

Can Computers Understand What is Happening?

Probabilistic Complex Event Recognition

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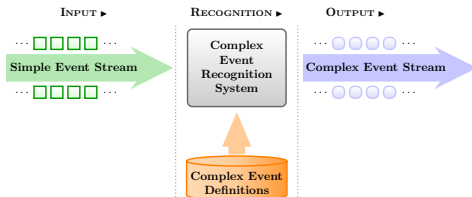
²University of Piraeus, Greece

³NKUA, Greece

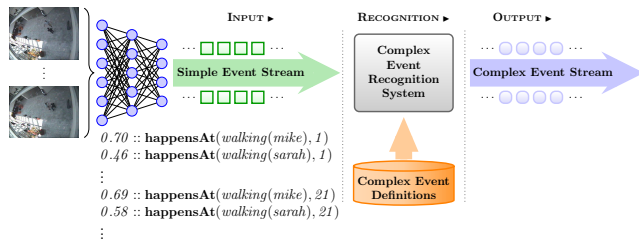
<https://cer.iit.demokritos.gr>



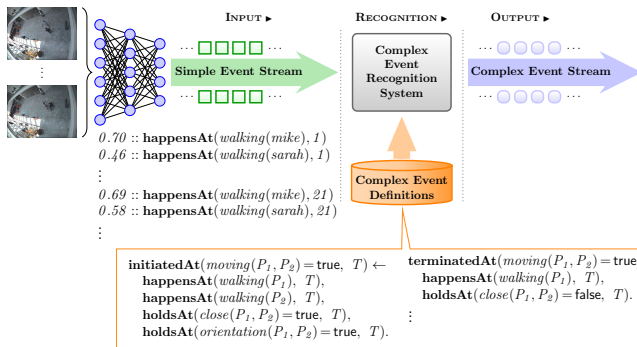
Complex Event Recognition under Uncertainty



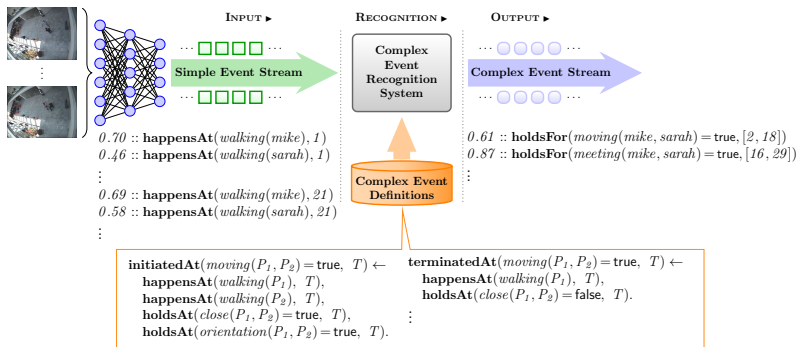
Complex Event Recognition under Uncertainty



Complex Event Recognition under Uncertainty



Complex Event Recognition under Uncertainty



Human Activity Recognition



<https://cer.iit.demokritos.gr> (activity recognition)

Event Calculus*

- A **logic programming language** for representing and reasoning about events and their effects.
- Key components:
 - **event** (typically instantaneous).
 - **fluent**: a property that may have different values at different points in time.

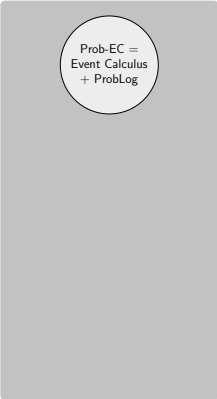
* Kowalski and Sergot, A Logic-based Calculus of Events. New Generation Computing, 1986.

Event Calculus*

- A **logic programming language** for representing and reasoning about events and their effects.
- Key components:
 - **event** (typically instantaneous).
 - **fluent**: a property that may have different values at different points in time.
- Built-in representation of **inertia**:
 - $F = V$ holds at a particular time-point if $F = V$ has been *initiated* by an event at some earlier time-point, and not *terminated* by another event in the meantime.

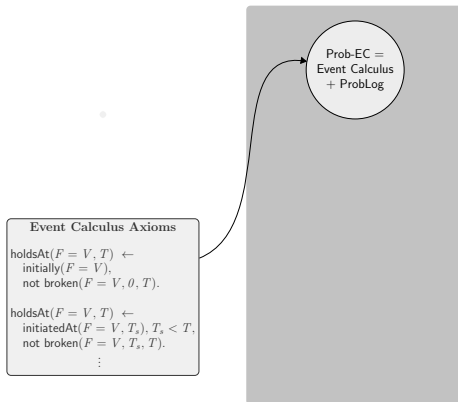
* Kowalski and Sergot, A Logic-based Calculus of Events. New Generation Computing, 1986.

Online Probabilistic Interval-Based Event Calculus

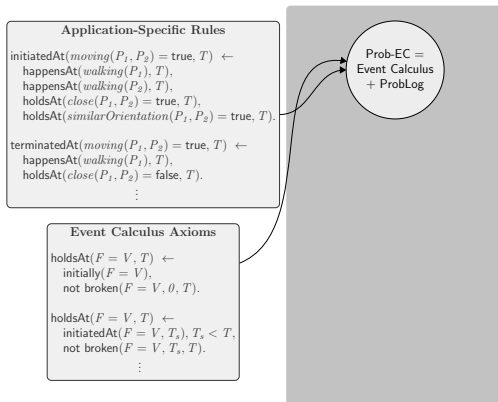


Prob-EC =
Event Calculus
+ ProbLog

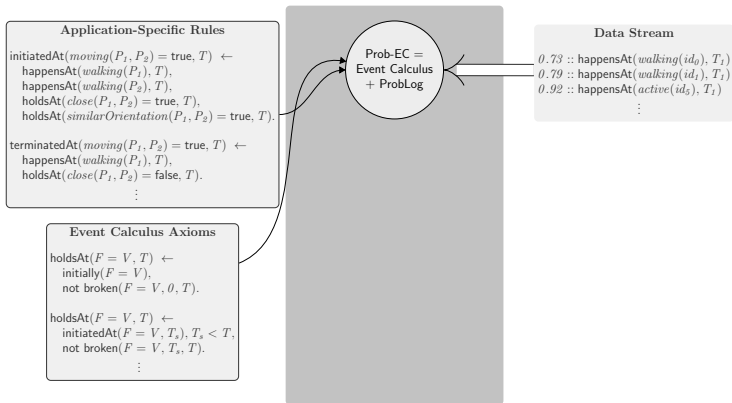
Online Probabilistic Interval-Based Event Calculus



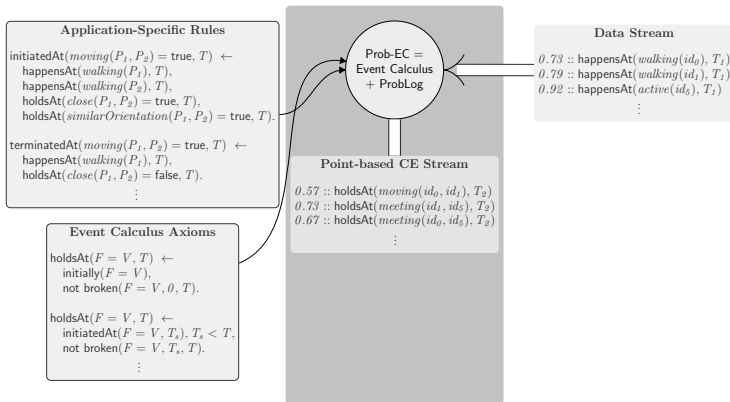
Online Probabilistic Interval-Based Event Calculus



Online Probabilistic Interval-Based Event Calculus



Online Probabilistic Interval-Based Event Calculus



Instantaneous Recognition

initiatedAt($moving(P_1, P_2) = \text{true}, T$) $\leftarrow$
 happensAt($walking(P_1), T$),
 happensAt($walking(P_2), T$),
 holdsAt($close(P_1, P_2) = \text{true}, T$),
 holdsAt($orientation(P_1, P_2) = \text{true}, T$).

terminatedAt($moving(P_1, P_2) = \text{true}, T$) $\leftarrow$
 happensAt($walking(P_1), T$),
 holdsAt($close(P_1, P_2) = \text{false}, T$).

