

# Automated Reasoning

## Lecture 14: Linear Temporal Logic II

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# Overview

- ▶ LTL in Isabelle
- ▶ Linear Temporal Logic over finite traces:  $\text{LTL}_f$
- ▶ A soft semantics for  $\text{LTL}_f$

# Recall: Specifications

We are interested in specifying behaviours of systems over time.

- ▶ Use **Temporal Logic**

Specifications are built from:

1. Primitive properties of individual states  
e.g., “is on”, “is off”, “is active”, “is in region”, “is at position”;
2. propositional connectives  $\wedge, \vee, \neg, \rightarrow$ ;
3. and **temporal** connectives: e.g.,
  - ▶ At **all times**, the system is not simultaneously *reading* and *writing*.
  - ▶ If a *request* signal is asserted **at some time**, a corresponding *grant* signal will be asserted **within 10 time units**.
  - ▶ The robot’s *position* will **eventually** be at **1 distance unit** from the shelf.

# Linear Temporal Logic

- ▶ We introduced a commonly used syntax for LTL in the last lecture but in this lecture will use a variant.
  - ▶ This is inspired in part by an existing formalisation in the AFP.
  - ▶ Easier to provide a “soft” semantics (for our discussion of LTL over finite traces).

Our syntax for LTL formulas  $\phi$ :

$$\phi ::= \text{true} \mid \text{false} \mid p \mid \neg p \mid \phi \wedge \phi \mid \phi \vee \phi \mid \mathbf{X}\phi \mid \mathbf{F}\phi \mid \mathbf{G}\phi \mid \phi \mathbf{W}\phi \mid \phi \mathbf{M}\phi$$

Note: We have an explicitly negated atom so a literal is an atom or the negation of an atom.

Temporal operator:

- ▶ **X** — Next
- ▶ **F** — Future; Eventually
- ▶ **G** — Globally; Always
- ▶ **W** — Weak Until
- ▶ **M** — Strong Release

# LTL – Informal Semantics for W and M

- ▶ Recall: The semantics of  $\phi_1 \mathbf{W} \phi_2$  does not require a state to be reached for which  $\phi_2$  holds, unlike  $\phi_1 \mathbf{U} \phi_2$ . Thus, we have:

$$\phi_1 \mathbf{W} \phi_2 \equiv \phi_1 \mathbf{U} \phi_2 \vee \mathbf{G} \phi_1$$

- ▶  $\phi \mathbf{M} \psi$  holds if there is a future position where both  $\phi$  becomes true, and  $\psi$  holds for all positions prior to and including that i.e.  $\phi$  ‘(strongly) releases’  $\psi$ .
  - ▶ It is equivalent to  $\neg(\neg\phi \mathbf{W} \neg\psi)$ .
  - ▶ Thus  $\mathbf{M}$  is the dual of  $\mathbf{W}$ .

**Note:** See previous lecture for the informal meaning of the other operators.

# LTL in Isabelle: Syntax

```
datatype (atoms_ltl: 'a) ltl =
  True_ltl                ("true")
| False_ltl               ("false")
| Prop_ltl 'a             ("prop'(_)'")
| NotProp_ltl 'a          ("nprop'(_)'")
| And_ltl "'a ltl" "'a ltl" ("_ and _" [82,82] 81)
| Or_ltl "'a ltl" "'a ltl"  ("_ or _" [81,81] 80)
| Next_ltl "'a ltl"        ("X _" [88] 87)
| Eventually_ltl "'a ltl"  ("F _" [88] 87)
| Always_ltl "'a ltl"      ("G _" [88] 87)
| WUntil_ltl "'a ltl" "'a ltl" ("W _" [84,84] 83)
| SRelease_ltl "'a ltl" "'a ltl" ("M _" [84,84] 83)
```

- ▶ Deep embedding: the syntax is defined as an explicit datatype in Isabelle/HOL.
- ▶ Note the use of the type variable “*a*”, thus e.g. `prop(q)` stands for atom *q* while `nprop(q)` stands for  $\neg q$ .
- ▶ Negation is not taken as primitive (and thus has no constructor).

