

# SOCIAL LAWS IN ALTERNATING TIME

Wiebe van der Hoek, Mark Roberts, Michael Wooldridge

University of Liverpool, UK

## Overview

- In this talk, I will:

Introduce a “social laws” framework for multiagent systems, based around the *alternating-time temporal logic* of Alur et al, and in particular, three key problems: *effectiveness*, *feasibility*, and *synthesis*.

- Structure of the talk:
  - introduce ATL;
  - introduce social laws;
  - introduce our social laws framework;
  - point to the future.

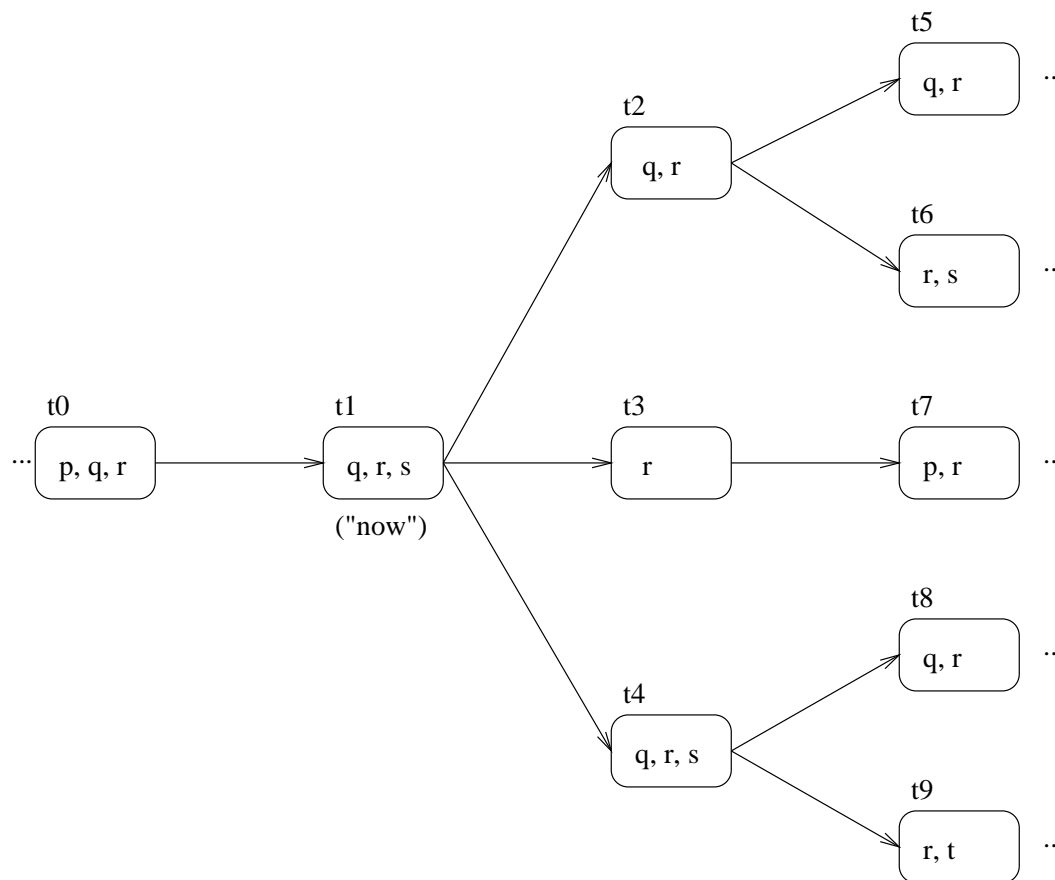
## ATL: A Logic for Multiagent Systems

- *Alternating-time Temporal Logic* (“ATL”) was introduced in 1997 as a logic for reasoning about *game-like distributed systems*: i.e., *multi-agent systems*.
- Main item of novelty: allows us to talk about *powers* of system components.
- Main reason why it’s exciting: it *generalises* CTL, the “generic” branching time logic, without appearing to be more computationally complex.

## Branching-time Temporal Logics

- Natural to view the possible computations of a system as a *tree* linear in the past, branching into the future.
- Branching corresponds to different ways in which non-determinism can be resolved.