0.70 :: **happensAt**($walking(mike), 1$).
0.46 :: **happensAt**($walking(sarah), 1$).

Instantaneous Recognition

initiatedAt(*moving*(P_1, P_2) = true, T) $\leftarrow$
 happensAt(*walking*(P_1), T),
 happensAt(*walking*(P_2), T),
 holdsAt(*close*(P_1, P_2) = true, T),
 holdsAt(*orientation*(P_1, P_2) = true, T).
terminatedAt(*moving*(P_1, P_2) = true, T) $\leftarrow$
 happensAt(*walking*(P_1), T),
 holdsAt(*close*(P_1, P_2) = false, T).

0.70 :: **happensAt**(*walking*(*mike*), 1).

0.46 :: **happensAt**(*walking*(*sarah*), 1).

$P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 1)) =$
 $P(\text{happensAt}(\text{walking}(\text{mike}), 1)) \times$
 $P(\text{happensAt}(\text{walking}(\text{sarah}), 1)) \times$
 $P(\text{holdsAt}(\text{close}(\text{mike}, \text{sarah}) = \text{true}, 1)) \times$
 $P(\text{holdsAt}(\text{orientation}(\text{mike}, \text{sarah}) = \text{true}, 1))$
 $= 0.7 \times 0.46 \times 1 \times 1 = 0.322$

Instantaneous Recognition

initiatedAt(*moving*(P_1, P_2) = true, T) $\leftarrow$
 happensAt(*walking*(P_1), T),
 happensAt(*walking*(P_2), T),
 holdsAt(*close*(P_1, P_2) = true, T),
 holdsAt(*orientation*(P_1, P_2) = true, T).
terminatedAt(*moving*(P_1, P_2) = true, T) $\leftarrow$
 happensAt(*walking*(P_1), T),
 holdsAt(*close*(P_1, P_2) = false, T).

0.70 :: **happensAt**(*walking*(*mike*), 1).
0.46 :: **happensAt**(*walking*(*sarah*), 1).

$P(\text{holdsAt}(CE = \text{true}, t)) =$
 $P(\text{initiatedAt}(CE = \text{true}, t-1) \vee$
 $(\text{holdsAt}(CE = \text{true}, t-1) \wedge$
 $\neg \text{terminatedAt}(CE = \text{true}, t-1)))$

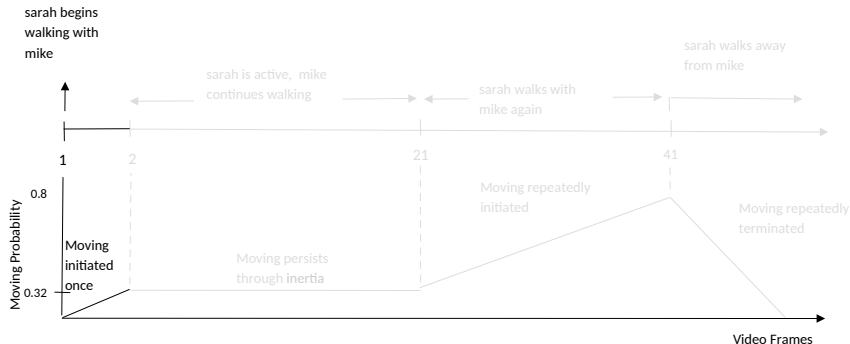
Instantaneous Recognition

initiatedAt(*moving*(P_1, P_2) = true, T) $\leftarrow$
 happensAt(*walking*(P_1), T),
 happensAt(*walking*(P_2), T),
 holdsAt(*close*(P_1, P_2) = true, T),
 holdsAt(*orientation*(P_1, P_2) = true, T).
terminatedAt(*moving*(P_1, P_2) = true, T) $\leftarrow$
 happensAt(*walking*(P_1), T),
 holdsAt(*close*(P_1, P_2) = false, T).

0.70 :: **happensAt**(*walking*(*mike*), 1).
0.46 :: **happensAt**(*walking*(*sarah*), 1).

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 2)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 1) \vee$
 ($\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 1) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 1)))$
 $= 0.322 + 0 \times 1 - 0.322 \times 0 \times 1 = 0.322$

Instantaneous Recognition

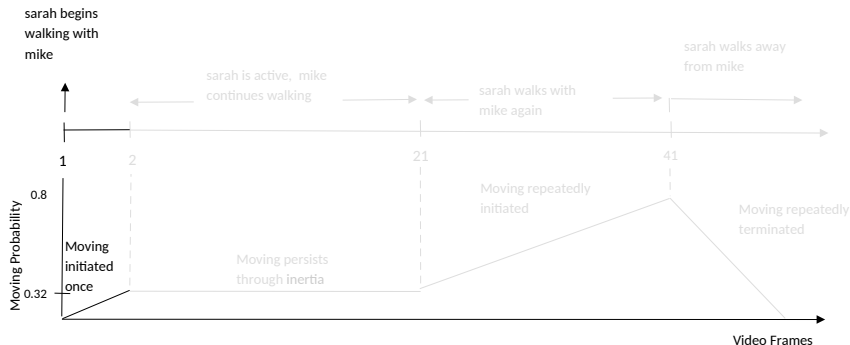


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$
 $\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.70 :: \text{happensAt}(\text{walking}(\text{mike}), 1).$
 $0.46 :: \text{happensAt}(\text{walking}(\text{sarah}), 1).$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 2)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 1) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 1) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 1)))$
 $= 0.322 + 0 \times 1 - 0.322 \times 0 \times 1 = 0.322$

Instantaneous Recognition

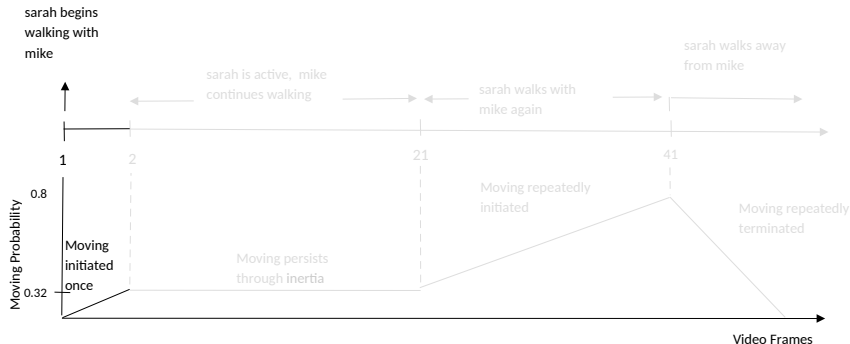


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.73 :: \text{happensAt}(\text{walking}(\text{mike}), 2).$
 $0.55 :: \text{happensAt}(\text{active}(\text{sarah}), 2). \dots$

Instantaneous Recognition



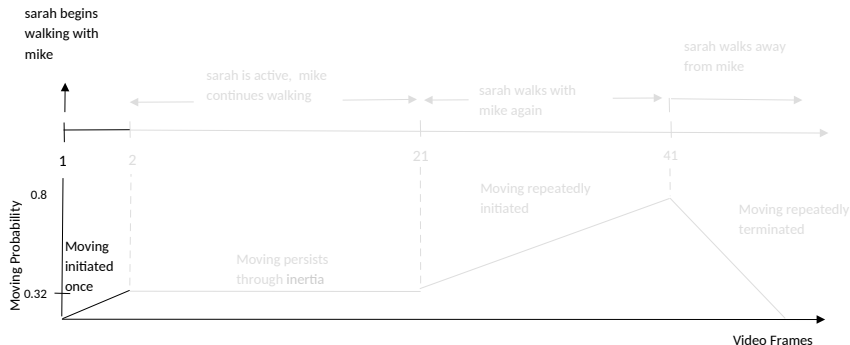
$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.73 :: \text{happensAt}(\text{walking}(\text{mike}), 2).$
 $0.55 :: \text{happensAt}(\text{active}(\text{sarah}), 2). \dots$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 3)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 2) \vee$
 $\quad (\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 2) \wedge$
 $\quad \neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 2)))$
 $= 0 + 0.322 \times 1 - 0 \times 0.322 \times 1 = 0.322$

