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# Movie Box Office Prediction With Self-Supervised and Visually Grounded Pretraining

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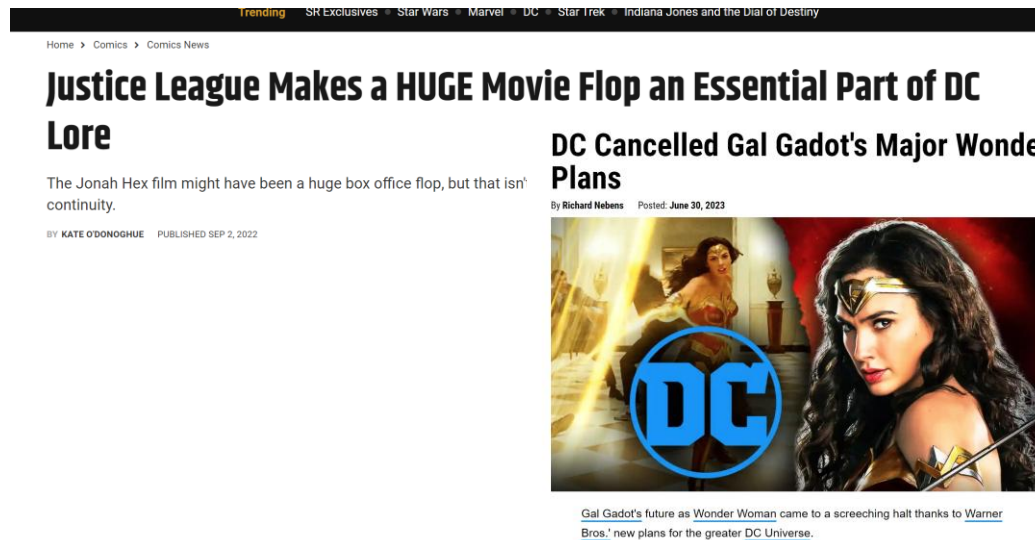
*13 July 2023*





# Introduction to Movie Box Office Prediction

- Movie investment carries significant risks
  - "Cleopatra (1963)", nearly ruined 20th Century Fox
  - "The Golden Compass (2007)", caused New Line Cinema absorbed into Warner Bros.
  - "Cutthroat Island(1995) “ made Carolco Pictures ceased to exist.



## The Flash's Box Office Drop Is The 2nd Worst Behind Another Big Superhero Flop

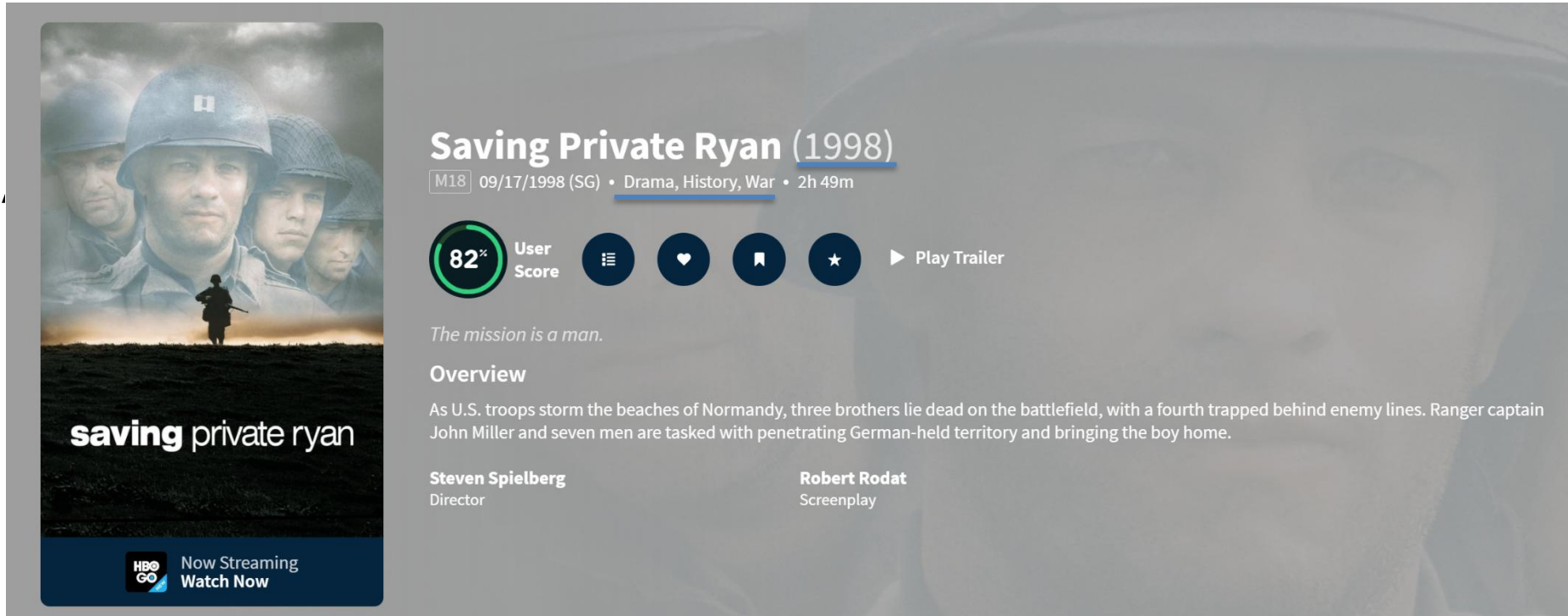
The Flash movie continues to underperform and disappoint at the box office in its second weekend, echoing a previous superhero bomb.