## A Branching time model



## Computation Tree Logic: CTL

- Extends propositional logic with
  - *path quantifiers* A, E
  - *tense modalities*  $\bigcirc, \diamond, \square, \mathcal{U}$
- Possible combinations of these are restricted as follows:

$A\bigcirc\varphi$	“on all paths, $\varphi$ is true next
$A\diamond\varphi$	“on all paths, $\varphi$ is eventually true
$A\square\varphi$	“on all paths, $\varphi$ is always true
$A\varphi\mathcal{U}\psi$	“on all paths, $\varphi$ is true until $\psi$
$E\bigcirc\varphi$	“on some path, $\varphi$ is true next
$E\diamond\varphi$	“on some path, $\varphi$ is eventually true
$E\square\varphi$	“on some path, $\varphi$ is always true
$E\varphi\mathcal{U}\psi$	“on some path, $\varphi$ is true until $\psi$

## Models for CTL

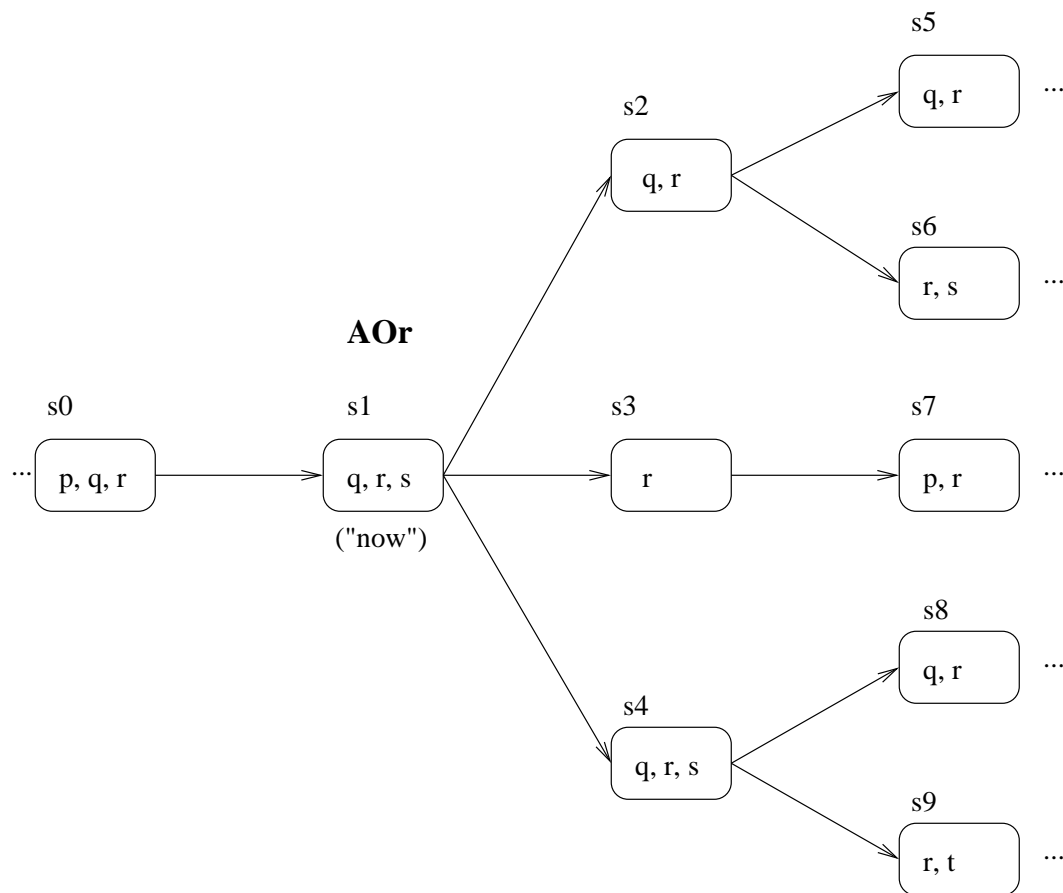
- Models for CTL are *Kripke structures*:

$$\langle S, R, \pi \rangle$$

where

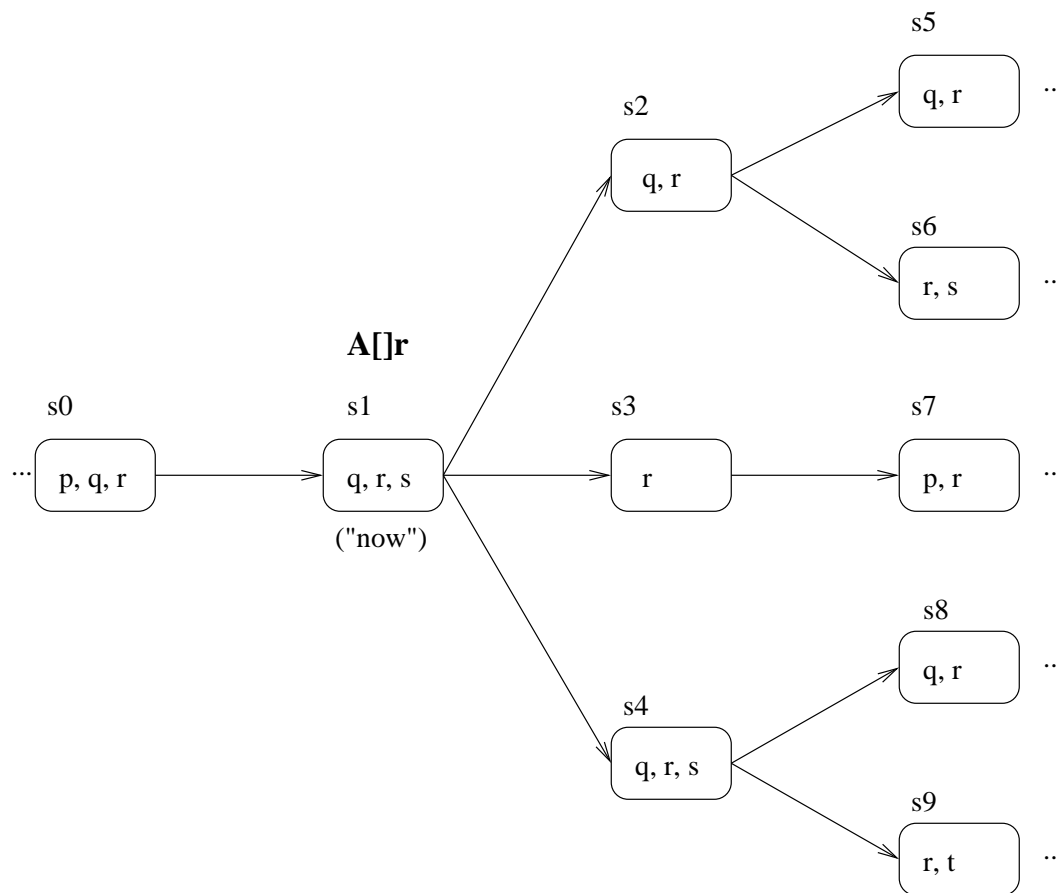
- $S$  is the set of possible system states
  - $R \subseteq S \times S$  is a total binary *next state* relation on  $S$
  - $\pi : S \rightarrow 2^{\Pi}$  says which propositions are true in each state.
- The branches are obtained by *unwinding* this relation, giving *paths* through the structure.