Instantaneous Recognition



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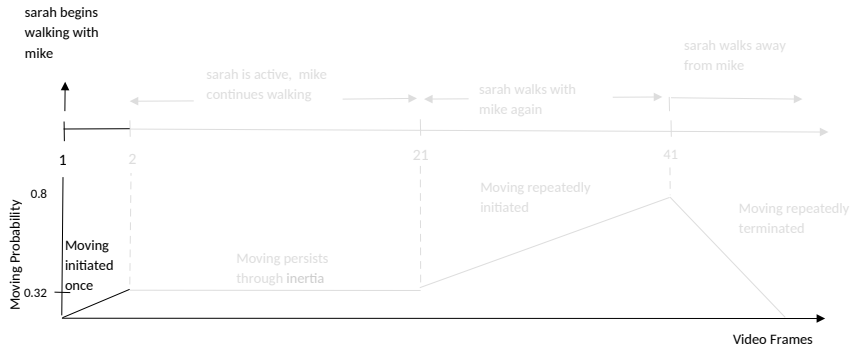
initiatedAt(moving( $P_1, P_2$ ) = true,  $T$ )  $\leftarrow$ 
  happensAt(walking( $P_1$ ),  $T$ ),
  happensAt(walking( $P_2$ ),  $T$ ),
  holdsAt(close( $P_1, P_2$ ) = true,  $T$ ),
  holdsAt(orientation( $P_1, P_2$ ) = true,  $T$ ).

terminatedAt(moving( $P_1, P_2$ ) = true,  $T$ )  $\leftarrow$ 
  happensAt(walking( $P_1$ ),  $T$ ),
  holdsAt(close( $P_1, P_2$ ) = false,  $T$ ).
  
```

```

0.45 :: happensAt(walking(mike), 20).
0.14 :: happensAt(active(sarah), 20).
  
```

Instantaneous Recognition



$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

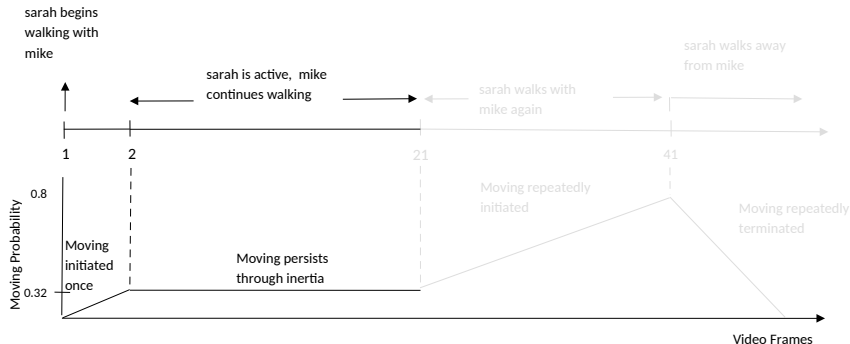
$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.45 :: \text{happensAt}(\text{walking}(\text{mike}), 20).$

$0.14 :: \text{happensAt}(\text{active}(\text{sarah}), 20).$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 21)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 20) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 20) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 20)))$
 $= 0 + 0.322 \times 1 - 0 \times 0.322 \times 1 = 0.322$

Instantaneous Recognition



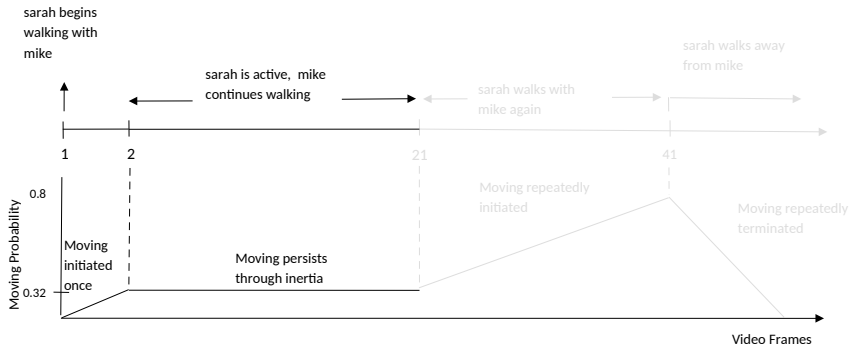
$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$
 $\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.45 :: \text{happensAt}(\text{walking}(\text{mike}), 20).$

$0.14 :: \text{happensAt}(\text{active}(\text{sarah}), 20).$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 21)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 20) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 20) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 20)))$
 $= 0 + 0.322 \times 1 - 0 \times 0.322 \times 1 = 0.322$

Instantaneous Recognition

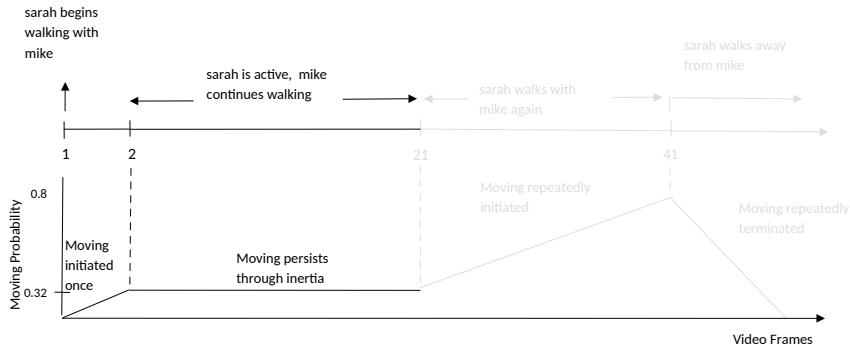


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.39 :: \text{happensAt}(\text{walking}(\text{mike}), 21).$
 $0.28 :: \text{happensAt}(\text{walking}(\text{sarah}), 21). \dots$

Instantaneous Recognition



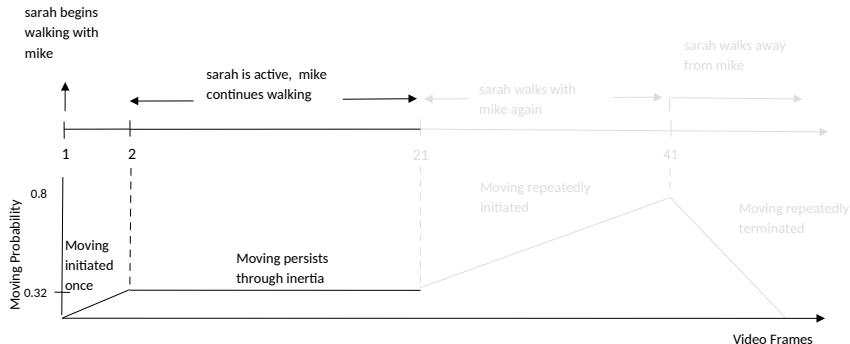
$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.39 :: \text{happensAt}(\text{walking}(\text{mike}), 21).$
 $0.28 :: \text{happensAt}(\text{walking}(\text{sarah}), 21). \dots$

$P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 21)) =$
 $P(\text{happensAt}(\text{walking}(\text{mike}), 21)) \times$
 $P(\text{happensAt}(\text{walking}(\text{sarah}), 21)) \times$
 $P(\text{holdsAt}(\text{close}(\text{mike}, \text{sarah}) = \text{true}, 21)) \times$
 $P(\text{holdsAt}(\text{orientation}(\text{mike}, \text{sarah}) = \text{true}, 21))$
 $= 0.39 \times 0.28 \times 1 \times 1 = 0.11$

Instantaneous Recognition

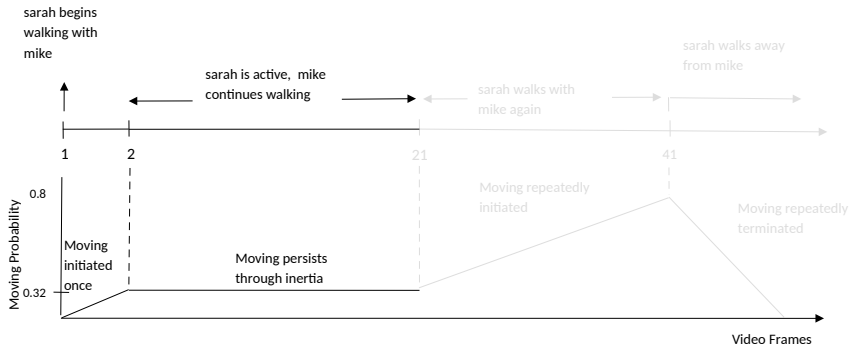


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$
 $\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.39 :: \text{happensAt}(\text{walking}(\text{mike}), 21).$
 $0.28 :: \text{happensAt}(\text{walking}(\text{sarah}), 21). \dots$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 22)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 21) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 21) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 21)))$
 $= 0.11 + 0.322 \times 1 - 0.11 \times 0.322 \times 1 = 0.39$

Instantaneous Recognition



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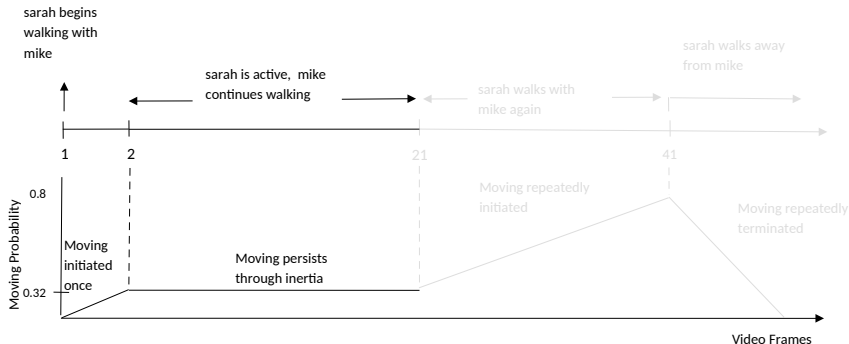
initiatedAt(moving( $P_1, P_2$ ) = true,  $T$ )  $\leftarrow$ 
  happensAt(walking( $P_1$ ),  $T$ ),
  happensAt(walking( $P_2$ ),  $T$ ),
  holdsAt(close( $P_1, P_2$ ) = true,  $T$ ),
  holdsAt(orientation( $P_1, P_2$ ) = true,  $T$ ).

terminatedAt(moving( $P_1, P_2$ ) = true,  $T$ )  $\leftarrow$ 
  happensAt(walking( $P_1$ ),  $T$ ),
  holdsAt(close( $P_1, P_2$ ) = false,  $T$ ).
  
