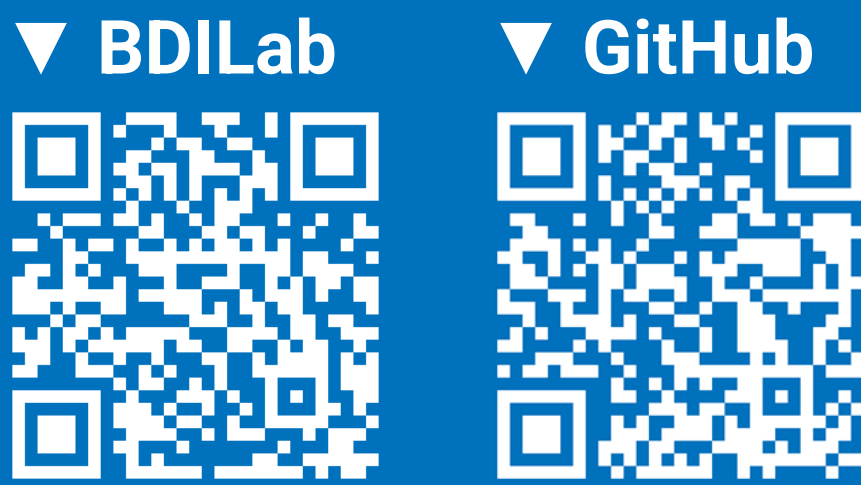


# Representation Learning on Hyper-Relational and Numeric Knowledge Graphs with Transformers



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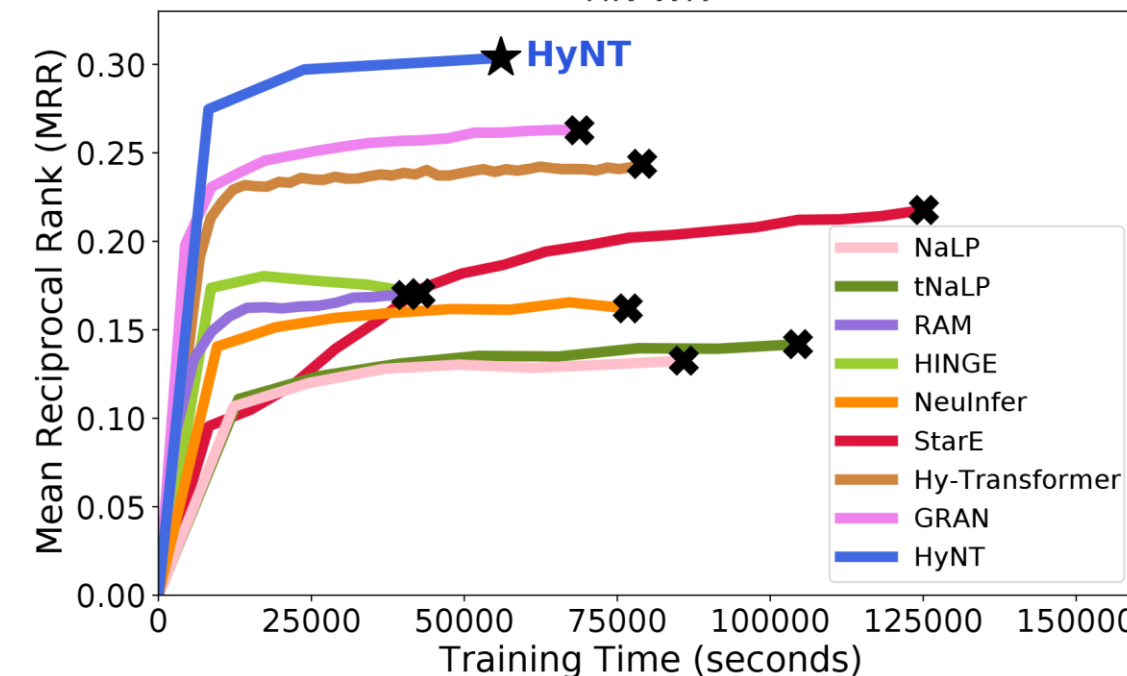
School of Computing, KAIST

