

Noether Embedding: Efficient Learning of Temporal Regularities

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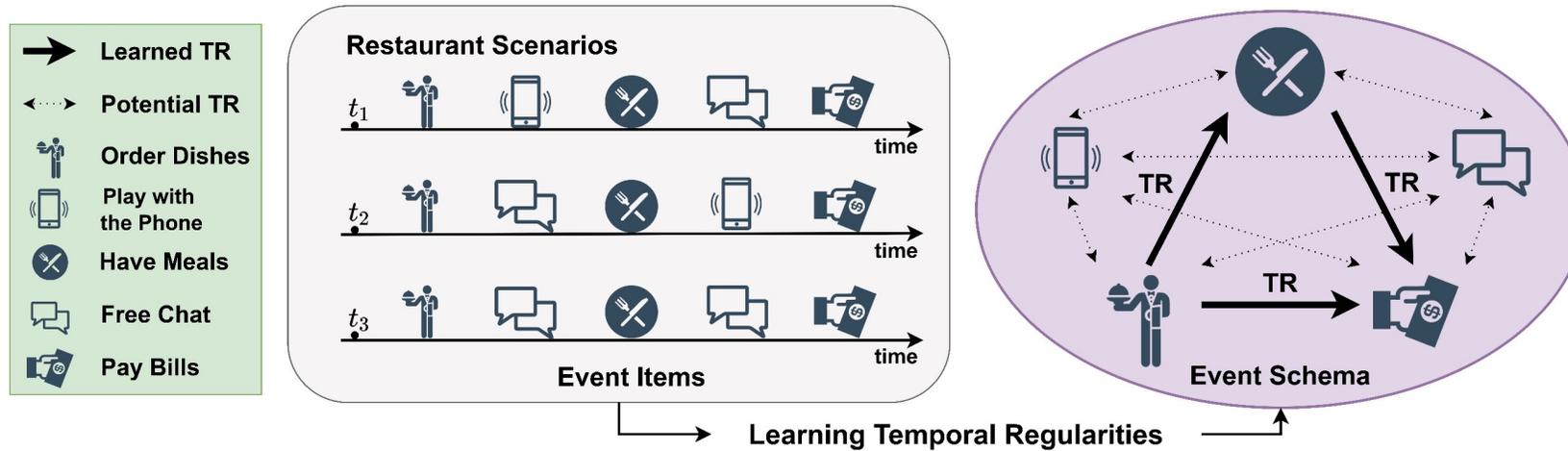
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Outline

- **Background & Motivation**
- **Problem Formulation**
- **Inspiration from Noether's Theorem**
- **Noether Embedding**
- **NE's Effectiveness, Efficiency, and Potential Applications**
- **Comparison to Related Work**

Background

Event schema is a relational memory structure in the brain that supports generalization.^[1]



Temporal regularity (TR) is a unit structure of event schemas,^[2]
which we formally define as **temporal associations that remain invariant to time shifts.**

$$(ev_b, t) \rightarrow (ev_h, t + \tau) \quad \forall t \in \mathbb{T}_a$$

[1] Ghosh, V. E. and Gilboa, A. What is a memory schema? a historical perspective on current neuroscience literature. *Neuropsychologia*, 53:104–114, 2014.

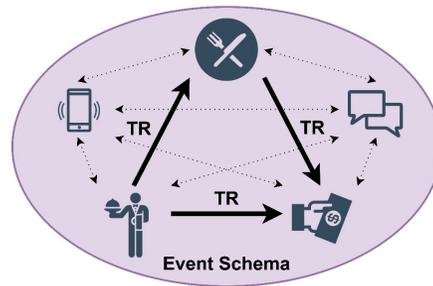
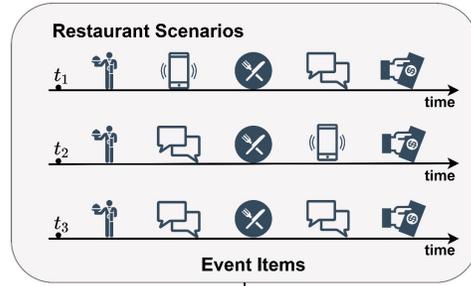
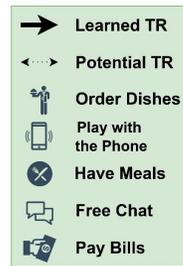
[2] McClelland, J. L., McNaughton, B. L., and O'Reilly, R. C. Why there are complementary learning systems in the hippocampus and neocortex: insights from the successes and failures of connectionist models of learning and memory. *Psychological review*, 102(3):419, 1995.

