodes are  $\top$  or  $\perp$ .

Reduced: No redundancy, identify isomorphic subgraphs.

Ordered: Variables in a given order, maximally once.

By “BDD” we always mean a reduced and ordered BDD.

## Example

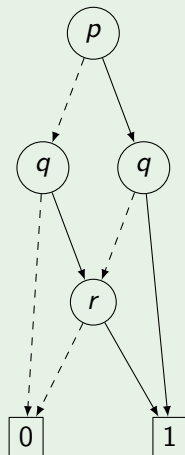




# Binary Decision Diagrams

## Example

$$(p \wedge q) \vee (q \wedge r) \vee (p \wedge r)$$



## Binary Decision Diagrams: Fun facts

Reduced and ordered BDDs provide a compact representation of boolean functions. Nice properties:

- ▶ Small: A BDD often takes less memory than a truth table.
- ▶ Canonical: All *equivalent* formulas have *identical* BDDs.
- ▶ Composable: Given the BDDs for  $\phi$  and  $\psi$  we can easily find those for  $\neg\phi$ ,  $\phi \wedge \psi$ ,  $\phi \vee \psi$ ,  $\forall p\phi$ ,  $\dots$
- ▶ Simultaneous composition: A BDD for  $\forall p(\phi \rightarrow \psi)$  can be constructed from BDDs for  $\phi$  and  $\psi$  “in one go”.

But: Only boolean! How to use BDDs for dynamic logics?

We generalize ideas in [SSL07] and [LSSC08] to cover full DEL.

## Knowledge Structures

## Interlude: Our DEL for today.

### Definition

The DEL language without common knowledge is:

$$\phi ::= p \mid \neg\phi \mid \phi \wedge \phi \mid K_i\phi \mid [!\phi]\phi \mid [\phi]_{\Delta}\phi$$

where  $p \in V$ ,  $i \in I$  and  $\Delta \subseteq I$ . Abbreviations for  $\vee$ ,  $\rightarrow$  and  $\leftrightarrow$ .

The *boolean formulas* over  $V$  are:

$$\phi ::= p \mid \neg\phi \mid \phi \wedge \phi$$

### Fact

Using Kripke Models  $\mathcal{M} = (W, (R_i)_{i \in I}, \text{Val})$  we can give semantics to this language and inductively define  $\mathcal{M}, w \models \phi$ .

# Knowledge Structures

Kripke models: Which situations can you distinguish?

Knowledge structure: What facts do you observe?

## Definition

A knowledge structure is a tuple  $\mathcal{F} = (V, \theta, O_1, \dots, O_n)$  where

- ▶ *Vocabulary*:  $V$  is a set of propositional variables
- ▶ *Law*:  $\theta$  is a boolean formula over  $V$
- ▶ *Observational variables*: for each agent  $i$ ,  $O_i \subseteq V$

An assignment  $s \subseteq V$  is a *state* of  $\mathcal{F}$  iff it satisfies  $\theta$ .

A pair  $(\mathcal{F}, s)$  is called a scenario.

The local state for  $i$  at  $s$  is  $s \cap O_i$ .

## Example

$$\mathcal{F} = (V = \{p\}, \theta = \top, O_1 = \{p\}, O_2 = \emptyset)$$

# DEL Semantics on Knowledge Structures

## Definition

Fix  $\mathcal{F} = (V, \theta, O_1, \dots, O_n)$  and a state  $s$  of  $\mathcal{F}$ .