# Example 1

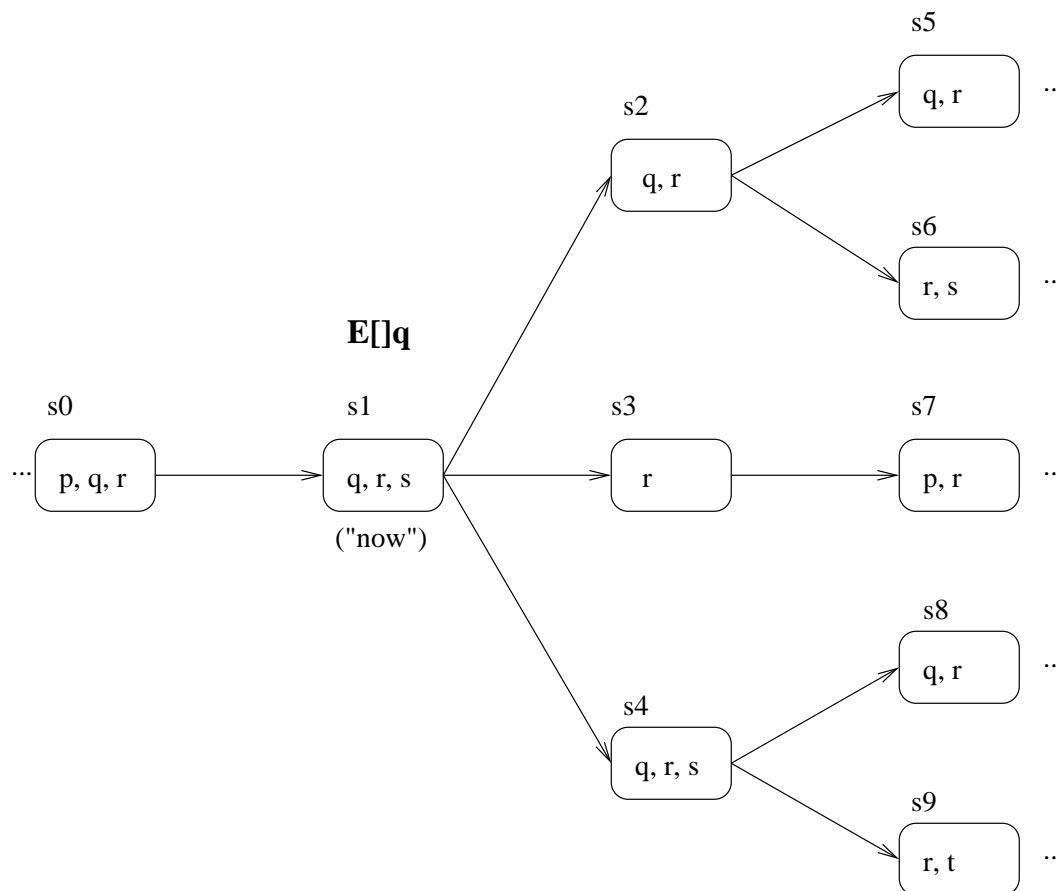




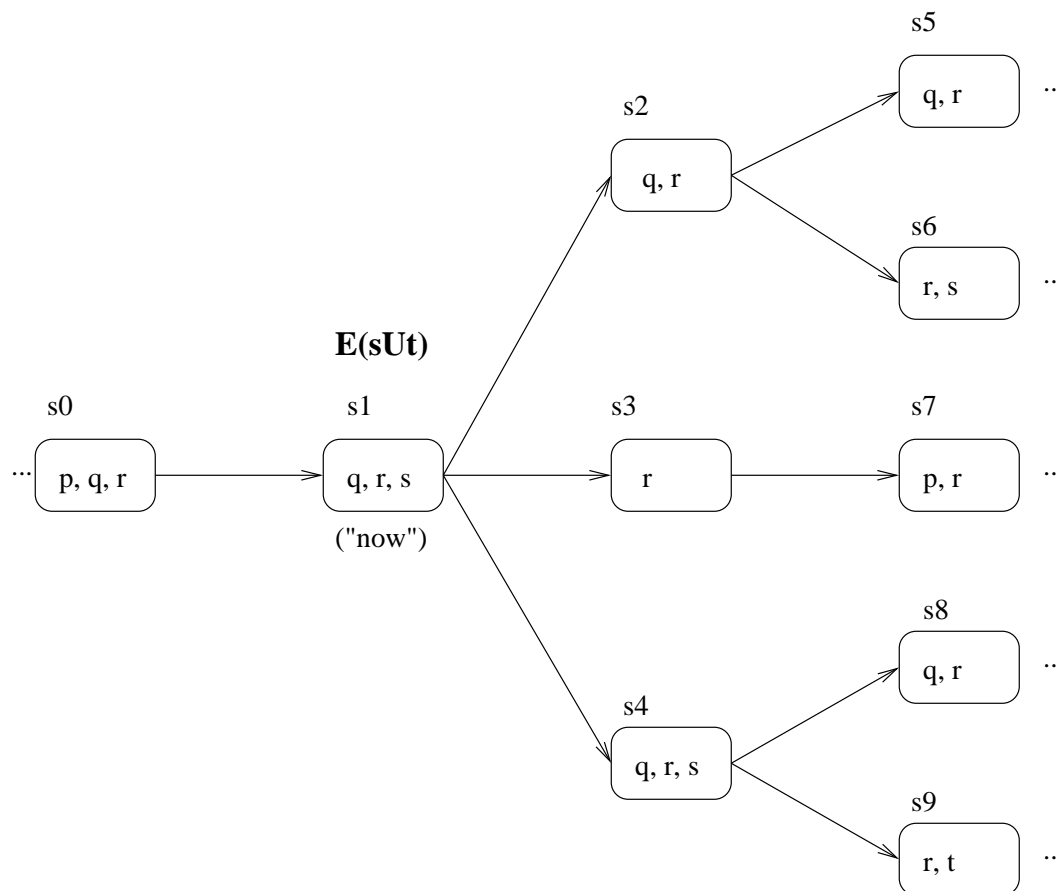
## Example 2



### Example 3



## Example 4



## Computational Properties of CTL

- *Satisfiability* problem for CTL:

*Given CTL formula  $\varphi$  is there some model that satisfies  $\varphi$ ?*

Time complexity: EXPTIME-complete.

