

# Hennessy-Milner Theorems via Galois Connections

Barbara König  
Universität Duisburg-Essen

Based on joint work with Harsh Beohar, Sebastian Gurke,  
Karla Messing

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## Motivation

Hennessy-Milner theorems connect behavioural equivalences with modal logics. Given two states  $x_1, x_2 \in X$  and formulas  $\varphi$ :

$$x_1 \sim x_2 \iff \forall \varphi: (x_1 \models \varphi \iff x_2 \models \varphi)$$

There is a metric analogue, where  $d$  is a pseudo-metric on the state space and formulas  $\varphi$  evaluate to real-valued predicates  $\llbracket \varphi \rrbracket: X \rightarrow [0, 1]$

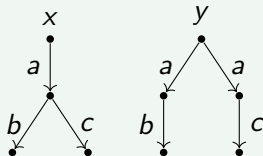
$$d(x_1, x_2) = \bigvee_{\varphi} | \llbracket \varphi \rrbracket(x_1) - \llbracket \varphi \rrbracket(x_2) |$$

Given a behavioural equivalence, the aim is typically to determine a modal logic that characterizes this equivalence

$\rightsquigarrow$  linear-time/branching-time spectrum [van Glabbeek],  
coalgebraic modal logics

# Motivation

## Example



$x, y$  are not bisimilar ( $x \not\sim y$ )

They are distinguished by  $\varphi = \diamond_a(\diamond_b \text{ true} \wedge \diamond_c \text{ true})$  where  $x \models \varphi, y \not\models \varphi$ .

# Motivation

Our contributions:

- Hennessy-Milner theorems can be obtained from the fact that least fixpoints are preserved by left adjoints (of a Galois connection).
- Rather than starting with the definition of a behavioural equivalence, we go the other way and derive fixpoint equations for behavioural equivalences/metrics from the modal logics. (Including compositionality results.)
- We obtain (new) fixpoint equations for decorated trace metrics.

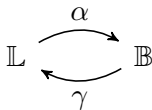
# Galois Connection

## Definition

Let  $\mathbb{L}$ ,  $\mathbb{B}$  be two complete lattices with order  $\sqsubseteq$ . A **Galois connection** from  $\mathbb{L}$  to  $\mathbb{B}$  is a pair  $\alpha \dashv \gamma$  of monotone functions such that

$$\alpha(l) \sqsubseteq m \iff l \sqsubseteq \gamma(m),$$

for all  $l \in \mathbb{L}$ ,  $m \in \mathbb{B}$ .



Intuition in our case:  $\mathbb{L}$  – logical universe,  $\mathbb{B}$  – behaviour universe

# General setting

$$\begin{array}{ccccc}
 & & \alpha & & \\
 \log \hookrightarrow & \mathbb{L} & \xrightarrow{\quad} & \mathbb{B} & \hookrightarrow \text{beh} \\
 & & \gamma & & \\
 & & \text{beh} = \alpha \circ \log \circ \gamma & & 
 \end{array}$$

## Compatibility

Let  $\log, c: \mathbb{L} \rightarrow \mathbb{L}$  be two monotone endo-functions on a lattice  $\mathbb{L}$ . We call  $\log$  **c-compatible** whenever  $\log \circ c \sqsubseteq c \circ \log$ .

Compatibility: concept borrowed from up-to techniques

## Theorem

Let  $\alpha: \mathbb{L} \rightarrow \mathbb{B}$ ,  $\gamma: \mathbb{B} \rightarrow \mathbb{L}$  be a Galois connection and let  $\log: \mathbb{L} \rightarrow \mathbb{L}$ ,  $\text{beh}: \mathbb{B} \rightarrow \mathbb{B}$  (both monotone).

- 1 Then  $\alpha \circ \log = \text{beh} \circ \alpha$  implies  $\alpha(\mu \log) = \mu \text{beh}$ .
- 2 Let  $c = \gamma \circ \alpha$  be the closure operator of the Galois connection and let  $\text{beh} = \alpha \circ \log \circ \gamma$ .  
Then  $c$ -compatibility of  $\log$  implies  $\alpha(\mu \log) = \mu \text{beh}$ .

$\mu$ : least fixpoint operator

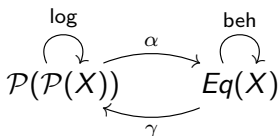
This theorem is well-known and goes back to work of Cousot & Cousot on abstract interpretation.