# LTL in Isabelle: Not operation

```
fun Not_ltl :: "'a ltl  $\Rightarrow$  'a ltl" ("not _" [85] 85)
where
  "not (true) = false"
| "not (false) = true"
| "not (prop(q)) = nprop(q)"
| "not (nprop(q)) = prop(q)"
| "not ( $\varphi$  and  $\psi$ ) = (not  $\varphi$ ) or (not  $\psi$ )"
| "not ( $\varphi$  or  $\psi$ ) = (not  $\varphi$ ) and (not  $\psi$ )"
| "not (X  $\varphi$ ) = X (not  $\varphi$ )"
| "not (F  $\varphi$ ) = G (not  $\varphi$ )"
| "not (G  $\varphi$ ) = F (not  $\varphi$ )"
| "not ( $\varphi$  W  $\psi$ ) = (not  $\varphi$ ) M (not  $\psi$ )"
| "not ( $\varphi$  M  $\psi$ ) = (not  $\varphi$ ) W (not  $\psi$ )"
```

- (Some of the) Dualities given in the previous lecture are now used to define negation.
- Negation is defined using primitive recursion (we could have used `primrec` instead of `fun`).
- We can easily prove by induction that  $\text{not}(\text{not } \phi) = \phi$ .

# LTL Semantics in Isabelle: Satisfaction by Path

**Satisfaction:**  $\pi \models^i \phi$  – “path at position  $i$  satisfies formula  $\phi$ ”

```
primrec position_semantics_ltl :: "[ 'a set word, nat, 'a ltl]  $\Rightarrow$  bool" ("_  $\models$ _" [80,80,80] 80)
where
  "  $\pi \models^i$  true = True"
| "  $\pi \models^i$  false = False"
| "  $\pi \models^i$  prop(q) = (q  $\in$   $\pi$  i)"
| "  $\pi \models^i$  nprop(q) = (q  $\notin$   $\pi$  i)"
| "  $\pi \models^i \varphi$  and  $\psi$  = ( $\pi \models^i \varphi \wedge \pi \models^i \psi$ )"
| "  $\pi \models^i \varphi$  or  $\psi$  = ( $\pi \models^i \varphi \vee \pi \models^i \psi$ )"
| "  $\pi \models^i X \varphi$  = ( $\exists j. j=i+1 \wedge \pi \models^j \varphi$ )"
| "  $\pi \models^i F \varphi$  = ( $\exists j \geq i. \pi \models^j \varphi$ )"
| "  $\pi \models^i G \varphi$  = ( $\forall j \geq i. \pi \models^j \varphi$ )"
| "  $\pi \models^i W \psi$  = (( $\forall j \geq i. \pi \models^j \varphi$ )  $\vee$  ( $\exists j \geq i. \pi \models^j \psi \wedge (\forall k. i \leq k \wedge k < j \longrightarrow \pi \models^k \varphi)$ ))"
| "  $\pi \models^i M \psi$  = ( $\exists j \geq i. \pi \models^j \varphi \wedge (\forall k. i \leq k \wedge k \leq j \longrightarrow \pi \models^k \psi)$ )"
```

- ▶ LTL semantics is defined as a primitive recursive function (relation) over our datatype.
- ▶ `type_synonym 'a word = nat  $\Rightarrow$  'a`
- ▶  $\pi :: \text{nat} \Rightarrow \text{'a set}$ , i.e.  $\pi(i)$  is the set of atoms that is true (in state) at position  $i$  of path  $\pi$ .