(So directly proving properties of systems using CTL looks to be hard.)

- *Model checking* problem for CTL:

*Given model  $M = \langle S, R, \pi \rangle$ , state  $s_0 \in S$ , and formula  $\varphi$ , is  $\varphi$  is true at state  $s_0$  in  $M$ ?*

Time complexity:  $O(|M| \cdot |\varphi|)$ .

## Alternating-time Temporal Logic

- In 1997, Alur, Henzinger & Kupferman proposed a natural variation of CTL known as *Alternating-time Temporal Logic* (ATL).
- Branching used to model evolution of a system controlled by a set of *agents*, which can affect the future by making *choices*.
- The particular future that will emerge depends on *combination* of choices that agents make.
- Thus: a temporal logic built on the notion of *agency*.

## Cooperation Modalities

- Path quantifiers A, E are replaced by *cooperation modalities*:

$$\langle\langle G \rangle\rangle \varphi$$

means

“group  $G$  can cooperate to ensure that  $\varphi$ ”

or, equivalently:

“ $G$  have a collective strategy to force  $\varphi$ ”

- ATL generalises CTL, since

$\langle\langle \emptyset \rangle\rangle$  is same as A

$\langle\langle \Sigma \rangle\rangle$  is same as E

## Computational Problems

- *Satisfiability* problem for ATL:  
EXPTIME-complete (van Drimmelen, 2003).  
Hence no worse than CTL.
- *Model-checking* problem:  
PTIME-complete.  
(Efficient model checkers have been implemented: MOCHA.)

## Example ATL Formulae

$\langle\langle m, j, w \rangle\rangle \Diamond \text{bored-audience}$

$m, j, w$  has a strategy for ensuring that the audience is eventually bored

$\neg \langle\langle m, j, w \rangle\rangle \Box \text{excited}$

$m, j, w$  has no strategy for ensuring that the audience is always excited

$\langle\langle g, w, b, t, b \rangle\rangle \Diamond \text{peace}$

$g, w, b$  and  $t, b$  have a strategy for ensuring that, eventually, there is peace (!)



## Social Laws

- Social laws are *coordination mechanisms*.

A set of rules imposed upon a multiagent system with goal of ensuring that some desirable behaviour will result.

- Work by *prohibiting* the performance of certain actions in certain states.
- Note that this is *not* social sciences *nor* is it law. . . it's computer science!

## Offline and Online Design

There are two ways in which social laws can come to exist in a system:

1. *Offline design*

Mechanisms are engineered at *design time*.

2. *Emergence at run-time*.

Agents develop the social laws at run-time; typically by co-learning, copying, ...

We are here interested in *offline design*.

## Previous Work

Offline design of social laws was first investigated by Moses, Shoham and Tennenholtz

- Moses and Tennenholtz investigate *artificial social systems* and issues that arise in their design
- Shoham and Tennenholtz propose the original framework of social laws and investigate the computational problems associated with them.
- Original framework is extended by Fitoussi and Tennenholtz to incorporate notions of minimal and simple social laws

## Useful Social Laws

- A *constraint* is defined to be a pair

$$\langle E', \alpha \rangle$$

where

- $E' \subseteq E$  is a set of environment states; and
- $\alpha \in Ac$  is an action.

A *social law* is a set  $sl$  of such constraints

- Each agent is associated with set  $F_i$  of *focal states*.

Social law is *useful* if after implementing it, it's possible for every agent to move from any of its focal states to any other focal state.

- Useful social law problem: NP-complete.