# Bisimilarity

We instantiate this framework and start with the simplest case: **bisimilarity** on labelled transition systems (with state space  $X$ ).

# Bisimilarity



$Eq(X)$ : set of all equivalences on  $X$ , ordered by  $\supseteq$

$$\alpha(\mathcal{S}) = \{(x_1, x_2) \in X \times X \mid \forall S \in \mathcal{S}: (x_1 \in S \Leftrightarrow x_2 \in S)\}$$

$$\gamma(R) = \{S \subseteq X \mid \forall (x_1, x_2) \in R: (x_1 \in S \Leftrightarrow x_2 \in S)\}$$

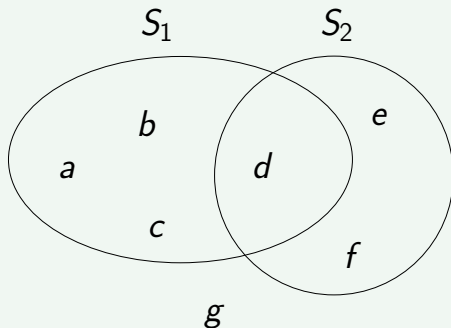
If  $\log \circ c \subseteq c \circ \log$  (where  $c = \gamma \circ \alpha$ ) and  $\text{beh} = \alpha \circ \log \circ \gamma$ :

$$\alpha(\mu \log) = \mu \text{beh}$$

( $\mu$ : least fixpoint. Contravariance!) This is the Hennessy-Milner theorem (logical equivalence = behavioural equivalence).

# Bisimilarity

Obtaining  $\alpha(\{S_1, S_2\})$  for  $S_1, S_2 \subseteq X = \{a, b, c, d, e, f, g\}$



# Bisimilarity

- **Logic function:**

$$\log(\mathcal{S}) = \bigcup_{a \in A} \diamond_a[cl_f(\mathcal{S})]$$

- $cl_f$  closes a set of sets under all finite boolean operations (empty conjunction: *true*, empty disjunction: *false*)
- $\diamond_a(\mathcal{S}) = \{x \in X \mid \exists y \in \mathcal{S}: x \xrightarrow{a} y\}$
- **Closure:**  $c = \alpha \circ \gamma$  closes a set of sets under all boolean operators  
 $\log$  is compatible with  $c$  if transition system is *finitely branching*
- **Behaviour function:** for  $R \in Eq(X)$

$$\begin{aligned} \text{beh}(R) &= \alpha(\log(\gamma(R))) = \\ &= \{(x_1, x_2) \mid \forall y_1: x_1 \xrightarrow{a} y_1 \exists y_2: x_2 \xrightarrow{a} y_2 \wedge (y_1, y_2) \in R \wedge \\ &\quad \forall y_2: x_2 \xrightarrow{a} y_2 \exists y_1: x_1 \xrightarrow{a} y_1 \wedge (y_1, y_2) \in R\} \end{aligned}$$

# Trace Metrics

Trace equivalence can be generalized to **trace metrics** [de Alfaro, Faella, Stoelinga] [Fahrenberg, Legay] that measures the distance between the sets of traces originating from two states.

Useful for systems with quantitative information (probabilities, weights, etc.) where behavioural equivalence is too strict.

Here: we generalize the trace inclusion preorder to a **directed trace metrics**.

First step: Extend transition systems with a metric  $d_A: A \times A \rightarrow [0, 1]$  on the label set  $A$ .

# Trace Metrics

## Preliminaries on metrics

- $DPMet(Y)$ : set of all **directed pseudo-metrics** on  $Y$ , i.e., functions  $d: Y \times Y \rightarrow [0, 1]$  such that
  - $d(y, y) = 0$  for all  $y \in Y$
  - $d(y_1, y_3) \leq d(y_1, y_2) + d(y_2, y_3)$  (triangle inequality) for all  $y_1, y_2, y_3 \in Y$ .
  - not necessarily symmetric ( $d(y_1, y_2) = d(y_2, y_1)$ )
- **Directed pseudometric space**: set  $Y$  with a directed pseudo-metric  $d$
- **Non-expansive functions** between pseudometric spaces  $(Y, d_Y), (Z, d_Z)$ : mapping  $f: Y \rightarrow Z$  with  $d_Z(f(y_1), f(y_2)) \leq d_Y(y_1, y_2)$  for all  $y_1, y_2 \in Y$ .