# LTL Semantics in Isabelle: Satisfaction by Path

- Alternative semantics for temporal operators are easily derived as theorems:

**lemma** neXt: " $\pi \models_i X \varphi = \pi \models_{(i+1)} \varphi$ "

**lemma** eventually: " $\pi \models_i F \varphi = (\exists j. \pi \models_{(i+j)} \varphi)$ "

**lemma** always: " $\pi \models_i G \varphi = (\forall j. \pi \models_{(i+j)} \varphi)$ "

**lemma** weak:

$$\pi \models_i \varphi W \psi = ((\forall j. \pi \models_{(i+j)} \varphi) \vee (\exists j. \pi \models_{(i+j)} \psi \wedge (\forall k < j. \pi \models_{(i+k)} \varphi)))$$

**lemma** strong\_release:

$$\pi \models_i \varphi M \psi = (\exists j. \pi \models_{(i+j)} \varphi \wedge (\forall k \leq j. \pi \models_{(i+k)} \psi))$$

# LTL Equivalences

- The expected equivalences follow:

lemma " $\pi \models_i G (\varphi \text{ and } \psi) = (\pi \models_i G \varphi \text{ and } G \psi)$ "

lemma " $\pi \models_i F (\varphi \text{ or } \psi) = (\pi \models_i F \varphi \text{ or } F \psi)$ "

lemma " $\pi \models_i \sigma W (\varphi \text{ or } \psi) = (\pi \models_i (\sigma W \varphi) \text{ or } (\sigma W \psi))$ "

lemma " $\pi \models_i (\varphi \text{ and } \psi) W \sigma = (\pi \models_i (\varphi W \sigma) \text{ and } (\psi W \sigma))$ "

lemma " $\pi \models_i G (F (G \varphi)) = \pi \models_i F (G \varphi)$ "

lemma " $\pi \models_i F (G (F \varphi)) = \pi \models_i G (F \varphi)$ "

lemma " $\pi \models_i G (F \varphi \text{ or } F \psi) = \pi \models_i G (F \varphi) \text{ or } G (F \psi)$ "

# LTL over Finite Traces

- ▶ Many real-world problems involve finite traces or paths rather than infinite ones, especially when dealing with terminating processes.
  - ▶ Example: trajectory constraints in AI planning and, in our research, constraining the finite trajectory being learnt by a neural network by asking it to e.g. avoid particular regions or reach specific points.
- ▶  $LTL_f$  is defined on finite traces.
  - ▶ It uses the same syntax as LTL.
  - ▶ Some changes needed to our understanding of the operators e.g. what is the meaning of  $X\perp$  at the last position?
- ▶ How can we formalise finite paths/traces in Isabelle/HOL?

# LTL<sub>f</sub> Semantics in Isabelle: “Computational Definition”

- We can formalise a semantics recursively over paths, which are represented as lists.

```
function semantics_ltl_f :: "[a set list, 'a ltl] ⇒ bool" ("_ ⊨_" [80,80] 80)
where
  "[] ⊨  $\varphi$  = False"
| "(s₀# $\pi$ ) ⊨ true = True"
| "(s₀# $\pi$ ) ⊨ false = False"
| "(s₀# $\pi$ ) ⊨ prop(q) = (q ∈ s₀)"
| "(s₀# $\pi$ ) ⊨ nprop(q) = (q ∉ s₀)"
| "(s₀# $\pi$ ) ⊨  $\varphi$  and  $\psi$  = ((s₀# $\pi$ ) ⊨  $\varphi$  ∧ (s₀# $\pi$ ) ⊨  $\psi$ )"
| "(s₀# $\pi$ ) ⊨  $\varphi$  or  $\psi$  = ((s₀# $\pi$ ) ⊨  $\varphi$  ∨ (s₀# $\pi$ ) ⊨  $\psi$ )"
| "(s₀# $\pi$ ) ⊨ X  $\varphi$  =  $\pi$  ⊨  $\varphi$ "
| "(s₀# $\pi$ ) ⊨ F  $\varphi$  = ((s₀# $\pi$ ) ⊨  $\varphi$  ∨  $\pi$  ⊨ F  $\varphi$ )"
| "(s₀# $\pi$ ) ⊨ G  $\varphi$  = ((s₀# $\pi$ ) ⊨  $\varphi$  ∧ (if  $\pi$  = [] then True else  $\pi$  ⊨ G  $\varphi$ ))"
| "(s₀# $\pi$ ) ⊨  $\varphi$  W  $\psi$  = (((s₀# $\pi$ ) ⊨  $\varphi$ ) ∧ (if  $\pi$  = [] then True else  $\pi$  ⊨  $\varphi$  W  $\psi$ )) ∨ (s₀# $\pi$ ) ⊨  $\psi$ "
| "(s₀# $\pi$ ) ⊨  $\varphi$  M  $\psi$  = (((s₀# $\pi$ ) ⊨  $\varphi$  and  $\psi$ ) ∨ (if  $\pi$  = [] then False else (s₀# $\pi$ ) ⊨  $\psi$  ∧  $\pi$  ⊨  $\varphi$  M  $\psi$ ))"
by pat_completeness auto
termination by size_change
```