```

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0.28 :: happensAt(walking(mike), 40).
0.18 :: happensAt(walking(sarah), 40).
  
```

Instantaneous Recognition



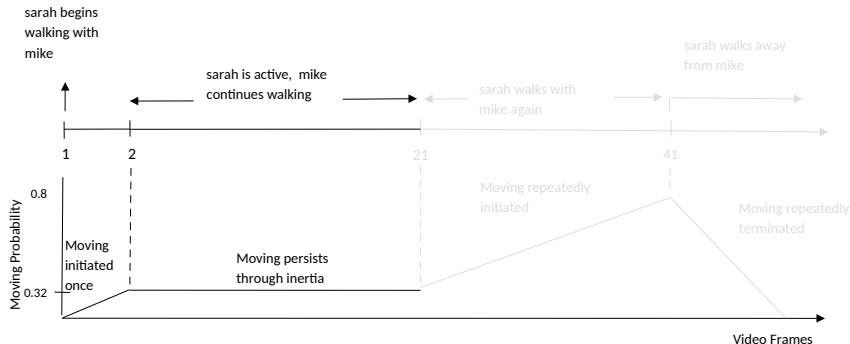
$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.28 :: \text{happensAt}(\text{walking}(\text{mike}), 40).$
 $0.18 :: \text{happensAt}(\text{walking}(\text{sarah}), 40).$

$P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 40)) =$
 $P(\text{happensAt}(\text{walking}(\text{mike}), 40)) \times$
 $P(\text{happensAt}(\text{walking}(\text{sarah}), 40)) \times$
 $P(\text{holdsAt}(\text{close}(\text{mike}, \text{sarah}) = \text{true}, 40)) \times$
 $P(\text{holdsAt}(\text{orientation}(\text{mike}, \text{sarah}) = \text{true}, 40))$
 $= 0.28 \times 0.18 \times 1 \times 1 = 0.05$

Instantaneous Recognition



$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

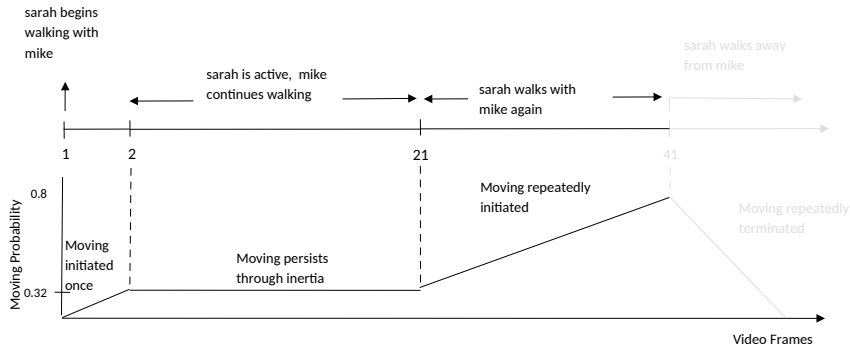
$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.28 :: \text{happensAt}(\text{walking}(\text{mike}), 40).$

$0.18 :: \text{happensAt}(\text{walking}(\text{sarah}), 40).$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 41)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 40) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 40) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 40)))$
 $= 0.05 + 0.79 \times 1 - 0.05 \times 0.79 \times 1 = 0.80$

Instantaneous Recognition

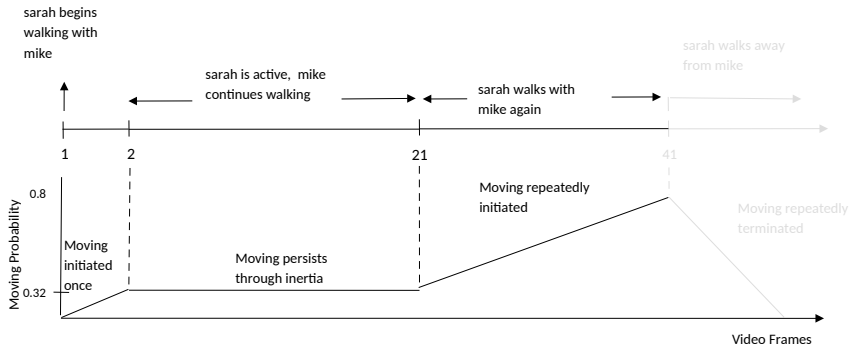


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$
 $\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.28 :: \text{happensAt}(\text{walking}(\text{mike}), 40).$
 $0.18 :: \text{happensAt}(\text{walking}(\text{sarah}), 40).$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 41)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 40) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 40) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 40)))$
 $= 0.05 + 0.79 \times 1 - 0.05 \times 0.79 \times 1 = 0.80$

Instantaneous Recognition

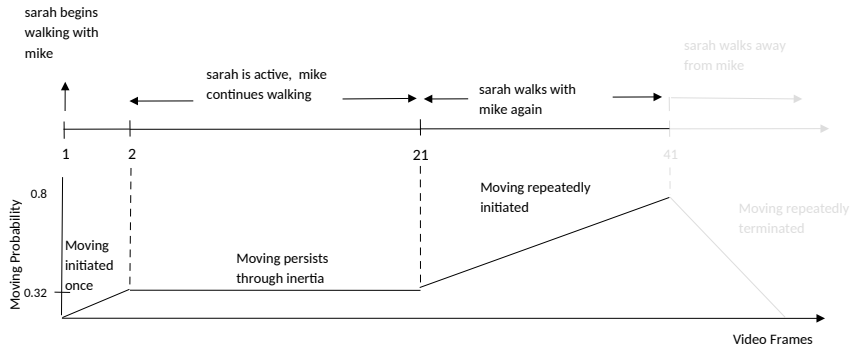


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.18 :: \text{happensAt}(\text{walking}(\text{mike}), 41).$
 $0.79 :: \text{happensAt}(\text{inactive}(\text{sarah}), 41). \dots$

Instantaneous Recognition



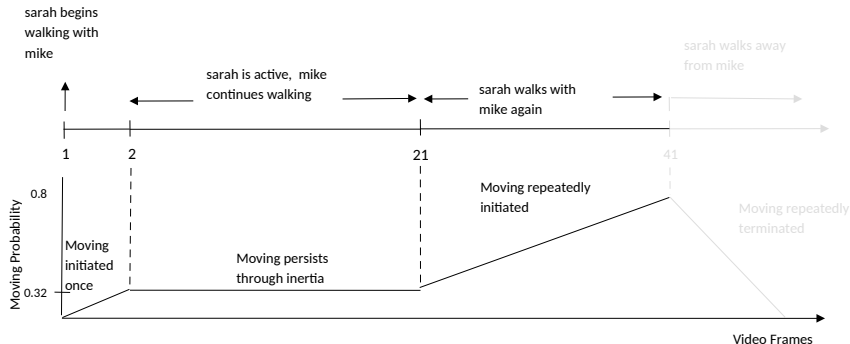
$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.18 :: \text{happensAt}(\text{walking}(\text{mike}), 41).$
 $0.79 :: \text{happensAt}(\text{inactive}(\text{sarah}), 41). \dots$

$P(\text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 41)) =$
 $P(\text{happensAt}(\text{walking}(\text{mike}), 41)) \times$
 $P(\text{holdsAt}(\text{close}(\text{mike}, \text{sarah}) = \text{false}, 41))$
 $= 0.18 \times 1 = 0.18$

Instantaneous Recognition

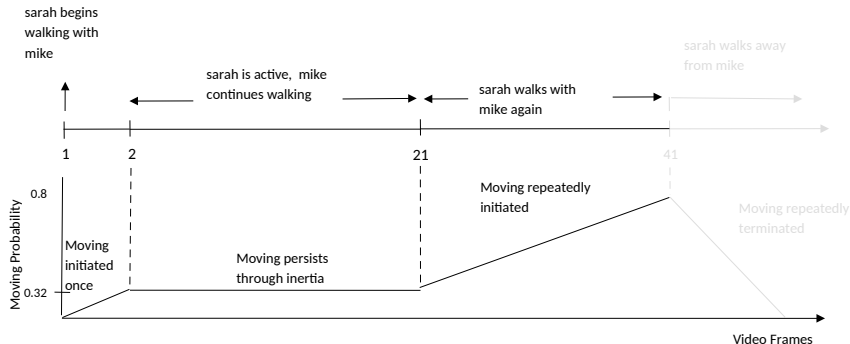


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$
 $\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$0.18 :: \text{happensAt}(\text{walking}(\text{mike}), 41).$
 $0.79 :: \text{happensAt}(\text{inactive}(\text{sarah}), 41). \dots$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 42)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 41) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 41) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 41)))$
 $= 0 + 0.8 \times (1 - 0.18) - 0 \times 0.8 \times (1 - 0.18) = 0.66$

Instantaneous Recognition



