Can Computers Understand what is Happening? The Run-Time Event Calculus

> Alexander Artikis^{1,2} Periklis Mantenoglou^{1,3}

¹NCSR Demokritos, Athens, Greece ²University of Piraeus, Greece ³NKUA, Greece

https://cer.iit.demokritos.gr







Complex Event Recognition



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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.

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 - 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.

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^{*}Michelioudakis et al, Online Semi-Supervised Learning of Composite Event Rules by Combining Structure and Mass-based Predicate Similarity. Machine Learning, 2024.

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- Various implementation routes[†].

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[†]Tsilionis et al, A Tensor-based Formalisation of the Event Calculus. IJCAI, 2024.

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where

. . .

conditions: $\begin{array}{ll} 0^{-K} happensAt(E_k, T), \\ 0^{-M} holdsAt(F_m = V_m, T), \\ 0^{-N} a temporal constraint_n \end{array}$

^{*}Artikis et al, An Event Calculus for Event Recognition. IEEE TKDE, 2015. https://github.com/aartikis/RTEC

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holdsFor(F = V, I)



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terminatedAt(*highSpeedNC*(*Vessel*) = true, T) \leftarrow happensAt(*end*(*withinArea*(*Vessel*, *nearCoast*) = true), T).



https://cer.iit.demokritos.gr (high speed near coast)

Fleet Management*



https://cer.iit.demokritos.gr (refuelling opportunities)

^{*}Tsilionis et al, Online Event Recognition from Moving Vehicles. Theory and Practice of Logic Programming, 2019.

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union_all($[I_1, I_2, I_3], I_u$)



 $\begin{array}{l} \mathsf{intersect_all}([I_1, I_2, I_3], I_i) \\ \mathsf{union_all}([I_1, I_2, I_3], I_u) \end{array}$







^{*}Mantenoglou et al, Complex Event Recognition with Allen Relations. Knowledge Representation and Reasoning (KR), 2023.

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Semantics



Semantics



Semantics



Proposition

An event description in RTEC is a locally stratified logic program*.

^{*}Mantenoglou et al, Stream Reasoning with Cycles. Knowledge Representation and Reasoning (KR), 2022.











Windowing



Windowing



Windowing



Windowing: Delayed Additions and Deletions



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^{*}Tsilionis et al, Incremental Event Calculus for Run-Time Reasoning. Journal of Al Research (JAIR), 2022.



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RTEC: Correctness and Complexity

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RTEC computes all maximal intervals of a fluent, and no other interval, provided that interval delays/retractions, if any, are tolerated by the window size.

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Complexity

The time to compute the maximal intervals of a fluent is linear to the window size.

Performance: Indicative Results





Indicative Comparative Analysis



§Logica: Language of Big Data, https://github.com/EvgSkv/logica.

¹Falcionelli et al., Indexing the event calculus: Towards practical human-readable personal health systems. Artificial Intelligence in Medicine, 2019.

^{*}Srinivasan et al., Learning explanations for biological feedback with delays using an event calculus. Machine Learning, 2022.

[†]Arias et al., Modeling and reasoning in event calculus using goal-directed constraint answer set programming. Theory and Practice of Logic Programming, 2022.

[‡]Beck et al., Ticker: A system for incremental asp-based stream reasoning. Theory and Practice of Logic Programming, 2017.

Indicative Comparative Analysis

Monitoring maritime activities with Allen relations

Window size		Reasoning Time (ms)		Output Intervals	
Days	Input Intervals	RTEC	D ² IA*	RTEC	D ² IA
1	19K	40	410	6K	6K
2	37K	65	592	9K	9K
4	74K	99	1.1K	16K	16K
8	148K	156	1.6K	32K	31K
16	297K	285	2.7K	77K	76K

^{*}Awad et al, D²IA: User-defined interval analytics on distributed streams. Information Systems, 2022.

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Next: Handle the lack of veracity of data streams.