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KAIST

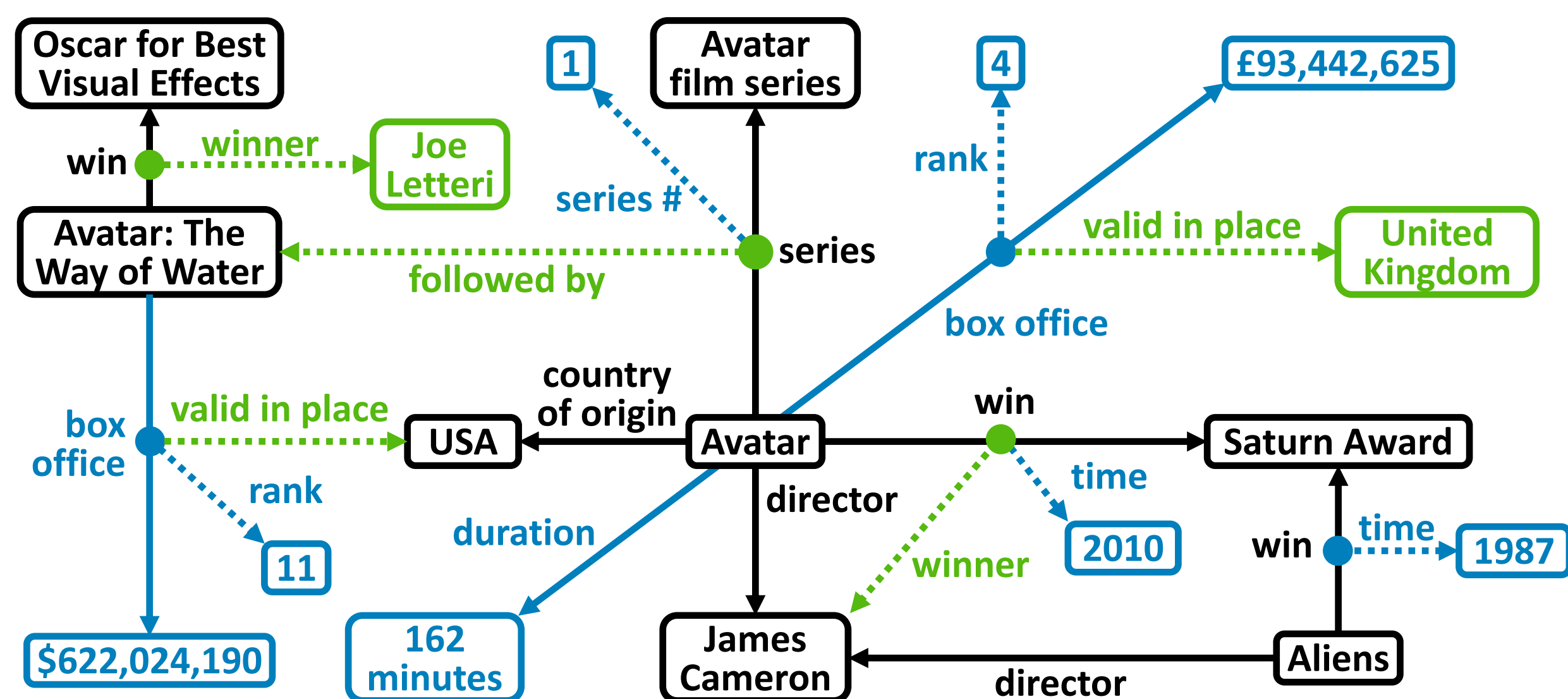
## Main Contributions

- Define **Hyper-relational and Numeric Knowledge Graphs (HN-KGs)**
  - Create 4 real-world HN-KG datasets
- Propose **HyNT**, Hyper-relational knowledge graph embedding with **N**umeric literals using **T**ransformers
  - Define a context transformer and a prediction transformer
  - Reduce the cost by learning compact representations of triplets and qualifiers
- HyNT significantly outperforms 12 different state-of-the-art methods for **link prediction**, **numeric value prediction**, and **relation prediction**