Motivation

Humans

**Data-efficient
TR formation** ^[1]

**Time-efficient
TR retrieval** ^[2]



Existing Event Embeddings

**Event
completion** ^[3]

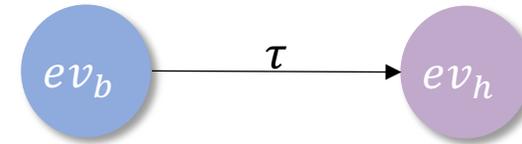
**Event
prediction** ^[4]

[1] Pudhiyidath, A., Roome, H. E., Coughlin, C., Nguyen, K. V., and Preston, A. R. Developmental differences in temporal schema acquisition impact reasoning decisions. *Cognitive neuropsychology*, 37(1-2):25–45, 2020.
 [2] Chaudhuri, R. and Fiete, I. Computational principles of memory. *Nature neuroscience*, 19(3): 394–403, 2016.
 [3] Cai, B., Xiang, Y., Gao, L., Zhang, H., Li, Y., and Li, J. Temporal knowledge graph completion: A survey. *arXiv preprint arXiv:2201.08236*, 2022.
 [4] Zhao, L. Event prediction in the big data era: A systematic survey. *ACM Computing Surveys (CSUR)*, 54(5):1–37, 2021.

Problem Formulation

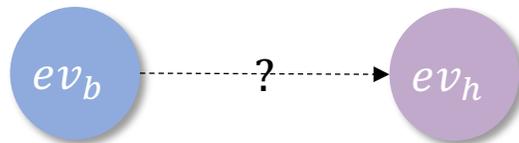
Aim: Learn TRs from event items $\{(ev, t)\}$

$$(ev_b, t) \rightarrow (ev_h, t + \tau) \quad \forall t \in \mathbb{T}_a$$

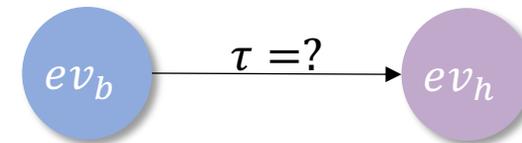


Tasks: TR Detection & TR Query

- **Detection:** Does (ev_h, ev_b) include a TR?



- **Query:** What is the τ for a TR (ev_h, ev_b) ?



Inspiration from Noether's Theorem

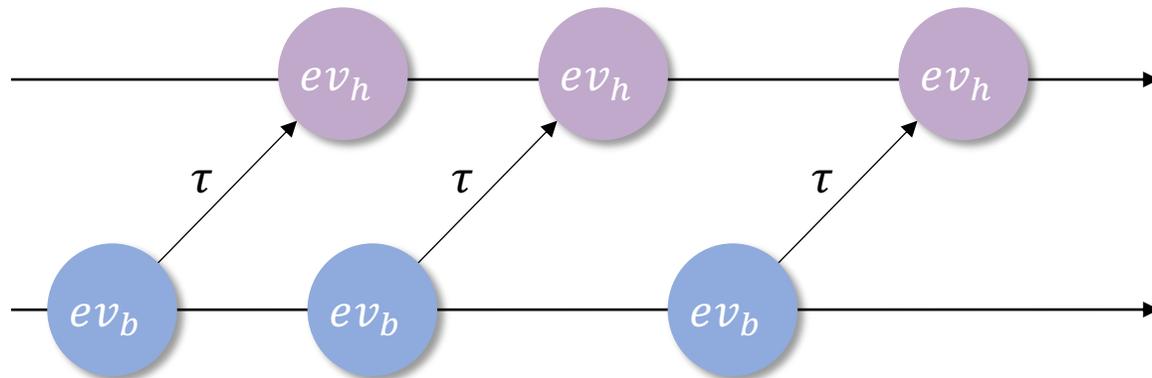
Noether's Theorem

Every differentiable symmetry of the action of a physical system with conservative forces has a corresponding conservation law.