```

initiatedAt(moving( $P_1, P_2$ ) = true,  $T$ )  $\leftarrow$ 
  happensAt(walking( $P_1$ ),  $T$ ),
  happensAt(walking( $P_2$ ),  $T$ ),
  holdsAt(close( $P_1, P_2$ ) = true,  $T$ ),
  holdsAt(orientation( $P_1, P_2$ ) = true,  $T$ ).
  
```

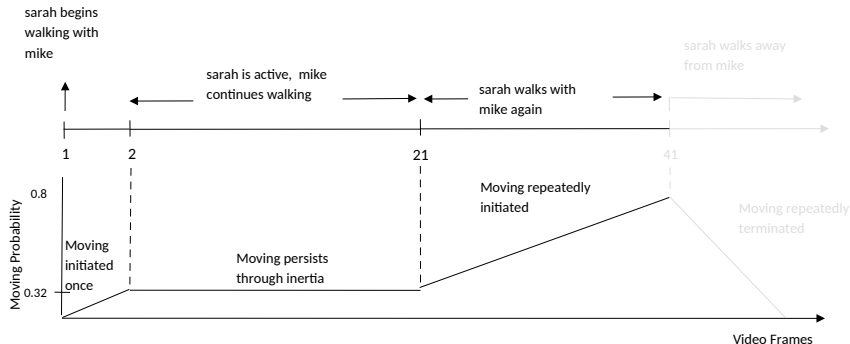
```

terminatedAt(moving( $P_1, P_2$ ) = true,  $T$ )  $\leftarrow$ 
  happensAt(walking( $P_1$ ),  $T$ ),
  holdsAt(close( $P_1, P_2$ ) = false,  $T$ ).
  
```