- This gives a more *computational* way of specifying the semantics, which can then be extracted faithfully as code (e.g. Ocaml or Haskell) and executed!
- We prove that this semantics is equivalently expressed in terms path suffixes (next slide).

# LTL<sub>f</sub> Semantics in Isabelle: *ith* Suffix of Path

- Semantics on terms of *ith* suffix of the path (see previous lecture) also uses a list but requires other list operations.

```
function semantics_ltl_f :: "[ 'a set list, 'a ltl] ⇒ bool" ("_ ⊨_" [80,80] 80)
  where
    "[] ⊨ φ = False"
  | "(s#π) ⊨ true = True"
  | "(s#π) ⊨ false = False"
  | "(s#π) ⊨ prop(q) = (q ∈ s)"
  | "(s#π) ⊨ nprop(q) = (q ∉ s)"
  | "(s#π) ⊨ φ and ψ = ((s#π) ⊨ φ ∧ (s#π) ⊨ ψ)"
  | "(s#π) ⊨ φ or ψ = ((s#π) ⊨ φ ∨ (s#π) ⊨ ψ)"
  | "(s#π) ⊨ X φ = π ⊨ φ"
  | "(s#π) ⊨ F φ = (∃ j < length (s#π). drop j (s#π) ⊨ φ)"
  | "(s#π) ⊨ G φ = (∀ j < length (s#π). drop j (s#π) ⊨ φ)"
  | "(s#π) ⊨ φ W ψ = (∀ j < length (s#π). drop j (s#π) ⊨ φ ∨ (∃ k ≤ j. drop k (s#π) ⊨ ψ))"
  | "(s#π) ⊨ φ M ψ = (∃ j < length (s#π). drop j (s#π) ⊨ φ ∧ (∀ k ≤ j. drop k (s#π) ⊨ ψ))"
by pat_completeness auto
termination by size_change
```

- The function `drop n xs` drops the first  $n$  elements of list  $xs$ .
- How do we deal with the end of the path? More generally, does  $\neg X\phi \equiv X \neg\phi$  still hold?

# LTL<sub>f</sub> Negation in Isabelle

Introduce a Weak Next operator:

```
definition WeakNext_ltl :: "'a ltl  $\Rightarrow$  'a ltl" ("Xw _" [88] 87)  
  where "WeakNext_ltl  $\varphi \equiv$  not X (not  $\varphi$ )"
```

that will enable us to define negation:

```
fun Notf_ltl :: "'a ltl  $\Rightarrow$  'a ltl" ("notf _" [85] 85)  
where  
  "notf (true) = false"  
| "notf (false) = true"  
| "notf (prop(q)) = nprop(q)"  
| "notf (nprop(q)) = prop(q)"  
| "notf ( $\varphi$  and  $\psi$ ) = (notf  $\varphi$ ) or (notf  $\psi$ )"  
| "notf ( $\varphi$  or  $\psi$ ) = (notf  $\varphi$ ) and (notf  $\psi$ )"  
| "notf (X  $\varphi$ ) = Xw (not  $\varphi$ )"  
| "notf (F  $\varphi$ ) = G (notf  $\varphi$ )"  
| "notf (G  $\varphi$ ) = F (notf  $\varphi$ )"  
| "notf ( $\varphi$  W  $\psi$ ) = (notf  $\varphi$ ) M (notf  $\psi$ )"  
| "notf ( $\varphi$  M  $\psi$ ) = (notf  $\varphi$ ) W (notf  $\psi$ )"
```