$$\begin{aligned}(\mathcal{F}, s) \models p &\iff p \in s \\(\mathcal{F}, s) \models \neg\alpha &\iff \text{not } (\mathcal{F}, s) \models \alpha \\(\mathcal{F}, s) \models \alpha \wedge \beta &\iff (\mathcal{F}, s) \models \alpha \text{ and } (\mathcal{F}, s) \models \beta \\(\mathcal{F}, s) \models K_i\alpha &\iff \text{for all states } s' \text{ of } \mathcal{F} : \\&\quad \text{if } s \cap O_i = s' \cap O_i, \text{ then } (\mathcal{F}, s') \models \alpha \\(\mathcal{F}, s) \models [!\psi]\alpha &\iff (\mathcal{F}, s) \models \psi \text{ implies } (\mathcal{F}^{!\psi}, s) \models \alpha \\&\quad \text{where } \|\psi\|_{\mathcal{F}} \text{ is on the next slide} \\&\quad \text{and } \mathcal{F}^{!\psi} = (V, \theta \wedge \|\psi\|_{\mathcal{F}}, O_1, \dots, O_n) \\(\mathcal{F}, s) \models [\psi]_{\Delta}\alpha &\iff (\mathcal{F}, s) \models \psi \text{ implies } (\mathcal{F}_{\psi}^{\Delta}, s \cup p_{\psi}) \models \alpha \\&\quad \text{where } p_{\psi} \text{ is fresh, } \|\psi\|_{\mathcal{F}} \text{ on next slide,} \\&\quad \mathcal{F}_{\psi}^{\Delta} := (V \cup p_{\psi}, \theta \wedge (p_{\psi} \rightarrow \|\psi\|_{\mathcal{F}}), O'_i) \\&\quad \text{and } O'_i = O_i \cup \{p_{\psi} \mid i \in \Delta\}\end{aligned}$$

# Everything is Boolean!

## Definition

Fix  $\mathcal{F} = (V, \theta, O_1, \dots, O_n)$ .

Translate  $\alpha \in \mathcal{L}(V)$  to a boolean formula  $\|\alpha\|_{\mathcal{F}}$ :

$$\begin{aligned} p &\mapsto p \\ \neg\beta &\mapsto \neg\|\beta\|_{\mathcal{F}} \\ \beta_1 \wedge \beta_2 &\mapsto \|\beta_1\|_{\mathcal{F}} \wedge \|\beta_2\|_{\mathcal{F}} \\ K_i\beta &\mapsto \forall(V \setminus O_i)(\theta \rightarrow \|\beta\|_{\mathcal{F}}) \\ [!\psi]\beta &\mapsto \|\psi\|_{\mathcal{F}} \rightarrow \|\beta\|_{\mathcal{F}!\psi} \\ [\psi]_{\Delta}\beta &\mapsto \|\psi\|_{\mathcal{F}} \rightarrow (\|\beta\|_{\mathcal{F}^{\psi}_{\Delta}})(\frac{p_{\psi}}{\mathbf{true}}) \end{aligned}$$

## Lemma

For all scenarios  $(\mathcal{F}, s)$  and all formulas  $\phi$ :

$$(\mathcal{F}, s) \models \phi \iff (\mathcal{F}, s) \models \|\phi\|_{\mathcal{F}}$$

This means we can use BDDs for all KNS and formulas ☺

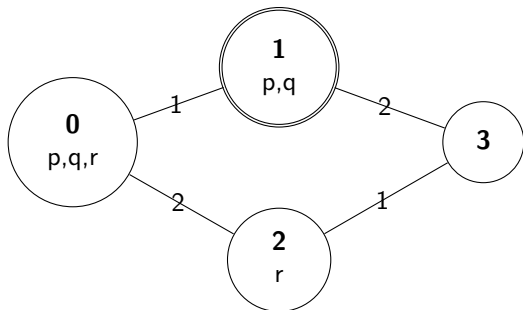
# Knowledge Structures are S5 Kripke Models

## Example

Consider this knowledge structure:

$$\mathcal{F} = (V = \{p, q, r\}, \theta = p \leftrightarrow q, O_1 = \{q\}, O_2 = \{r\})$$

Here we have  $\mathcal{F}, \{p, q\} \models K_1 q$  and  $\mathcal{F}, \{p, q\} \models \neg K_2 q$ .