```

1.00 :: happensAt(walking(mike), 49).
0.96 :: happensAt(inactive(sarah), 49).
  
```

Instantaneous Recognition



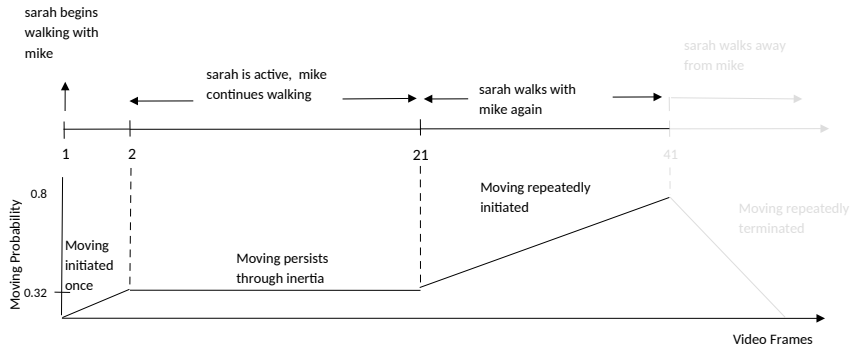
$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$1.00 :: \text{happensAt}(\text{walking}(\text{mike}), 49).$
 $0.96 :: \text{happensAt}(\text{inactive}(\text{sarah}), 49).$

$P(\text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 49)) =$
 $P(\text{happensAt}(\text{walking}(\text{mike}), 49)) \times$
 $P(\text{holdsAt}(\text{close}(\text{mike}, \text{sarah}) = \text{false}, 49))$
 $= 1 \times 1 = 1$

Instantaneous Recognition



$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$

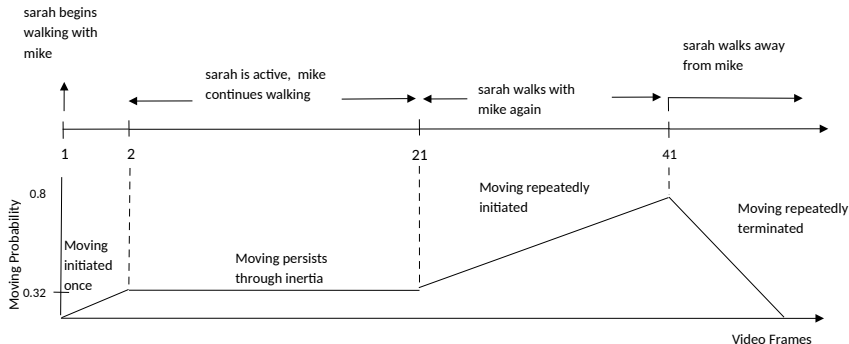
$\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$1.00 :: \text{happensAt}(\text{walking}(\text{mike}), 49).$

$0.96 :: \text{happensAt}(\text{inactive}(\text{sarah}), 49).$

$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 50)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 49) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 49) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 49)))$
 $= 0 + 0.07 \times 0 - 0 \times 0.07 \times 0 = 0$

Instantaneous Recognition

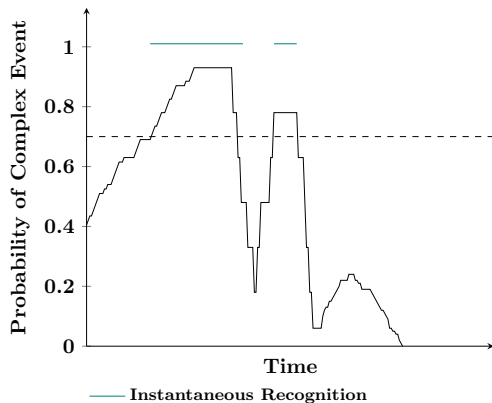


$\text{initiatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{happensAt}(\text{walking}(P_2), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{true}, T),$
 $\text{holdsAt}(\text{orientation}(P_1, P_2) = \text{true}, T).$
 $\text{terminatedAt}(\text{moving}(P_1, P_2) = \text{true}, T) \leftarrow$
 $\text{happensAt}(\text{walking}(P_1), T),$
 $\text{holdsAt}(\text{close}(P_1, P_2) = \text{false}, T).$

$1.00 :: \text{happensAt}(\text{walking}(\text{mike}), 49).$
 $0.96 :: \text{happensAt}(\text{inactive}(\text{sarah}), 49).$

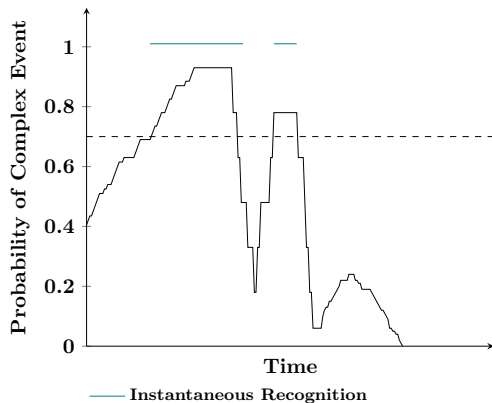
$P(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 50)) =$
 $P(\text{initiatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 49) \vee$
 $(\text{holdsAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 49) \wedge$
 $\neg \text{terminatedAt}(\text{moving}(\text{mike}, \text{sarah}) = \text{true}, 49)))$
 $= 0 + 0.07 \times 0 - 0 \times 0.07 \times 0 = 0$

Instantaneous Recognition*



*Skarlatidis et al, A Probabilistic Logic Programming Event Calculus. Theory & Practice of Logic Programming, 2015.

Instantaneous Recognition*

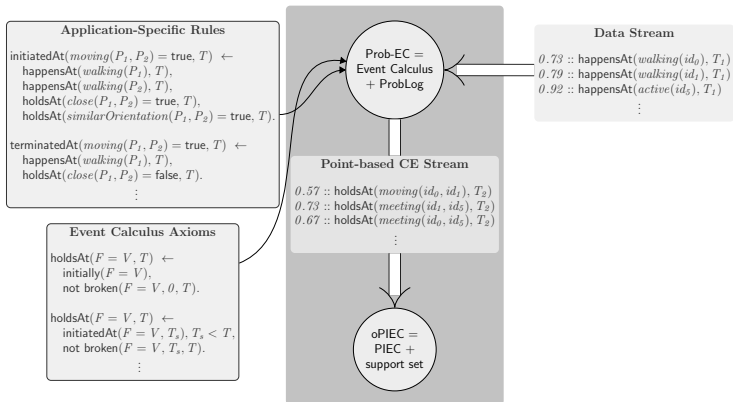


Higher accuracy than crisp reasoning in the presence of:

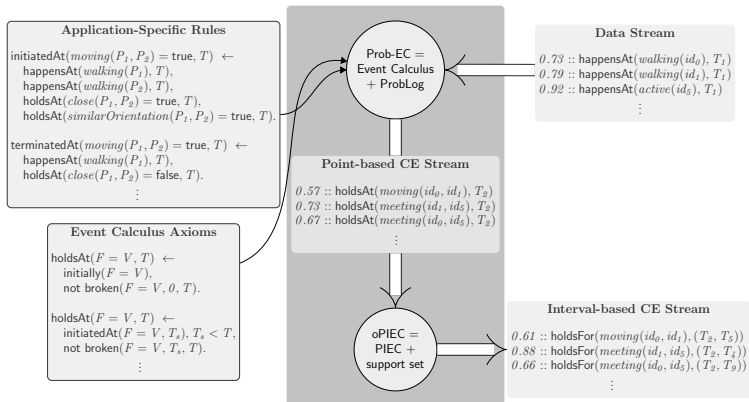
- several initiations and terminations;
- few probabilistic conjuncts.

*Skarlatidis et al, A Probabilistic Logic Programming Event Calculus. Theory & Practice of Logic Programming, 2015.

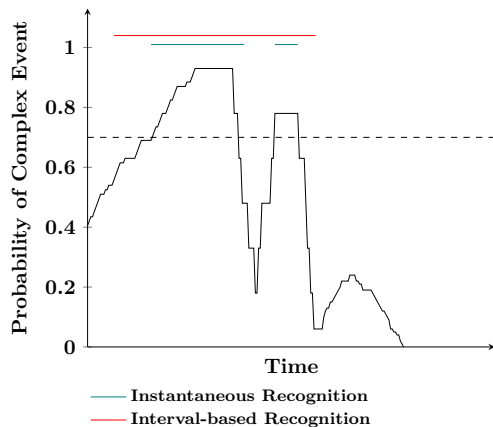
Online Probabilistic Interval-Based Event Calculus



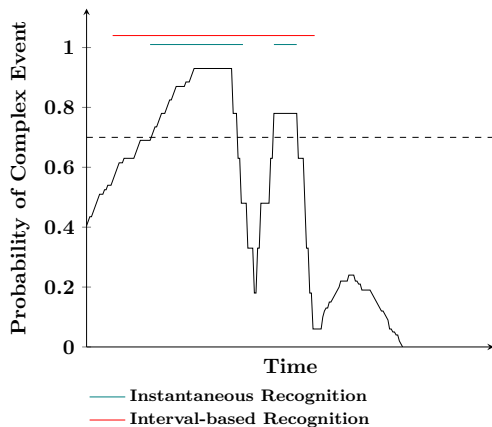
Online Probabilistic Interval-Based Event Calculus



Instantaneous vs Interval-based Recognition

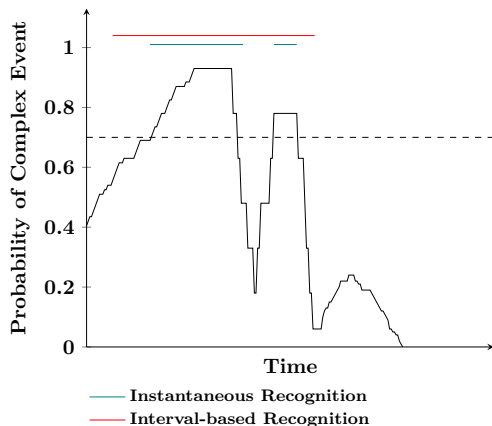


Instantaneous vs Interval-based Recognition



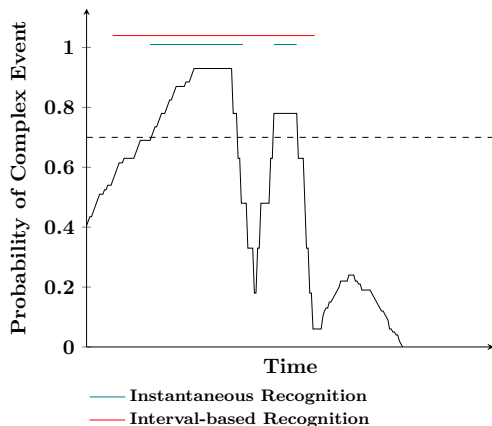
- **Interval Probability:** average probability of the time-points it contains.

Instantaneous vs Interval-based Recognition



- **Interval Probability:** average probability of the time-points it contains.
- **Probabilistic Maximal Interval:**
 - interval probability above a given threshold;
 - no super-interval with probability above the threshold.

Instantaneous vs Interval-based Recognition



- **Interval Probability:** average probability of the time-points it contains.