BY NICHOLAS BRADY PUBLISHED 3 DAYS AGO



# Introduction to Movie Box Office Prediction

- Features Collection: TMDb website



**Saving Private Ryan (1998)**  
M18 09/17/1998 (SG) • Drama, History, War • 2h 49m

**82%** User Score

[Play Trailer](#)

*The mission is a man.*

**Overview**

As U.S. troops storm the beaches of Normandy, three brothers lie dead on the battlefield, with a fourth trapped behind enemy lines. Ranger captain John Miller and seven men are tasked with penetrating German-held territory and bringing the boy home.

**Steven Spielberg**  
Director

**Robert Rodat**  
Screenplay

**HBO GO** Now Streaming Watch Now

# Challenges

How to learn an effective representation for movie box prediction

## Data Sparsity

- Near-synonym keywords
- Missing keywords

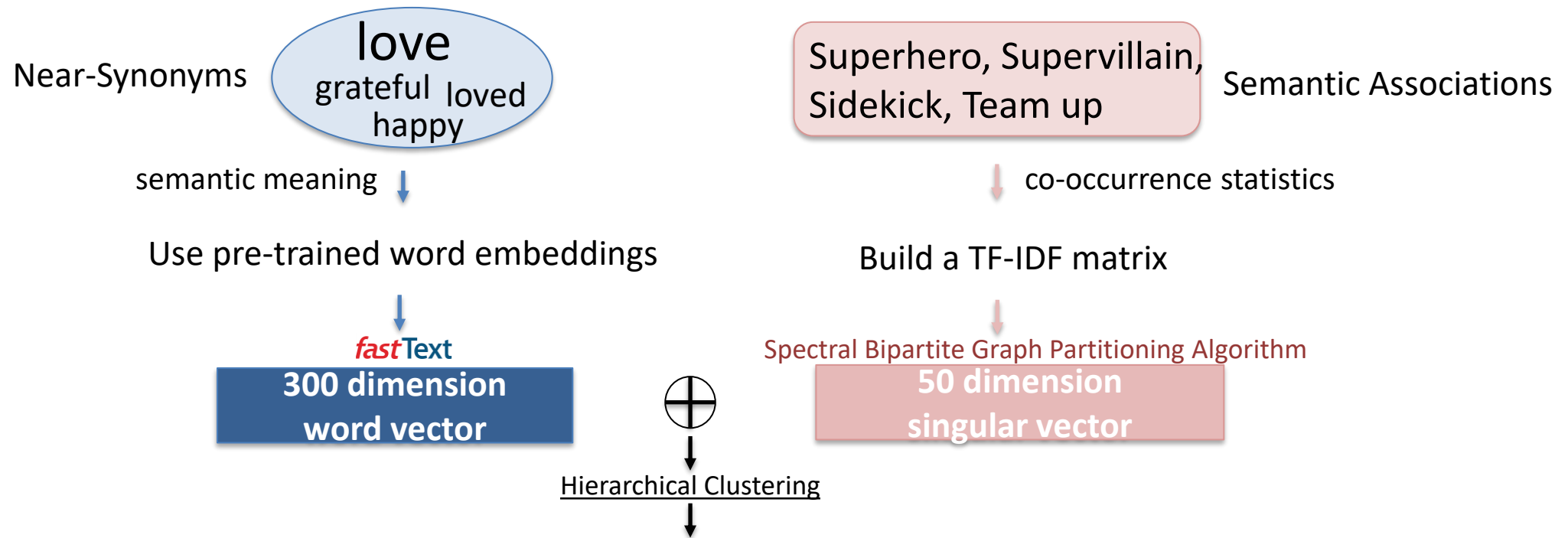
## Idiosyncrasy

- Robots in Sci-fi vs Robots on assembly line
- BERT embedding does not fit well

## Multi-Modal

- Movie poster under-utilized

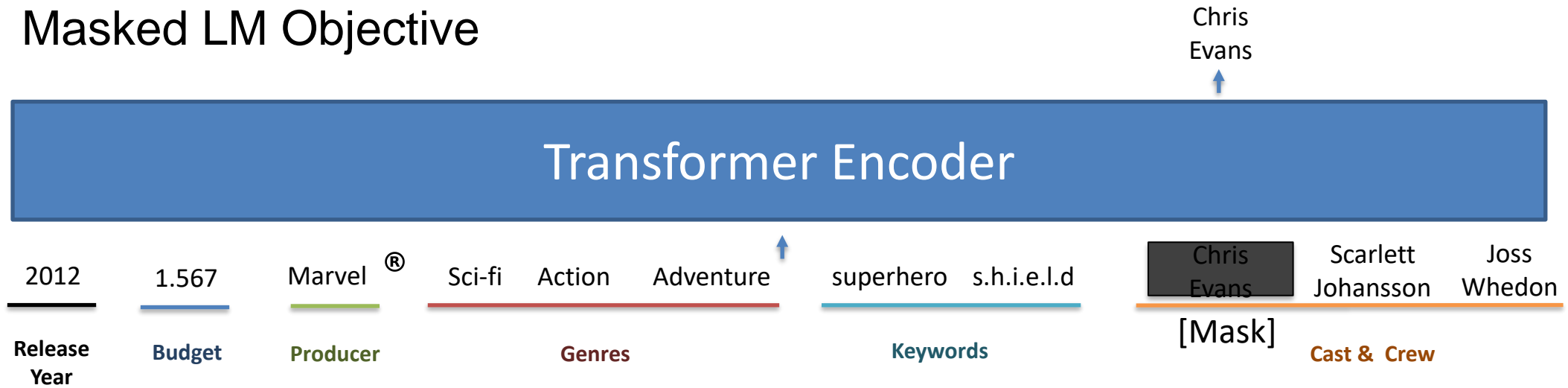
# Keywords Clustering



| Cluster Label     | Elements   |
|-------------------|--|
| drama             | 'love', 'loved', 'hate', 'unhappy', 'waiting', 'happy', 'grateful', 'lucky', 'expecting', 'loving'                             |