Specifically, time-translation symmetry corresponds to energy conservation.

Inspiration

TRs inherently possess
time-translation symmetries



Event embedding $\mathbf{q}(t; ev)$

Local energy $g(\mathbf{q}(t, ev_b), \mathbf{q}(t + \tau, ev_h))$



TR validity



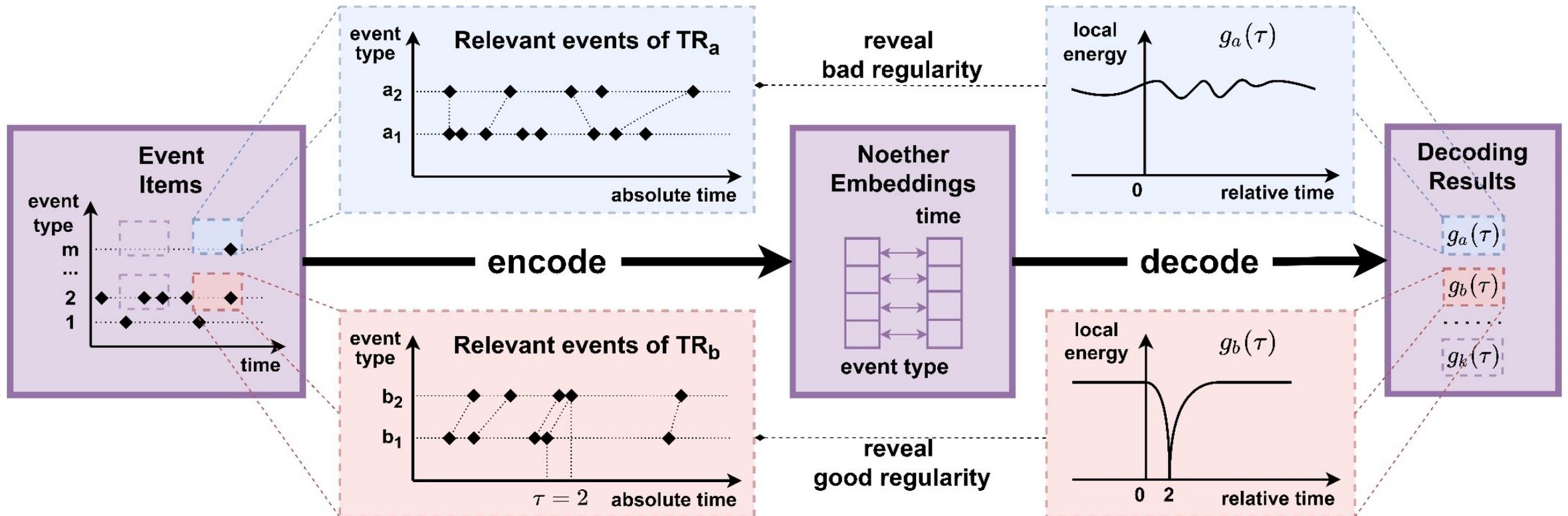
Conserved

Noether Embedding (NE)