## Hyper-relational and Numeric Knowledge Graphs (HN-KGs)

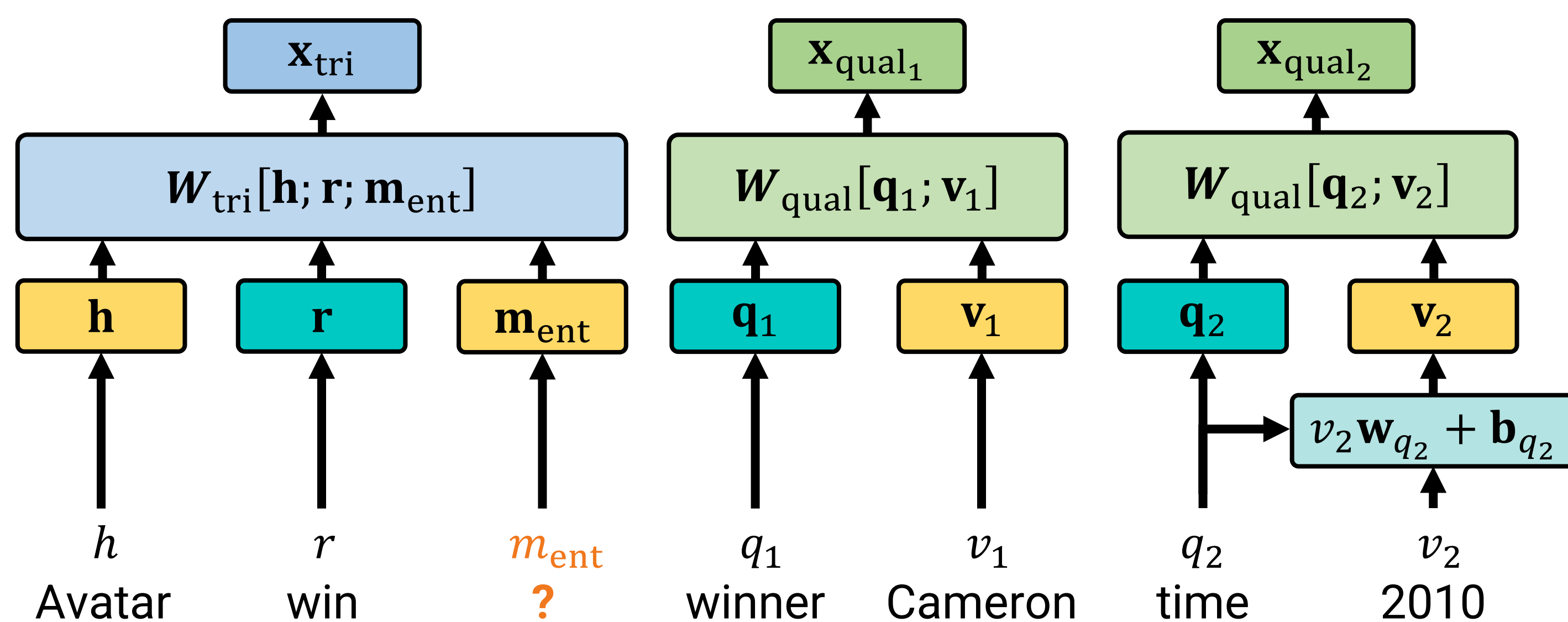
- Hyper-relational Knowledge Graphs**
  - Attach a set of **qualifiers** to a triplet to enrich information
  - Existing methods assume that all entities are **discrete** objects
- Hyper-relational and Numeric Knowledge Graphs**
  - Contain both hyper-relational facts and numeric values