- **Probabilistic Maximal Interval:**
 - interval probability above a given threshold;
 - no super-interval with probability above the threshold.
- Probabilistic maximal interval computation via **maximal non-negative sum interval** computation.

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
ln	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
L	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5

$$L[i] = ln[i] - \mathcal{T}$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
l_n	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
L	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5

$$\sum_{i=s}^e L[i] \geq 0 \Leftrightarrow P([s, e]) \geq \mathcal{T}$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9

$$prefix[i] = \sum_{j=1}^i L[j]$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>										

$$dp[i] = \max_{i \leq j \leq n} (prefix[j])$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>										-0.9

$$dp[10] = \max_{10 \leq j \leq 10} (prefix[j])$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>									-0.9	-0.9

$$dp[9] = \max_{9 \leq j \leq 10} (prefix[j])$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>								-0.9	-0.9	-0.9

$$dp[8] = \max_{8 \leq j \leq 10} (prefix[j])$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>							-0.9	-0.9	-0.9	-0.9

$$dp[7] = \max_{7 \leq j \leq 10} (prefix[j])$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>						-0.4	-0.9	-0.9	-0.9	-0.9

$$dp[6] = \max_{6 \leq j \leq 10} (prefix[j])$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dp[i] = \max_{i \leq j \leq 10} (prefix[j])$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[s, e] = \begin{cases} dp[e] - prefix[s-1] & \text{if } s > 1 \\ dp[e] & \text{if } s = 1 \end{cases}$$

Interval-based Recognition

Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[s, e] = \begin{cases} dp[e] - prefix[s-1] & \text{if } s > 1 \\ dp[e] & \text{if } s = 1 \end{cases}$$

$$dprange[s, e] \geq 0 \Rightarrow \exists e^* : e^* \geq e, P([s, e^*]) \geq \mathcal{T}$$

Interval-based Recognition

	$\uparrow\downarrow$									
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

Interval-based Recognition

	$\uparrow\downarrow$									
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[1, 1] = dp[1] = 0.1 \geq 0$$

Interval-based Recognition

	$\uparrow$	$\downarrow$								
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

Interval-based Recognition

	$\uparrow$	$\downarrow$								
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[1, 2] = dp[2] = 0.1 \geq 0$$

Interval-based Recognition

	$\uparrow$		$\downarrow$							
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[1, 3] = dp[3] = 0.1 \geq 0$$

Interval-based Recognition

	$\uparrow$			$\downarrow$						
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[1, 4] = dp[4] = 0.1 \geq 0$$

Interval-based Recognition

	$\uparrow$				$\downarrow$					
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[1, 5] = dp[5] = 0 \geq 0$$

Interval-based Recognition

	$\uparrow$					$\downarrow$				
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[1, 6] = dp[6] = -0.4 < 0$$

Interval-based Recognition

	$\uparrow$					$\downarrow$				
Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[1, 6] = dp[6] = -0.4 < 0$$

Interval-based Recognition

		↑				↓				
Time	1	2	3	4	5	6	7	8	9	10
<i>ln</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[2, 6] = dp[6] - prefix[1] = 0.1 \geq 0$$

Interval-based Recognition

		↑					↓			
Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[2, 7] = dp[7] - prefix[1] = -0.4 < 0$$