Event embedding $q(t; ev) = u(ev) \circ r(t)$

Score function $f(t; ev) = \sum_{i=1}^d \text{Real}(q(t; ev))_i$

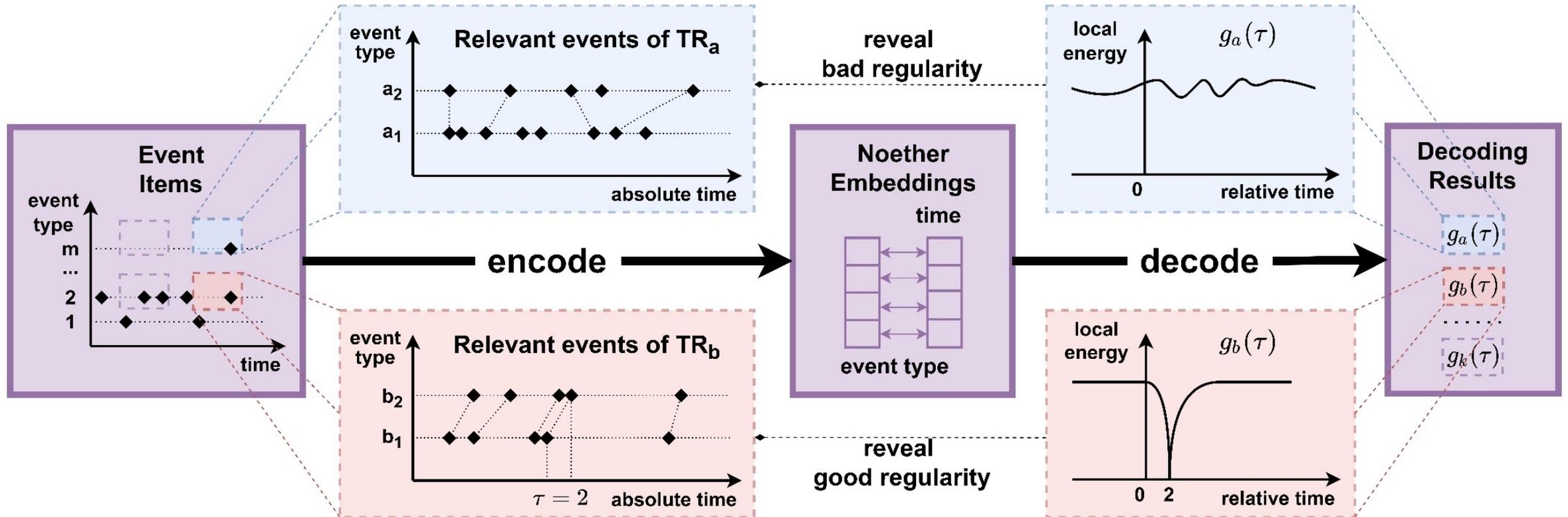
Loss function $L(\xi; C_p, C_n) = \left(\frac{1}{\sqrt{d}}f(\xi) - C_p\right)^2 + \frac{1}{N} \sum \left(\frac{1}{\sqrt{d}}f(\xi') - C_n\right)^2$



Noether Embedding (NE)

Decoding function

$$g(\tau) = |\mathbf{u}_b - \mathbf{u}_h \circ \mathbf{r}(\tau)|^2 \quad (= |\mathbf{u}_b \circ \mathbf{r}(t) - \mathbf{u}_h \circ \mathbf{r}(t + \tau)|^2 = |\mathbf{q}_b(t) - \mathbf{q}_h(t + \tau)|^2)$$