- An example of a hyper-relational fact
  - $((\text{Avatar}, \text{win}, \text{Saturn\_Award}), \{(\text{winner}, \text{James\_Cameron}), (\text{time}, 2010)\})$ 
    - Primary Triplet
    - Qualifier 1
    - Qualifier 2
- Predictions on HN-KGs
  - Link Prediction:** Predict a missing **discrete entity**
  - Numeric Value Prediction:** Predict a missing **numeric value**
  - Relation Prediction:** Predict a missing **relation**
  - The missing component can be in either a **primary triplet** or a **qualifier**

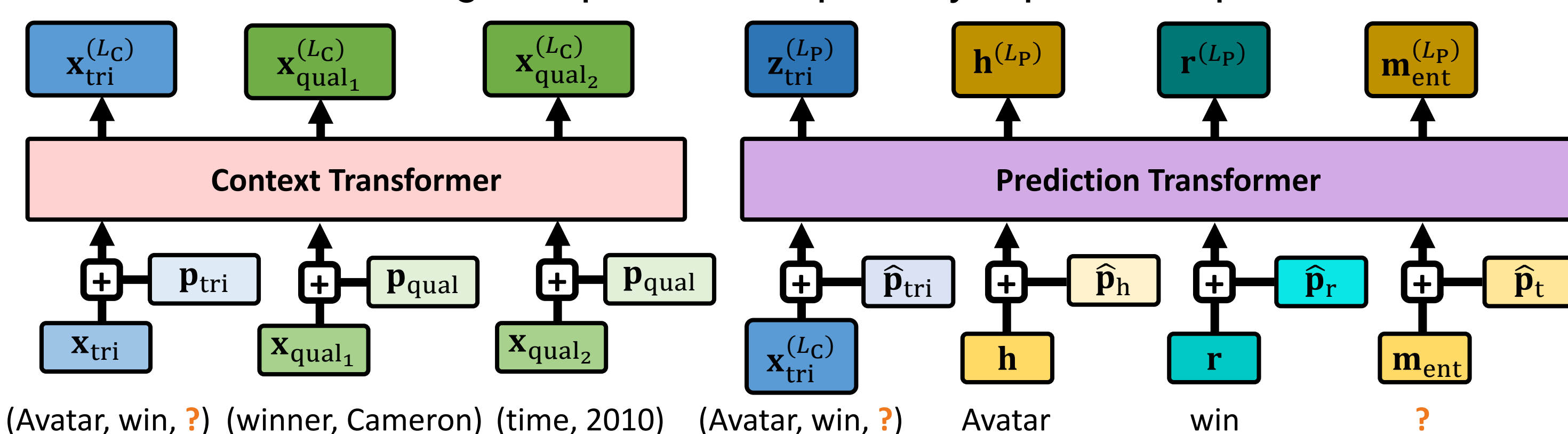
## Triplet/Qualifier Encoding

- Convert a triplet/qualifier to a representation vector



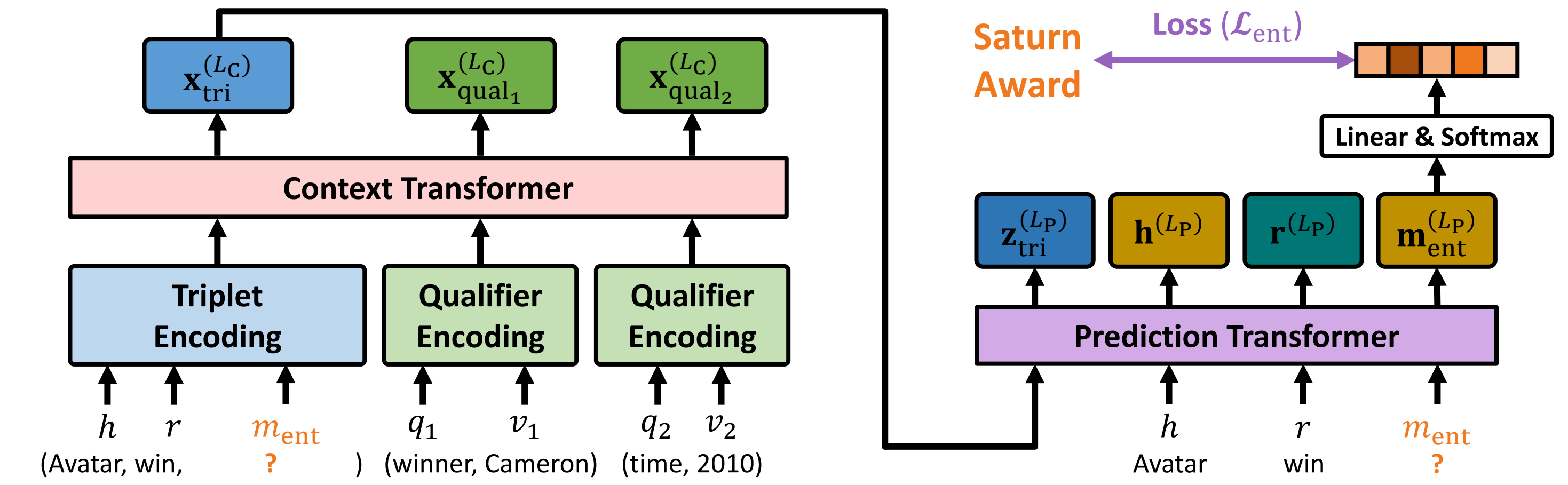
## Context & Prediction Transformers

- Context Transformer**
  - Learn the representations of a primary triplet and the qualifiers
- Prediction Transformer**
  - Predict a missing component in a primary triplet or a qualifier