| superhero-related | 'superhero', 'villainess', 'villain', 'symbiote', 'sidekick', 'superhuman', 'teamup', 'nemesis', 'superheroes', 'supervillain' |
| psycho-related    | 'psycho', 'psychotic', 'pyromaniac', 'psychopathic', 'homicidal', 'deranged'   |

# Self-Supervised Learning Pretraining

- Masked LM Objective



- Numerical Embedding (compute the distance to an anchor vector)

$$\text{NE}_i(x) = \exp\left(-\frac{\|x - q_i\|_2}{\sigma^2}\right), \text{ where } \{q_i\}_{i=0}^{D-1} \text{ are } D \text{ evenly spaced number over } [-10, 10]$$



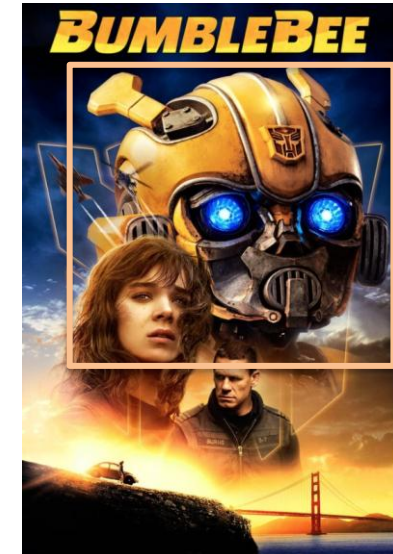
# Visual Grounding

- 'robot' have different meaning in film



Sparrow, Amazon's latest warehouse robot, leverages computer vision and artificial intelligence to recognise and handle millions of items

VS



- Use off-the-shelf object detection model (e.g., VinVL, 1500+ object labels )
  - Extract local feature maps for all the detections (discard tiny objects and titles)

# Visual grounding

- Contrastive Learning Objective

$\mathcal{Z}_i = \{z_m\}_{m=1}^M$  : visual features for  $M$  objects on the poster  $\longleftrightarrow$  similar  $\mathcal{X}_i = \{x_k\}_{k=1}^K$  : contextualized embeddings of the  $K$  keywords