## Social Laws in Alternating Time

- We have developed a new framework, in which social laws can be expressed in ATL
- ATL provides a natural language for expressing social laws:
  - can capture *liveness* and *safety* properties, as it generalises CTL
  - can capture *powers* that agents can/should have.
- Some problems associated with social laws, such as the effectiveness, feasibility and synthesis problems reduce directly to ATL model checking problems
- Checked with existing model checkers, such as MOCHA

## Semantic Structures: AATSSs

$$\langle Q, q_0, Ag, Ac_1, \dots, Ac_n, \rho, \tau, \Phi, \pi \rangle$$

- $Q$  is a finite, non-empty set of *states*;
- $q_0 \in Q$  is the *initial state*;
- $Ag = \{1, \dots, n\}$  is a set of *agents*;
- $Ac_i$  is a set of *actions*;
- $\rho : Ac_1 \cup \dots \cup Ac_n \rightarrow 2^Q$  is an *action precondition function*;
- $\tau : Q \times (Ac_1 \times \dots \times Ac_n) \rightarrow Q$  is a *(partial) system transition function*;
- $\Phi$  is a set of *atomic propositions*;
- $\pi : Q \rightarrow 2^\Phi$  is an interpretation function.

(Some obvious *coherence constraints* are required.)

## Social Laws Framework

In our framework, a social law consists of two parts:

1. An *objective*

What we *want to achieve* with this social law.

2. A *behavioural constraint*

The mechanism by which we will achieve it.

## The Formal Definition

A *social law* is a pair

$$(\varphi, \beta)$$

where:

- $\varphi$  is an ATL formula called the *objective* of the law
- $\beta : (Ac_1 \cup \dots \cup Ag_n) \rightarrow 2^Q$  is a *behavioural constraint*.



## Behavioural Constraints

- Behavioural constraints are required to be “reasonable” (every agent must be able to do *something*).
- Implement  $\beta$  in AATS  $S$  = eliminate from  $S$  all transitions forbidden by  $\beta$
- Implementation of  $\beta$  is an *update* on AATSS, resulting in a new AATS
- AATS obtained from  $S$  by implementing  $\beta$  denoted by  $S \dagger \beta$

## Implementing Behavioural Constraints

$$S \dagger \beta = \langle Q, q_0, Ag, Ac_1, \dots, Ac_n, \rho', \tau', \Phi, \pi \rangle,$$

where:

$$1. \forall \alpha \in Ac,$$

$$\rho'(\alpha) = \rho(\alpha) \setminus \beta(\alpha)$$

$$2. \forall q \in Q, \forall j \in J_{Ag},$$

$$\tau'(q, j) = \begin{cases} \tau(q, j) & \text{if } (q, j) \in \text{dom } \tau \text{ and } \forall i \in Ag, q \notin \beta(j_i) \\ \text{undefined} & \text{otherwise} \end{cases}$$

$$3. \text{ All other components of } S \dagger \beta \text{ are as in } S.$$

Note that  $S \dagger \beta$  can be computed in time polynomial in size of  $S, \beta$ .

## Update Properties

Consider *universal* and *existential* sublanguages of ATL, denoted  $\mathcal{L}^u$  and  $\mathcal{L}^e$ :

$$\mathcal{L}^u \quad v ::= p \mid \neg p \mid v \wedge v \mid v \vee v \mid \langle\langle\rangle\rangle \bigcirc v \mid \langle\langle\rangle\rangle \diamond v \mid \langle\langle\rangle\rangle \Box v \mid \langle\langle\rangle\rangle v \mathcal{U} v$$

$$\mathcal{L}^e \quad \epsilon ::= p \mid \neg p \mid \epsilon \wedge \epsilon \mid \epsilon \vee \epsilon \mid \langle\langle Ag \rangle\rangle \bigcirc \epsilon \mid \langle\langle Ag \rangle\rangle \diamond \epsilon \mid \langle\langle Ag \rangle\rangle \Box \epsilon \mid \langle\langle Ag \rangle\rangle \epsilon \mathcal{U} \epsilon$$

where  $p \in \Phi$ .

## Properties of Updates

Suppose we have AATS  $S$ , a behavioural constraint  $\beta$ , a state  $q$  in  $S$ , and formulae  $v \in \mathcal{L}^u$ ,  $\epsilon \in \mathcal{L}^e$ . Then:

1. *Implementation preserves universal properties:*

If  $S, q \models v$  then  $S \dagger \beta, q \models v$ .

2. *Existential properties of updated systems are there in the original system:*

If  $S \dagger \beta, q \models \epsilon$  then  $S, q \models \epsilon$ .