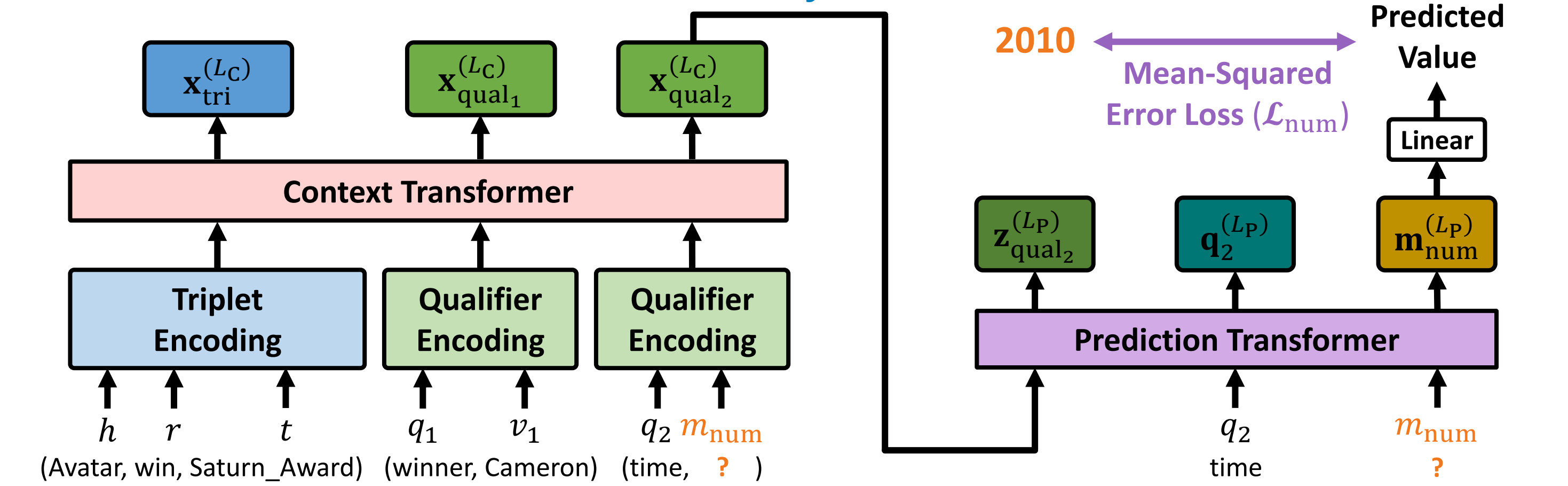


## Training & Predictions of HyNT

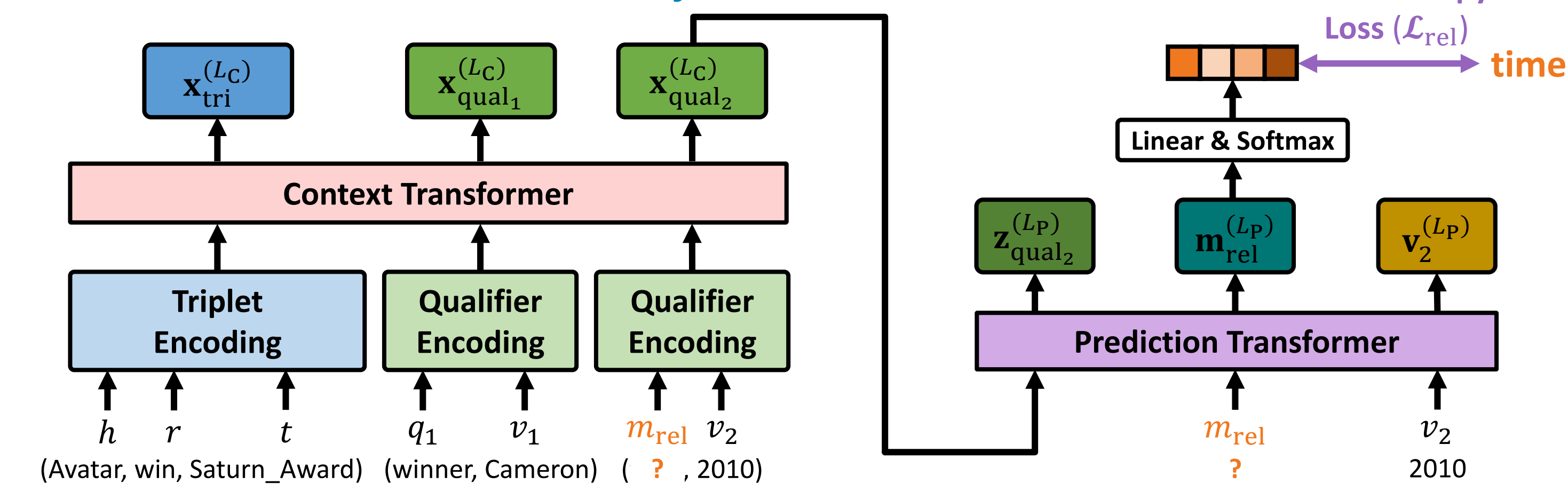
- Link Prediction Loss of HyNT**



- Numeric Value Prediction Loss of HyNT**



- Relation Prediction Loss of HyNT**



- Loss of HyNT:**  $\mathcal{L} := \mathcal{L}_{\text{ent}} + \lambda_1 \cdot \mathcal{L}_{\text{rel}} + \lambda_2 \cdot \mathcal{L}_{\text{num}}$

## Experimental Results

- Baseline methods:** TransEA, MT-KGNN, KBLN, LiteralE, NaLP, tNaLP, RAM, HINGE, NeuInfer, StarE, Hy-Transformer, GRAN

- Link Prediction Results (MRR,  $\uparrow$ )**

		HN-WK	HN-YG	HN-FB	HN-FB-S
Primary	Best-baseline	0.2627	0.1951	0.2602	0.5077
	HyNT	<b>0.3037</b>	<b>0.2035</b>	<b>0.4544</b>	<b>0.5079</b>
All	Best-baseline	0.2901	0.1951	-	<b>0.5873</b>
	HyNT	<b>0.3254</b>	<b>0.2035</b>	<b>0.5022</b>	0.5796