- ▶ It then follows that  $\text{not}_f(\text{not}_f(\phi)) = \phi$ .
- ▶ Most of the usual LTL equivalences are recovered.

## From Discrete to RobustSemantics

- ▶ Recall: We are interested in specifying behaviours of systems over time.
- ▶ Informally, can we find a way of specifying such behaviours rigorously and then incorporating them into the training of a neural network to ensure it (attempts) to satisfy them?
  - ▶ Issue: The semantics of LTL and  $LTL_f$  are boolean so either a formula is satisfied or it is not. So, they are not robust (and “learning”-friendly).
  - ▶ Can we formulate a soft (robust) semantics that will approximate the standard one and thus enable us to quantify by how much a  $LTL_f$  constraint is violated during learning i.e. figure out the constraint-related *loss*?
  - ▶ Moreover, we would like this function to be differentiable so that we can use it to minimise the constraint loss during back propagation (cf. Mark’s Guest Lecture).

# Some Soft Functions

- We formalise soft versions of the functions max and minimum:

```
fun Max_gamma :: "real  $\Rightarrow$  real  $\Rightarrow$  real  $\Rightarrow$  real" ("_  $\sqcap_\gamma$  _" [82,82] 81) where
  "(a  $\sqcap_\gamma$  b) = (if  $\gamma \leq 0$  then max a b
                 else  $\gamma * \ln (\exp (a / \gamma) + \exp (b / \gamma)))$ "

fun Min_gamma :: "real  $\Rightarrow$  real  $\Rightarrow$  real  $\Rightarrow$  real" ("_  $\sqcup_\gamma$  _" [81,81] 80) where
  "(a  $\sqcup_\gamma$  b) = (if  $\gamma \leq 0$  then min a b
                 else  $-\gamma * \ln (\exp (-a / \gamma) + \exp (-b / \gamma)))$ "
```

- Each of these soft functions takes an additional smoothing parameter  $\gamma$ .
- We show that as  $\gamma \rightarrow 0$ ,  $\sqcap_\gamma \rightarrow \max$ ,  $\sqcup_\gamma \rightarrow \min$ , and that they are differentiable for  $\gamma > 0$ .