## Effective Social Laws

- A social law  $(\varphi, \beta)$  is *effective* in  $S$  if

$$S \upharpoonright \beta, q_0 \models \langle\langle\rangle\rangle \Box \varphi$$

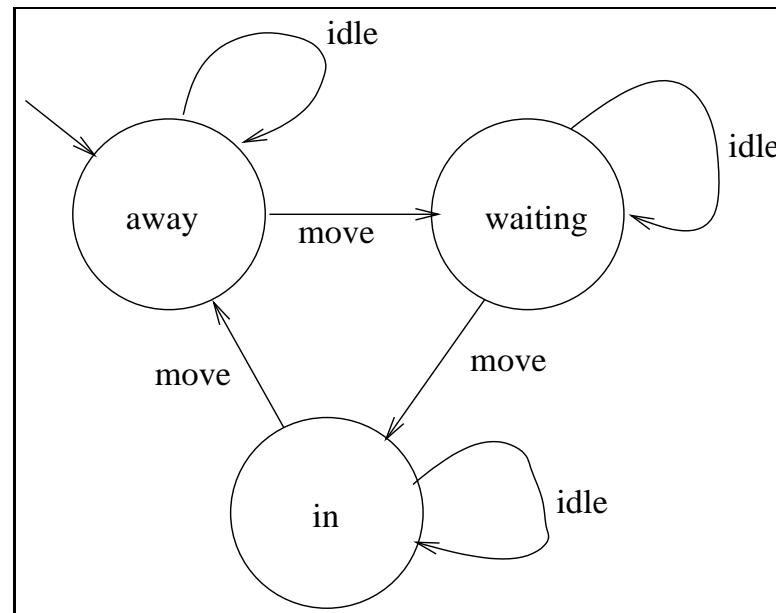
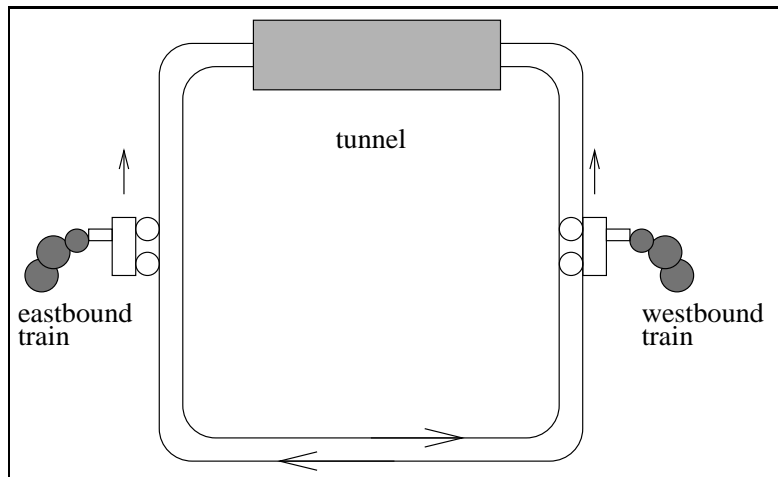
That is, if after implementing it, the objective is guaranteed to hold.

- Thus, the Effectiveness Problem is:

Given  $S$  and  $(\varphi, \beta)$  over  $S$ , determine whether  $(\varphi, \beta)$  is effective in  $S$ .

- Implies effectiveness problem may be solved in time polynomial in the size of  $S$  and  $\varphi$ .

## An Example System



## Example Social Laws

- Obvious requirement: the trains don't crash:

$$O_1 = \neg(in_E \wedge in_W)$$

- Consider the behavioural constraint  $\beta_1$  such that:
  - when both agents are waiting to enter the tunnel, the eastbound train is prevented from moving;
  - when the westbound train is already in the tunnel and the eastbound train is waiting to enter the tunnel, then the eastbound train is prevented from moving; and
  - when the eastbound train is already in the tunnel and the westbound train is waiting to enter the tunnel, then the westbound train is prevented from moving.
- $(O_1, \beta_1)$  is an effective social law in the trains system.

## More Examples

- But consider the social law  $\beta_2$  that simply *prevents both trains from moving*:  $(O_1, \beta_2)$  is also effective!
- Refine our original objective:

$$O_2 = O_1 \wedge \bigwedge_{i \in \{E, W\}} \left( \begin{array}{l} (away_i \rightarrow \langle\langle i \rangle\rangle \diamond waiting_i) \wedge \\ (waiting_i \rightarrow \langle\langle i \rangle\rangle \diamond (in_i \wedge O_1)) \wedge \\ (in_i \rightarrow \langle\langle i \rangle\rangle \bigcirc away_i) \end{array} \right)$$

- $\beta_3$  forbids trains from lingering in tunnel, but is otherwise the same as  $\beta_1$ :  $(O_2, \beta_3)$  is effective.