Addition and subtraction are modified to stay within  $[0,1]$

# Trace Metrics

## Directed Hausdorff metric

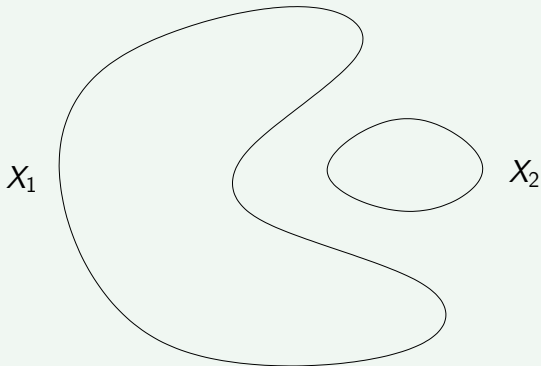
Lifting a directed metric space  $(X, d)$  to  $(\mathcal{P}(X), d^H)$ : let  $X_1, X_2 \subseteq X$ :

$$d^H(X_1, X_2) = \max_{x_1 \in X_1} \min_{x_2 \in X_2} d(x_1, x_2)$$

- For each element  $x_1 \in X_1$  take the closest element  $x_2 \in X_2$  and measure the distance  $d(x_1, x_2)$
- Take the maximum of all such distances.

# Trace Metrics

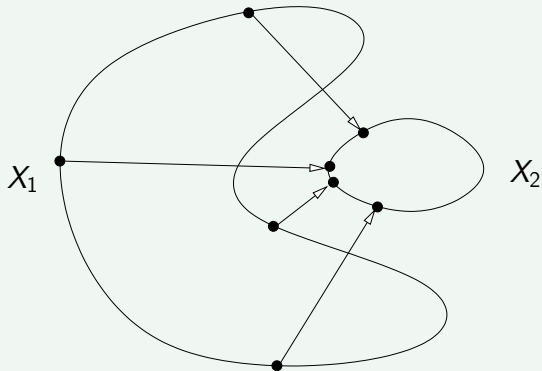
Example: Directed Hausdorff metric





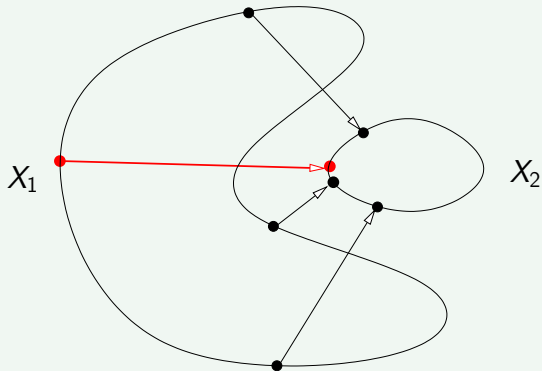
# Trace Metrics

## Example: Directed Hausdorff metric



# Trace Metrics

## Example: Directed Hausdorff metric



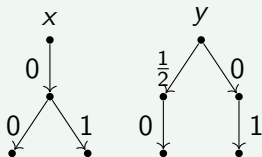
# Trace Metrics

## Trace Distance of two states $x, y$

- Let  $Tr(x) \subseteq A^*$  be the set of finite traces of  $x$ .
- The distance of two traces  $\sigma_1, \sigma_2$  is defined as
  - $d_{Tr}(\sigma_1, \sigma_2) = 1$  if  $|\sigma_1| \neq |\sigma_2|$
  - $d_{Tr}(\varepsilon, \varepsilon) = 0$
  - $d_{Tr}(a_1\sigma'_1, a_2\sigma'_2) = \max\{d_A(a_1, a_2), d_{Tr}(\sigma'_1, \sigma'_2)\}$   
(sup-metric)
- Given two states  $x, y$ :

$$d(x, y) = (d_{Tr})^H(Tr(x), Tr(y))$$

## Trace Metrics



For simplicity we restrict to traces of length 2:

$$d(x, y) = (d_{\text{Tr}})^H(\{00, 01\}, \{\frac{1}{2}0, 01\})$$

Hence:  $d(x, y) = \frac{1}{2}$ .