# Knowledge Structures are S5 Kripke Models

## Theorem

*For any knowledge structure  $\mathcal{F} = (V, \theta, O_i)$  there is an equivalent S5 Kripke model, namely:*

$$\mathcal{M} = (W := \mathcal{P}(V), R_i := \{(s, t) \mid s \cap O_o = t \cap O_i\}, Val := \text{id})$$

## Theorem

*For any S5 Kripke model, there is an equivalent knowledge structure. (Idea: Label all equivalence classes with (sets of) additional propositions to describe the observations.)*

## Muddy Children

## Muddy Children

Father: "At least one of you has mud on their face."

Father: "Do you know if you have mud on your face?"

Nobody reacts.

Father: "Do you know if you have mud on your face?"

Nobody reacts.

Father: "Do you know if you have mud on your face?"

All three children: "Yes, I do have."

## Muddy Children in DEL

The fact that after “Do you know if you have mud on your face?” nobody reacts means that nobody knows their own state:

$$\psi := \bigwedge_{i \in I} (\neg(K_i p_i \vee K_i \neg p_i))$$

The whole story:

$$[!(p_1 \vee p_2 \vee p_3)][!\psi][!\psi](\bigwedge_{i \in I} (K_i p_i))$$

For details and the standard analysis with Kripke semantics, see for example [DHK07, p. 93-96].

# Muddy Children on Knowledge Structures

Initial Model:

$$\mathcal{F}_0 = \left( \begin{array}{l} \{p_1, p_2, p_3\}, \top, \\ O_1 = \{p_2, p_3\} \\ O_2 = \{p_1, p_3\} \\ O_3 = \{p_1, p_2\} \end{array} \right)$$

Announcing a formula means to add ( $\wedge$ ) it to the law ( $\theta$ ) of the knowledge structure. After “At least one of you is dirty”:

$$\mathcal{F}_1 = \left( \begin{array}{l} \{p_1, p_2, p_3\}, p_1 \vee p_2 \vee p_3, \\ O_1 = \{p_2, p_3\} \\ O_2 = \{p_1, p_3\} \\ O_3 = \{p_1, p_2\} \end{array} \right)$$

## Muddy Children on Knowledge Structures

After “At least one of you is dirty.”:

$$\mathcal{F}_1 = \left( \begin{array}{l} \{p_1, p_2, p_3\}, p_1 \vee p_2 \vee p_3, \\ O_1 = \{p_2, p_3\} \\ O_2 = \{p_1, p_3\} \\ O_3 = \{p_1, p_2\} \end{array} \right)$$

Announcing an **epistemic** formula means to add its **boolean equivalent** to the law of the knowledge structure.

After “Nobody knows their own state.”:

$$\mathcal{F}_2 = \left( \begin{array}{l} \{p_1, p_2, p_3\}, (p_1 \vee p_2 \vee p_3) \wedge \|\psi\|_{\mathcal{F}_1}, \\ O_1 = \{p_2, p_3\} \\ O_2 = \{p_1, p_3\} \\ O_3 = \{p_1, p_2\} \end{array} \right)$$

## Muddy Children on Knowledge Structures

$$\begin{aligned}\|K_1 p_1\|_{\mathcal{F}_1} &= \forall(V \setminus O_1)(\theta_1 \rightarrow \|p_1\|_{\mathcal{F}_1}) \\ &= \forall p_1((p_1 \vee p_2 \vee p_3) \rightarrow p_1) \\ &= ((\top \vee p_2 \vee p_3) \rightarrow \top) \wedge ((\perp \vee p_2 \vee p_3) \rightarrow \perp) \\ &= \top \wedge (\neg(p_2 \vee p_3)) \\ &= \neg(p_2 \vee p_3)\end{aligned}$$

$$\begin{aligned}\|K_1 \neg p_1\|_{\mathcal{F}_1} &= \forall(V \setminus O_1)(\theta_1 \rightarrow \|\neg p_1\|_{\mathcal{F}_1}) \\ &= \forall p_1((p_1 \vee p_2 \vee p_3) \rightarrow \neg p_1) \\ &= ((\top \vee p_2 \vee p_3) \rightarrow \neg \top) \wedge ((\perp \vee p_2 \vee p_3) \rightarrow \neg \perp) \\ &= (\top \rightarrow \perp) \wedge ((p_2 \vee p_3) \rightarrow \top) \\ &= \perp\end{aligned}$$