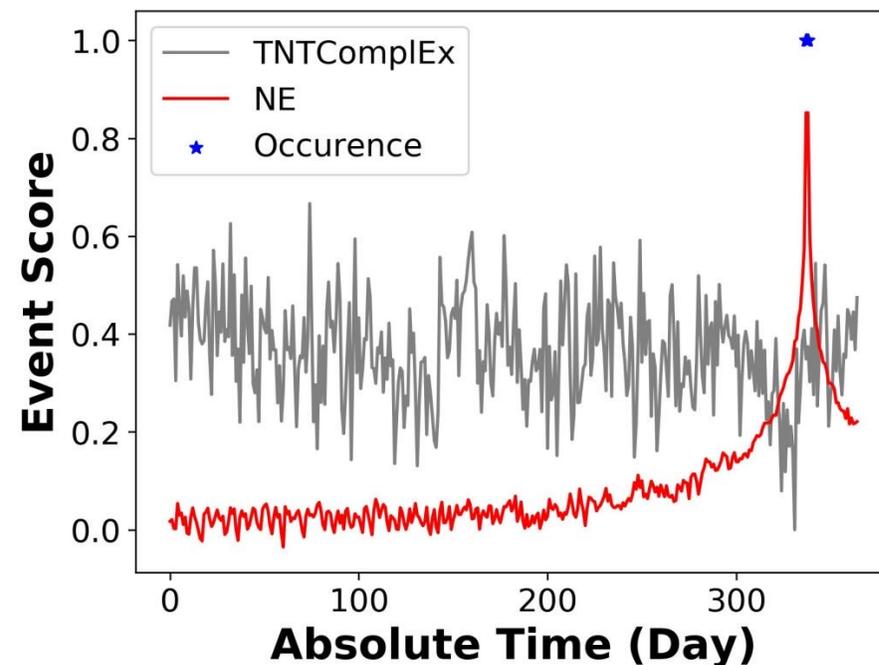
Effectiveness of NE

Table 1: Statistical results on ICEWS14, ICEWS18, and GDELT

Embedding	TR Detection (F1)			TR Query (r)		
	D14	D18	GDELT	D14	D18	GDELT
TNTComplex	0.26	0.18	0.08	0.08	0.08	0.01
DE-Simple	0.22	0.20	-	0.09	0.09	-
TASTER	0.18	0.15	0.08	0.09	0.09	0.00
TeRo	0.43	0.64	0.16	0.08	0.08	0.01
BoxTE	0.40	0.40	0.18	0.08	0.08	0.01
ATISE	0.40	0.44	0.18	0.08	0.08	0.01
NE with $g'(\tau)$	0.78	0.79	0.48	0.85	0.83	0.83
NE with $g(\tau)$	0.82	0.83	0.51	0.87	0.86	0.85

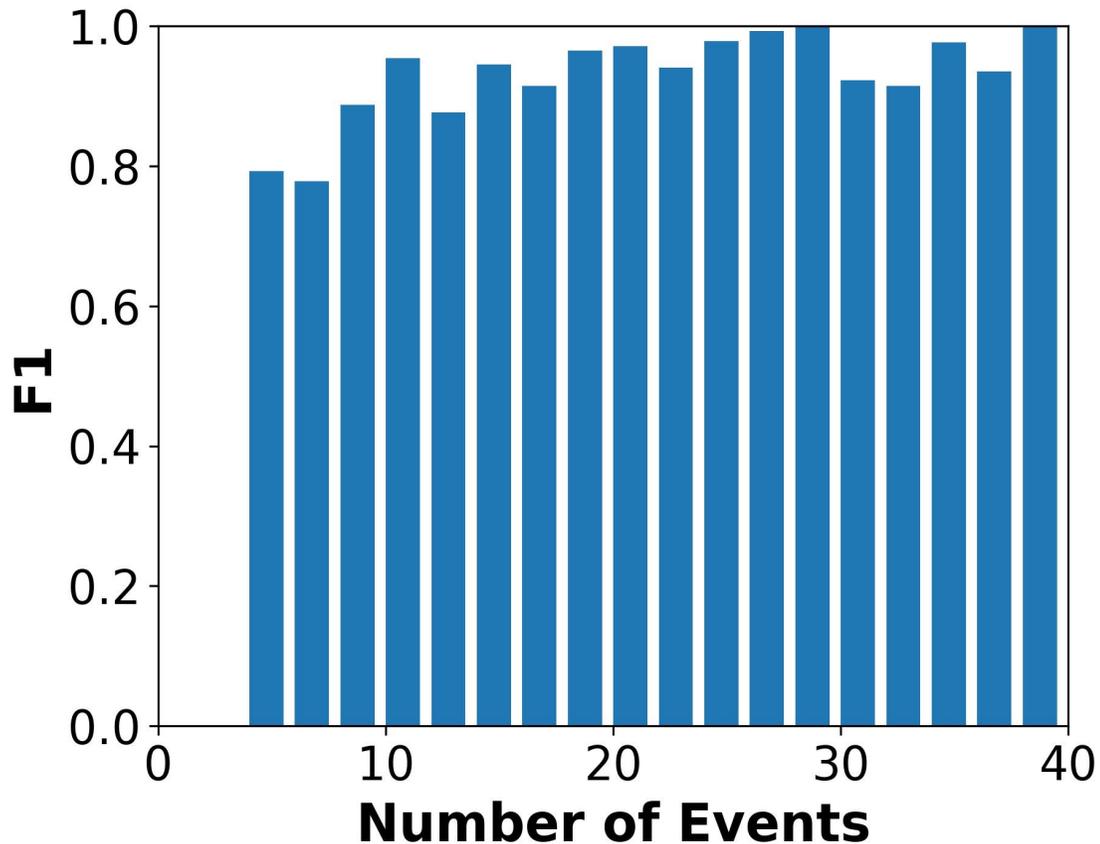
$$g'(\tau) = \frac{\sum_{t, t+\tau \in \mathbb{T}_a} f'_b(t) f'_h(t+\tau)}{\sum_{t \in \mathbb{T}_a} f'_b(t) \cdot \sum_{t \in \mathbb{T}_a} f'_h(t)}$$

Baselines **over-apply** the generalization capabilities of distributed representations, which **hinders the fit of event occurrences**.