## The Feasibility Problem

Given  $S$  and a formula  $\varphi$  of ATL representing an objective, does there exist a  $\beta$  such that  $(\varphi, \beta)$  is an effective social law in  $S$ .

- This problem is...  
NP-complete for arbitrary ATL objectives formulae...  
NP-complete for CTL-objectives...  
...and even NP-complete for  $\mathcal{L}^u$  objectives!
- But it is polynomial for propositional logic objectives.

## Synthesis

Given  $S$  and a formula  $\varphi$  of ATL representing an objective, exhibit a behavioural constraint  $\beta$  such that  $(\varphi, \beta)$  is an effective social law in  $S$  if such a constraint exists, otherwise answer “no”.

- In general, requires solving NP-hard optimisation problem. . .

## Synthesis for Propositional Objectives

- There is a link between *model checking* and *synthesis*.
- Intuition: An objective is feasible if the agents could cooperate to make it work.

Suppose  $\varphi$  is a propositional logic formula (representing an objective), and  $S$  is an AATS. Then:

$$S \models \langle\langle Ag \rangle\rangle \Box \varphi \text{ iff } \varphi \text{ is feasible in } S.$$

- So, we get synthesis here as a “side effect” of model checking.

## This Result Does Not Generalise

- This result does not hold for arbitrary formulae.
- Consider the following objective:

$$p \wedge \langle\langle i \rangle\rangle \diamond \neg p$$

- On the one hand want to delete all  $\neg p$  states (to ensure  $\Box p$ )
- But on the other hand, we need them to ensure  $\Box \langle\langle i \rangle\rangle \diamond \neg p$ .

## Synthesis for ATL Objectives

So, for arbitrary ATL objectives, we have:

Suppose  $v \in \mathcal{L}^u$  is a universal ATL formula (representing an objective), and  $S$  is an AATS. Then:

$S \models \langle\langle Ag \rangle\rangle \Box v$  implies  $v$  is feasible in  $S$ .

## Referring to Actions

- Suppose we want to refer to legality of actions *explicitly* in objectives.
- For each action  $\alpha$ , let  $\ell(\alpha)$  be a proposition meaning “ $\alpha$  is legal”.

$$O_4 = \neg(in_E \wedge in_W) \wedge \ell(move_E) \wedge \ell(move_W)$$

$$O_5 = \neg(in_E \wedge in_W) \wedge \ell(move_E)$$

$$O_6 = \neg(in_E \wedge in_W) \wedge \ell(move_W)$$

## Implementation using MOCHA

The idea:

- add “controller” agents, which control the  $\ell(\alpha)$  variables
- change the pre-condition of each action to include this variable.

agent <i>name</i> reads <i>in</i> writes <i>out</i> $P_1 \mapsto \alpha_1;$ $P_2 \mapsto \alpha_2;$ $\dots$ $P_k \mapsto \alpha_k.$	$\Rightarrow$	agent <i>name</i> reads <i>in</i> writes <i>out</i> $\ell(\alpha_1) \wedge P_1 \mapsto \alpha_1;$ $\ell(\alpha_2) \wedge P_2 \mapsto \alpha_2;$ $\dots$ $\ell(\alpha_k) \wedge P_k \mapsto \alpha_k.$
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## Feasibility Again

Again, for propositional objectives, feasibility = model checking.

Suppose  $\varphi$  is a propositional logic formula (representing an objective), and  $S$  is an AATS. Then

$$S^\circ \models \langle\langle \text{controllers} \rangle\rangle \Box \left( \varphi \wedge \left[ \bigwedge_{i \in Ag} \bigvee_{\alpha \in Ac_i} \ell(\alpha) \right] \right)$$

iff  $\varphi$  is feasible in  $S$ .



## That Example Again

$$O_4 = \neg(in_E \wedge in_W) \wedge \ell(move_E) \wedge \ell(move_W)$$

... is not feasible, but. . .

$$O_5 = \neg(in_E \wedge in_W) \wedge \ell(move_E)$$

is feasible, and so is

$$O_6 = \neg(in_E \wedge in_W) \wedge \ell(move_W)$$

## Future Work

- Extend the framework to incorporate *knowledge*  
(Feasibility for ATEL is NP-complete.)
- Algorithmic synthesis of social laws using algorithms
- Examples without trains.