- Many-to-Many

Positive pairs:  $(\mathbf{x}, \mathbf{z}) \in \mathcal{X}_i \times \mathcal{Z}_i$

$$\mathcal{L}_{\text{VG}} = -\frac{1}{N} \sum_{i=1}^N \log \left( \frac{\text{sim}(i, i)}{\text{sim}(i, i) + \sum_{(i', j')} \text{sim}(i', j')} \right)$$

↓

$$\text{sim}(i, i) = \sum_{(\mathbf{x}, \mathbf{z}) \in \mathcal{X}_i \times \mathcal{Z}_i} \exp\left(\frac{\mathbf{x}^\top \mathbf{z}}{\|\mathbf{x}\|_2 \|\mathbf{z}\|_2}\right)$$



# Challenges

How to learn an effective representation for movie box prediction

Data Sparsity

Idiosyncrasy

Multi-Modal

Keywords  
Clustering

Self-Supervised  
Pretraining

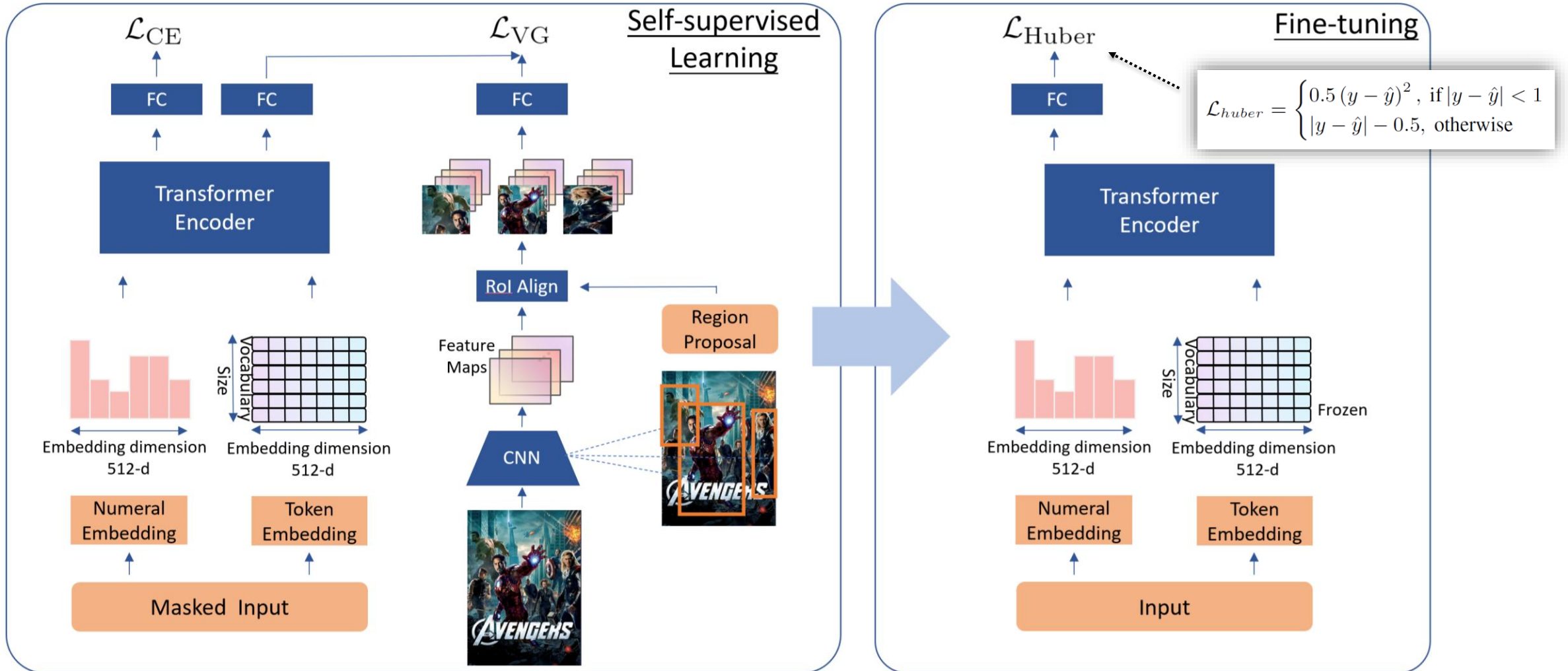
Visual  
Grounding



# Model Architecture

An example of input with textual and numerical features:

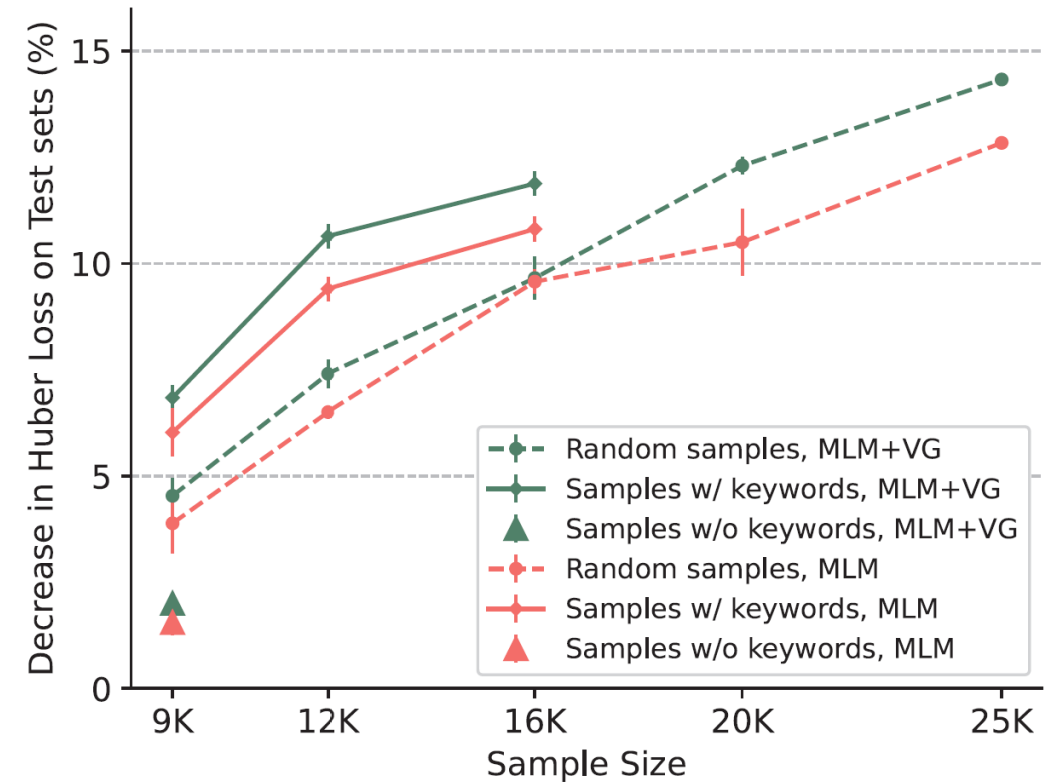
[CLS][PG-13]1.5678[Genres][Action][Sci-Fi][Keywords][shield][superhero][Directors][Joss Whedon][Actors][Chris Evans][SEP]



# Main Results

1. Our best model shows a 14.5% of accuracy improvement compared to BERTsmall.
2. Independent to the sample size, VG method consistently improve the result.

| Model                                 | Test Huber Loss(% improvement) |                 |
|---------------------------------------|--------------------------------|-----------------|
| <b>Numerical features only</b>        |                                |                 |
| Random Forest                         | 0.3677 (−3.5%)                 |                 |
| <b>Textual and numerical features</b> |                                |                 |
| BERT <sub>small</sub> finetuned       | 0.3553 (baseline)              |                 |
| BERT <sub>medium</sub> finetuned      | 0.3446 (2.5%)                  |                 |
| <b>Our models</b>                     | <b>Clustering</b>              | <b>Keywords</b> |
| Random init.                          | 0.3290 (7.4%)                  | 0.3265 (8.1%)   |
| + MLM pretraining                     | 0.3109 (12.5%)                 | 0.3133 (11.8%)  |
| + VG pretraining                      | 0.3070 (13.6%)                 | 0.3109 (12.5%)  |
| BERT embeddings init.                 | 0.3137 (11.7%)                 | 0.3249 (8.6%)   |
| + MLM pretraining                     | 0.3102 (12.7%)                 | 0.3226 (9.2%)   |
| + VG pretraining                      | 0.3037 (14.5%)                 | 0.3182 (10.4%)  |



# Qualitative Eval – Image Retrieval



Fig 4a: Use the contextualized word embedding of the keyword ‘love’ in the context of a romantic movie *One Day* (2009) to retrieval movie posters. The ground truth shows up as the top 6<sup>th</sup> .



# Conclusion

- We propose to pretrain a transformer network with masked language modeling and visual grounding objectives tailored to the film industry context.
- Compared to BERT embedding, the contextualized and visual grounded representation improve the box office prediction accuracy.
- We constructed a large dataset for community to continue exploring the movie box office prediction task.

Paper:



GitHub:





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