Interval-based Recognition



Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

$$dprange[2, 7] = dp[7] - prefix[1] = -0.4 < 0$$

Interval-based Recognition



Time	1	2	3	4	5	6	7	8	9	10
<i>In</i>	0	0.5	0.7	0.9	0.4	0.1	0	0	0.5	1
<i>L</i>	-0.5	0	0.2	0.4	-0.1	-0.4	-0.5	-0.5	0	0.5
<i>prefix</i>	-0.5	-0.5	-0.3	0.1	0	-0.4	-0.9	-1.4	-1.4	-0.9
<i>dp</i>	0.1	0.1	0.1	0.1	0	-0.4	-0.9	-0.9	-0.9	-0.9

Interval-based Recognition*

Interval Computation Correctness

An interval is computed iff it is a probabilistic maximal interval.

* Artikis et al, A Probabilistic Interval-based Event Calculus for Activity Recognition. Annals of Mathematics and Artificial Intelligence, 2021.

Interval-based Recognition*

Interval Computation Correctness

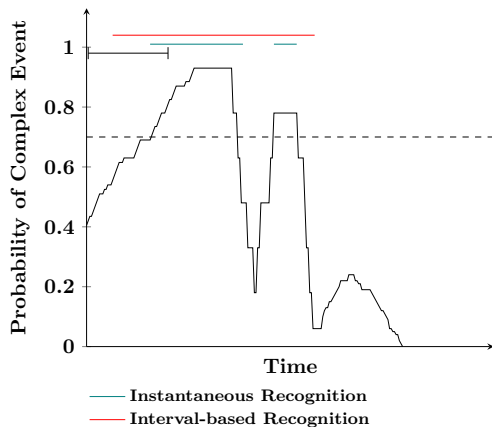
An interval is computed iff it is a probabilistic maximal interval.

Complexity

The computation of probabilistic maximal intervals is linear to the dataset size.

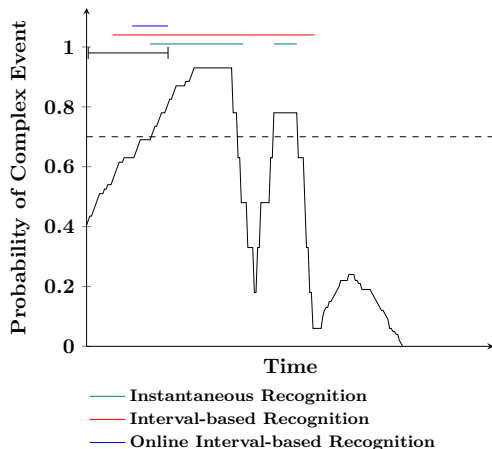
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Online Interval-based Recognition



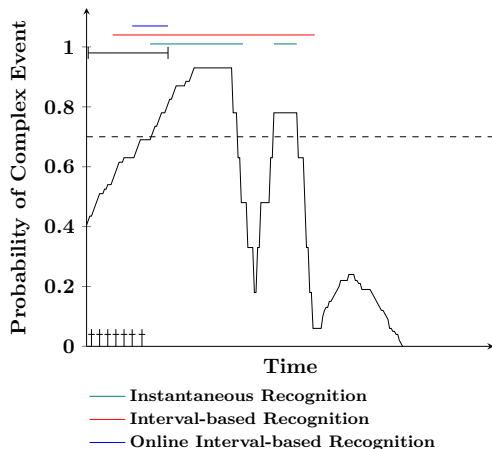
- Windowing.

Online Interval-based Recognition



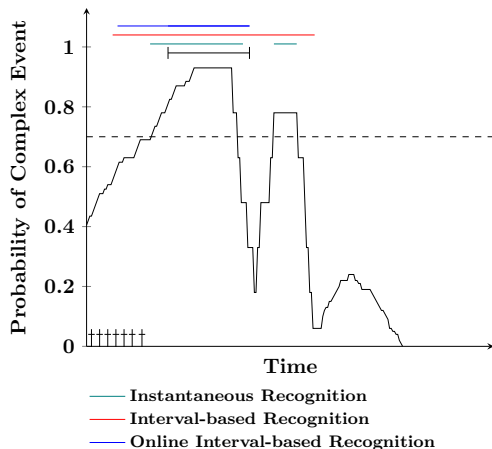
- Windowing.
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Online Interval-based Recognition



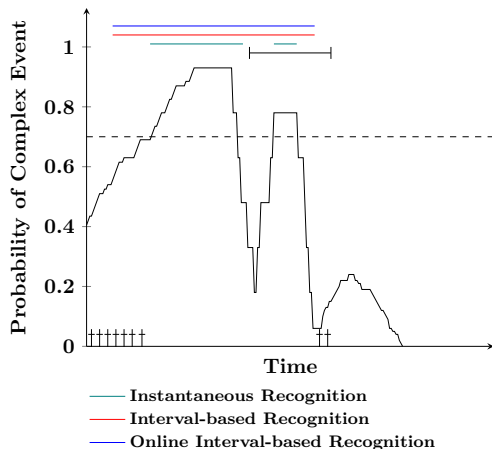
- Windowing.
- Probabilistic maximal interval computation.
- Caching **potential starting points**.
 - Discard time-point t iff there is a $t' < t$ that can be the starting point of a probabilistic maximal interval including t .

Online Interval-based Recognition



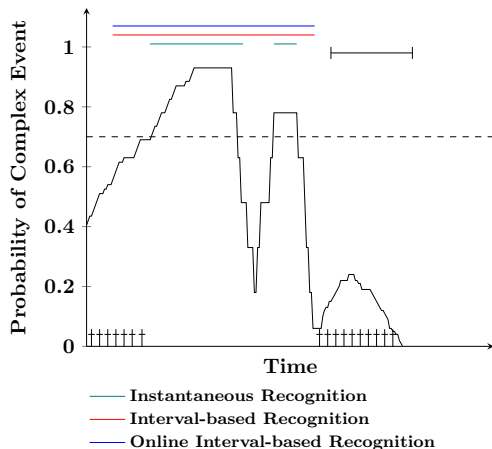
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Memory Minimality

A time-point is cached iff it may be the starting point of a future probabilistic maximal interval.

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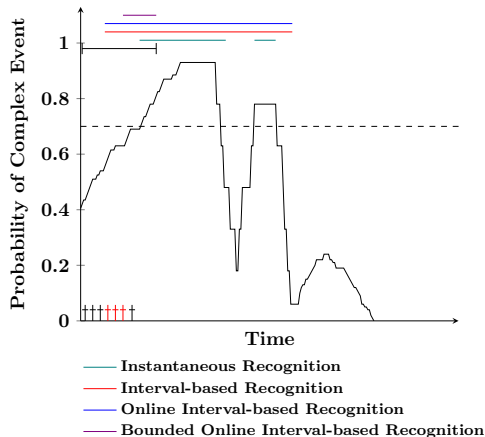
Interval Computation Correctness

An interval is computed iff it is a probabilistic maximal interval given the data seen so far.

Complexity

The computation of probabilistic maximal intervals is linear to the window and memory size.

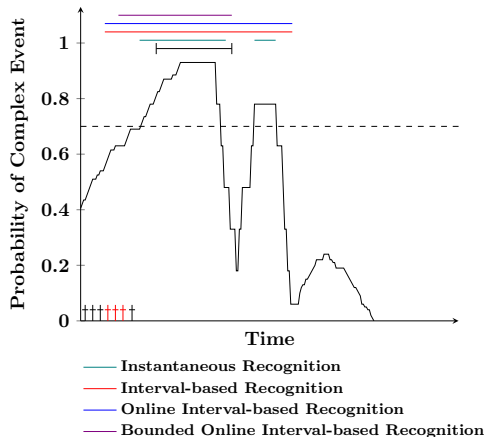
Bounded Online Interval-based Recognition*



- Complex event duration statistics favor more recent potential starting points.

* Mantenoglou et al, Online Event Recognition over Noisy Data Streams. International Journal of Approximate Reasoning, 2023. <https://github.com/Periklismant/oPIEC>

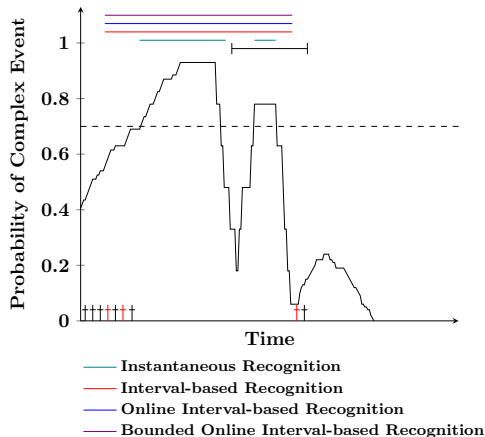
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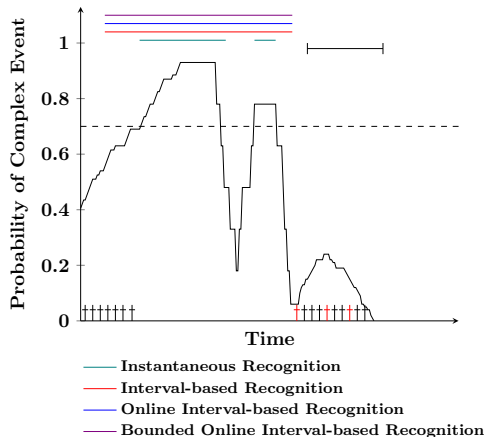
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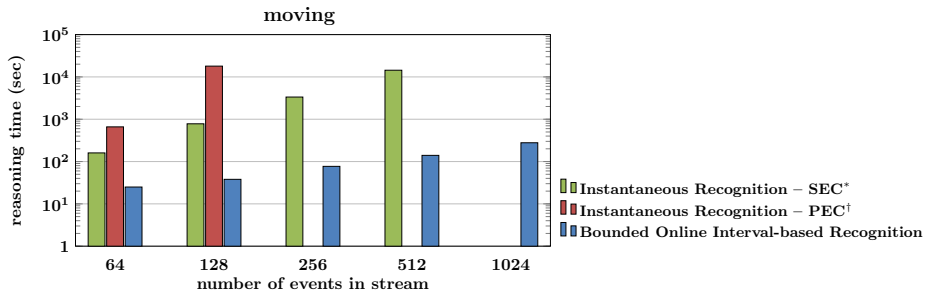
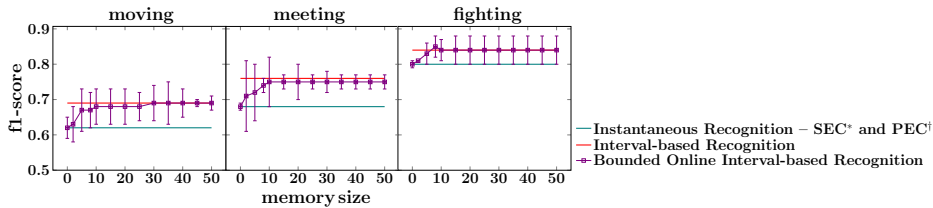
Bounded Online Interval-based Recognition*



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- Comparable accuracy to batch reasoning.

* Mantenoglou et al, Online Event Recognition over Noisy Data Streams. International Journal of Approximate Reasoning, 2023. <https://github.com/Periklismant/oPIEC>

Indicative Experimental Results



* McAreevey et al., The event calculus in probabilistic logic programming with annotated disjunctions. AAMAS, 2017.

† D'Asaro et al., Probabilistic reasoning about epistemic action narratives. Artificial Intelligence, 2021.

Summary

Complex event recognition over noisy streams:

- Probabilistic reasoning $\rightarrow$ robust complex event recognition.

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Next:

- **Forecast** complex events.