# A Simple Soft LTL<sub>f</sub> Semantics in Isabelle

- This is a simple formalisation but it illustrates how we can move from a boolean to a real-valued semantics:

```
function soft_semantics_ltl_f :: "[ 'a set list, real, 'a ltl ]  $\Rightarrow$  real" ("_  $\models_s$  _" [80,80,80] 80)
where
  "[ ]  $\models_s \gamma \varphi = 1$ "
| "(s# $\pi$ )  $\models_s \gamma$  true = 0"
| "(s# $\pi$ )  $\models_s \gamma$  false = 1"
| "(s# $\pi$ )  $\models_s \gamma$  prop(q) = (if (q  $\in$  s) then 0 else 1)"
| "(s# $\pi$ )  $\models_s \gamma$  nprop(q) = (if (q  $\notin$  s) then 0 else 1)"
| "(s# $\pi$ )  $\models_s \gamma \varphi$  and  $\psi = ((s# $\pi$ )  $\models_s \gamma \varphi$ )  $\sqcap_\gamma$  ((s# $\pi$ )  $\models_s \gamma \psi$ )"
| "(s# $\pi$ )  $\models_s \gamma \varphi$  or  $\psi = ((s# $\pi$ )  $\models_s \gamma \varphi$ )  $\sqcup_\gamma$  ((s# $\pi$ )  $\models_s \gamma \psi$ )"
| "(s# $\pi$ )  $\models_s \gamma$  X  $\varphi = \pi \models_s \gamma \varphi$ "
| "(s# $\pi$ )  $\models_s \gamma$  F  $\varphi = ((s# $\pi$ )  $\models_s \gamma \varphi$ )  $\sqcup_\gamma$  ( $\pi \models_s \gamma$  F  $\varphi$ )"
| "(s# $\pi$ )  $\models_s \gamma$  G  $\varphi = ((s# $\pi$ )  $\models_s \gamma \varphi$ )  $\sqcap_\gamma$  (if  $\pi = []$  then 0 else  $\pi \models_s \gamma$  G  $\varphi$ )"
| "(s# $\pi$ )  $\models_s \gamma \varphi$  W  $\psi =$ 
  (((s# $\pi$ )  $\models_s \gamma \varphi$ )  $\sqcap_\gamma$  (if  $\pi = []$  then 0 else  $\pi \models_s \gamma \varphi$  W  $\psi$ ))  $\sqcup_\gamma$  ((s# $\pi$ )  $\models_s \gamma \psi$ )"
| "(s# $\pi$ )  $\models_s \gamma \varphi$  M  $\psi =$ 
  ((s# $\pi$ )  $\models_s \gamma \varphi$  and  $\psi$ )  $\sqcup_\gamma$  (if  $\pi = []$  then 1 else ((s# $\pi$ )  $\models_s \gamma \psi$ )  $\sqcap_\gamma$  ( $\pi \models_s \gamma \varphi$  M  $\psi$ ))"
by pat_completeness auto
termination by size_change$$$$ 
```

- We have a literal translation of the standard LTL<sub>f</sub> semantics we presented earlier, with  $\wedge$  and  $\vee$  replaced by our softmax  $\sqcap_\gamma$  and softmin  $\sqcup_\gamma$  functions respectively.

# Is our Soft Semantics Sound?

Yes! We can formally prove:

- ▶ with  $\gamma = 0$ , it is exactly the standard boolean semantics:

$$\text{lemma } "(\pi \models_{s0} \varphi = 0) = (\pi \models \varphi)"$$

- ▶ but more importantly, we have:

$$\text{lemma } "((\lambda\gamma. \pi \models_{s\gamma} \varphi) -0\rightarrow 0) = (\pi \models \varphi)"$$

where, mathematically, the theorem states:

$$\lim_{\gamma \rightarrow 0} (\pi \models_{s\gamma} \phi) = 0 \iff \pi \models \phi$$

- ▶ We thus have a well-defined semantic function
  - ▶ In principle it can be used to implement the loss with respect to LTL constraints during learning by a neural network.
  - ▶ It will be rather limited and excruciatingly slow though; so a much more sophisticated formalisation is needed for practical use.
- ▶ Next Lecture: An overview of how this can be adapted and used for rigorous, constrained neural learning.

# Summary

- ▶ Linear Temporal Logic (H&R 3.2)
  - ▶ Syntax and Semantics: see also  
<https://www.isa-afp.org/sessions/ltl>.
- ▶ Linear Temporal Logic over finite traces:  $LTL_f$ 
  - ▶ See, for example: *Linear Temporal Logic and Linear Dynamic Logic on Finite Traces* by De Giacomo and Vardi,  
<https://www.cs.rice.edu/~vardi/papers/ijcai13.pdf>
- ▶ A soft semantics for  $LTL_f$ 
  - ▶ See our paper: *Constrained Training of Neural Networks via Theorem Proving*,  
<https://ceur-ws.org/Vol-3311/paper2.pdf> and  
<https://arxiv.org/pdf/2207.03880>
  - ▶ See also: *Elaborating on Learned Demonstrations with Temporal Logic Specifications* by Innes and Ramamoorthy,  
<https://arxiv.org/pdf/2002.00784>