- Numeric Value Prediction Results (RMSE,  $\downarrow$ )**

		HN-WK	HN-YG	HN-FB	HN-FB-S
Primary	Best-baseline	0.0761	0.0778	0.0637	0.0656
	HyNT	<b>0.0548</b>	<b>0.0706</b>	<b>0.0517</b>	<b>0.0532</b>
All	Best-baseline	0.0820	0.1123	-	0.0627
	HyNT	<b>0.0405</b>	<b>0.0694</b>	<b>0.0558</b>	<b>0.0499</b>

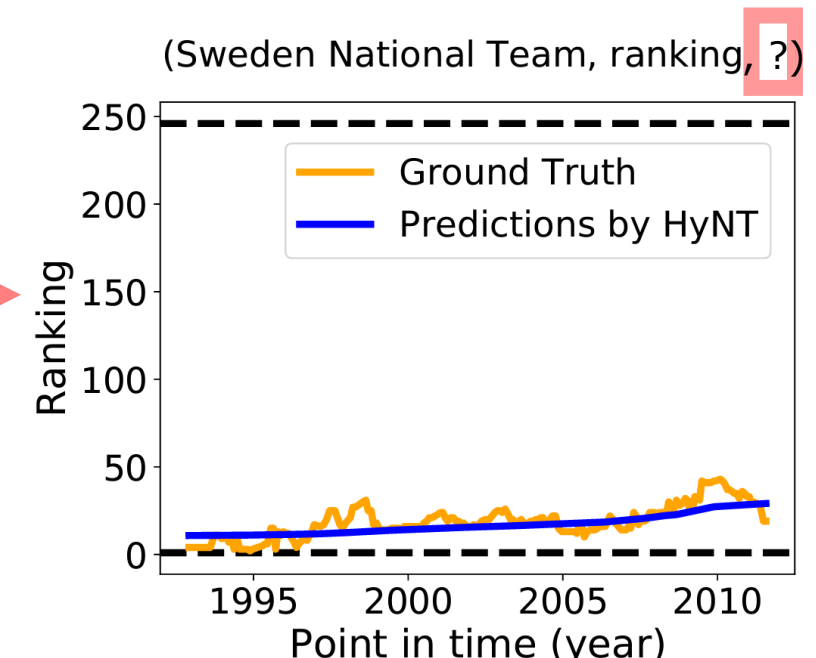
- Relation Prediction Results (MRR,  $\uparrow$ )**

		HN-WK	HN-YG	HN-FB	HN-FB-S
Primary	Best-baseline	0.9285	0.8347	-	<b>0.9845</b>
	HyNT	<b>0.9474</b>	<b>0.8797</b>	<b>0.9809</b>	0.9815
All	Best-baseline	0.9599	0.8548	-	<b>0.9918</b>
	HyNT	<b>0.9706</b>	<b>0.8944</b>	<b>0.9860</b>	0.9902

- Visualization of Numeric Value Predictions**

- Numeric value prediction problems in a particular form

$((\text{Sweden National Team}, \text{ranking}, ?), \{(\text{point in time}, 1992)\})$   
 $((\text{Sweden National Team}, \text{ranking}, ?), \{(\text{point in time}, 1993)\})$   
 $((\text{Sweden National Team}, \text{ranking}, ?), \{(\text{point in time}, 1994)\})$



## Conclusion & Future Work

- Introduce the concept and real-world datasets for **Hyper-relational and Numeric Knowledge Graphs (HN-KGs)**
- Propose **HyNT** to solve **link prediction**, **numeric value prediction**, and **relation prediction** on HN-KGs
- HyNT significantly outperforms 12 different state-of-the-art methods
- Plan to extend HyNT to **inductive learning scenarios** where new entities and relations appear at test time