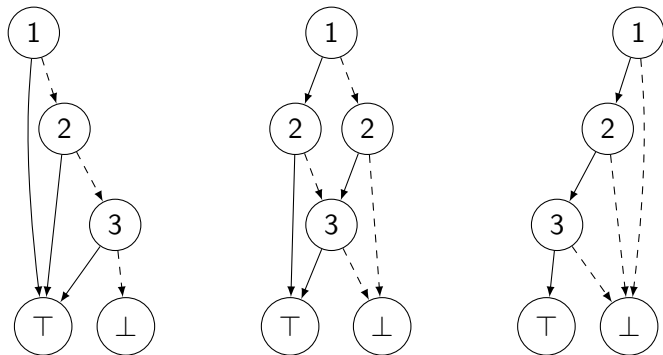
Similarly for agents 2 and 3. The announced formula then becomes

$$\begin{aligned}\|\psi\|_{\mathcal{F}_1} &= \|\bigwedge_{i \in I} (\neg(K_i p_i \vee K_i \neg p_i))\|_{\mathcal{F}_1} = \bigwedge_{i \in I} \|\neg(K_i p_i \vee K_i \neg p_i)\|_{\mathcal{F}_1} \\ &= \neg(\neg(p_2 \vee p_3)) \wedge \neg(\neg(p_1 \vee p_3)) \wedge \neg(\neg(p_1 \vee p_2)) \\ &= (p_2 \vee p_3) \wedge (p_1 \vee p_3) \wedge (p_1 \vee p_2)\end{aligned}$$

# Muddy Children on Knowledge Structures

$$\mathcal{F}_k = \left( \begin{array}{l} \{p_1, p_2, p_3\}, \theta_k, \\ \begin{array}{l} O_1 = \{p_2, p_3\} \\ O_2 = \{p_1, p_3\} \\ O_3 = \{p_1, p_2\} \end{array} \end{array} \right)$$

where  $\theta_1$ ,  $\theta_2$  and  $\theta_3$  are given by:





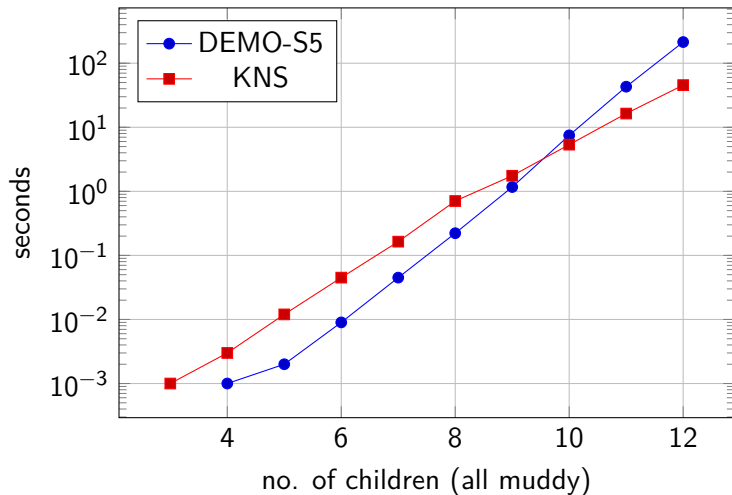
## **A Benchmark for Epistemic Model Checking**

For  $n$  children ( $m$  of them muddy) how many announcements of  $\psi$  are needed until someone knows their own state?