Efficiency of NE

Data-efficient TR formation



Time-efficient TR retrieval

$$g(\tau) = \|\mathbf{u}_b - \mathbf{u}_h \circ \mathbf{r}(\tau)\|^2$$

$$g'(\tau) = \frac{\sum_{t, t+\tau \in \mathbb{T}_a} f'_b(t) f'_h(t+\tau)}{\sum_{t \in \mathbb{T}_a} f'_b(t) \cdot \sum_{t \in \mathbb{T}_a} f'_h(t)}$$

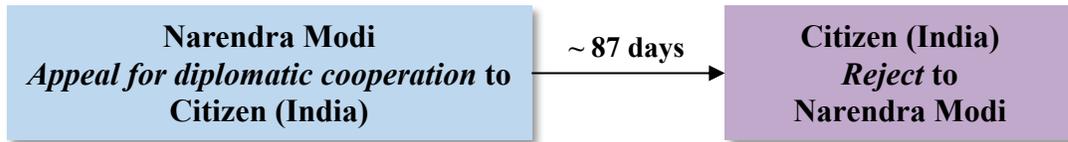
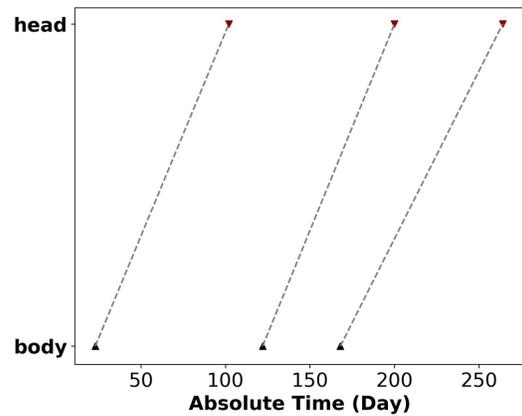
Storage-efficient TR memory

$$\begin{aligned} S(\mathbf{NE}) &= S(\text{ev} - \text{vector}) + S(\text{time} - \text{vector}) \\ &= (2Nd + 2d) \times 64 \text{bit} \end{aligned}$$

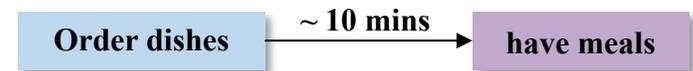
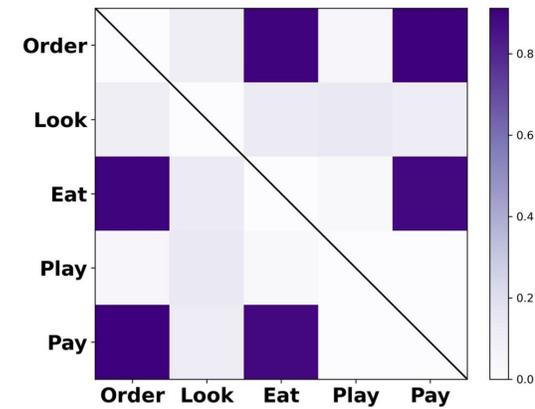
$$\begin{aligned} S(\mathbf{CT}) &= S(\mathbf{TR}) + S(\text{event}) \\ &= N^2 \times T_a \times \log_2 \frac{n}{N} \text{bit} + N \times \frac{n}{N} \times \log_2 T_a \text{bit} \end{aligned}$$

NE's Potential Applications

Social event prediction



Personal decision making



Memory-constrained scenarios, such as the edge

Comparison to Related Work

- **Event schema induction (in natural language processing):**

organize known event regularities **already given as priors** for the extracting algorithm and focus on the schemas for use



Our tasks are designed to learn event regularities **directly from experience without supervision.**

- **Temporal rule mining (in data mining)**

the mined rules and source events are generally stored as symbolic representations in list form



Using event embeddings, NE strikes a balance between **storage efficiency and storage accuracy.**

- **Embedding models of structured data (in representation learning)**

embed (temporal) knowledge graphs, generally supporting completion, inference, and prediction tasks



NE for the first time enables event embeddings to achieve **data-efficient TR formation, storage-efficient TR memory, and time-efficient TR retrieval.**

Thank you