# Trace Metrics

We apply our recipe and use the following **Galois connection**:

$$\text{log} \left( \begin{array}{c} \curvearrowright \\ (\mathcal{P}([0, 1]^X), \subseteq) \end{array} \right) \begin{array}{c} \xrightarrow{\alpha} \\ \xleftarrow{\gamma} \end{array} \left( \begin{array}{c} \text{DPMet}(\mathcal{P}(X)), \leq \\ \curvearrowright \text{beh} \end{array} \right)$$

$$\alpha(\mathcal{F})(X_1, X_2) = \bigvee_{f \in \mathcal{F}} (\tilde{f}(X_1) - \tilde{f}(X_2))$$

$$\gamma(d) = \{f \in [0, 1]^X \mid \tilde{f} \text{ is non-expansive wrt. } d\}$$

$$\text{where } \tilde{f}: \mathcal{P}(X) \rightarrow [0, 1] \quad \tilde{f}(X') = \bigvee_{x \in X'} f(x) \\ (f: X \rightarrow [0, 1])$$

# Trace Metrics

Logic function:

$$\text{Modality: } \quad \bigcirc_a f(x) = \bigvee \{ \overline{D}_a(b) \wedge f(x') \mid x \xrightarrow{b} x' \}$$

where  $a \in A$ ,  $f: X \rightarrow [0, 1]$ ,  $\overline{D}_a(b) = 1 - d_A(b, a)$ .

$$\log(\mathcal{F}) = \bigcup_{a \in A} \bigcirc_a [c/\text{sh}(\mathcal{F})] \cup \{1\},$$

where  $c/\text{sh}$  closes a set of functions under constant shifts  
 $(f \mapsto f + c, f - c, c \in [0, 1])$ .

The logic function is compatible with the closure of the Galois connection (shifts are needed for compatibility).

# Trace Metrics

## Completeness:

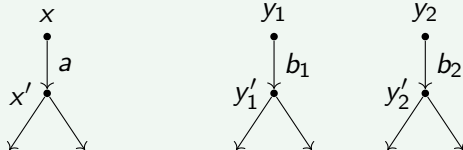
In order to convince ourselves that the logic is complete, we construct a **distinguishing formula**  $\varphi$ .

To obtain the trace distance of states  $x, y$ , take the trace  $a_1 \dots a_n \in Tr(x)$  of  $x$  that is farthest from any trace in  $Tr(y)$ .

Define  $\varphi = \bigcirc_{a_1} \cdots \bigcirc_{a_n} 1$

# Trace Metrics

Fixpoint function/equation (special case):  $\text{beh} = \alpha \circ \log \circ \gamma$



$$\begin{aligned} & \text{beh}(d)(\{x\}, \{y_1, y_2\}) \\ = & (d_A(a, b_1) \wedge d_A(a, b_2)) \vee (d_A(a, b_1) \wedge d(\{x'\}, \{y_2'\})) \\ & \vee (d_A(a, b_2) \wedge d(\{x'\}, \{y_1'\})) \vee d(\{x'\}, \{y_1', y_2'\}) \end{aligned}$$

This result depends on the fact that  $([0, 1], \leq)$  is a **distributive** lattice.



## Further Results

We can also handle . . .

- Preorders
- Behavioural metrics
- Decorated trace equivalences and metrics (completed traces, readiness, failure, etc.)

on labelled transition systems respectively metric transition systems.

# Coalgebra & Fibrations

**Future work:** generalization for coalgebras living in Eilenberg-Moore categories (with Lutz Schröder, Jonas Forster, Paul Wild)

It is our aim to show that the following three notions induce the same metric:

- behavioural metrics obtained in the graded monad setting
- graded modal logics
- least solution of a fixpoint equation defined on the determinized coalgebra

# Conclusion

## Related Work

- Fahrenberg, Legay, Thrane: Characterization of the metric linear-time/branching-time spectrum via games. Does not treat logics and fixpoint equations for trace metrics are different.
- Klin (e.g. in Klin's PhD thesis): different handling of the closure, does not treat behavioural metrics.
- Dual adjunction: functor on the “logic universe” characterizes the *syntax* of the logics rather than the *semantics*. Fibrational setup deviates from [Kupke, Rot].
- Approximating family [Komorida, Katsumata, Kupke, Rot, Hasuo]: related to our notion of compatibility.