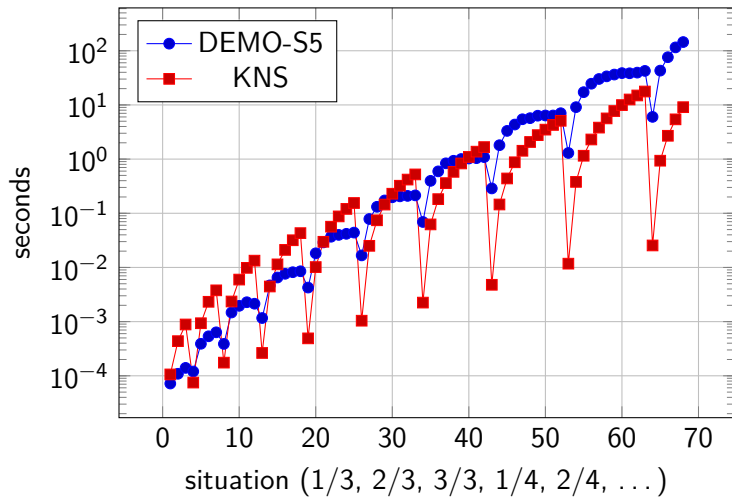
Hint:  $m - 1$

We compare DEMO-S5 from [vE14] with the new BDD approach.

## Performance: Overview



## Performance: Details



## Product Update

## Group announcements

Our language also contains  $[\psi]_{\Delta}\phi$  which says that after announcing  $\psi$  to the group  $\Delta$ ,  $\phi$  holds.

On Knowledge Structures, this is true at  $s$  of

$$\mathcal{F} = (V, \theta, (O_i)_{i \in I})$$

iff  $\mathcal{F}, s \models \psi$  implies that  $\phi$  is true at  $s \cup \{p_{\psi}\}$  of

$$\mathcal{F}' = (V \cup \{p_{\psi}\}, \theta \wedge (p_{\psi} \rightarrow \|\psi\|_{\mathcal{F}}), (O'_i)_{i \in I})$$

where  $O'_i$  contains  $p_{\psi}$  iff  $i \in \Delta$ .

NB: This is *not a secret* announcement.

Now generalize this idea for any S5 product update from [BMS98].

# The general case: Product Update

Similar to the product update

Kripke Model  $\otimes$  Action Model  $\rightarrow$  Kripke Model

we can define a general update mechanism

KNS  $\otimes$  Knowledge Transformer  $\rightarrow$  KNS

## Definition

A *knowledge transformer* for vocabulary  $V$  is a tuple

$$\mathcal{X} = (V^+, \mu, O_1^+, \dots, O_n^+)$$

where

1.  $V^+$  is a set of atomic propositions such that  $V \cap V^+ = \emptyset$ ,
2.  $\mu$  is a (possibly epistemic!) formula over  $V \cup V^+$  and
3.  $O_i^+ \subseteq V^+$  for all agents  $i$ .

# The general case: Product Update

## Definition

Given  $\mathcal{F} = (V, \theta, O_1, \dots, O_n)$  and  $\mathcal{X} = (V^+, \mu, O_1^+, \dots, O_n^+)$ , let

$$\mathcal{F} \otimes \mathcal{X} := (V \cup V^+, \theta \wedge \|\mu\|_{\mathcal{F}}, O_1 \cup O_1^+, \dots, O_n \cup O_n^+)$$

## Theorem

*For every Knowledge Transformer there is an equivalent S5 Action Model and vice versa.*

## Open Questions



# Open Questions

- ▶ How is the performance on other puzzles and protocols? (Sum and Product, Consecutive Numbers, Russian Cards, Dining Cryptographers, ...)
- ▶ Compare to the number triangle approach in [GS11]?
- ▶ Optimize the  $S5 \rightarrow KNS$ -translation by “recognizing” observational propositions before adding new ones?
- ▶ Encode non- $S5$  models in something similar to KNS?
- ▶ Factual change as in [BEK06]?
- ▶ Could KNS help us to find new axiomatizations or completeness proofs?

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Thank you!



Image credit: *Wacken-Besucher feiern im Schlamm*, Focus Online, 05.08.2012.