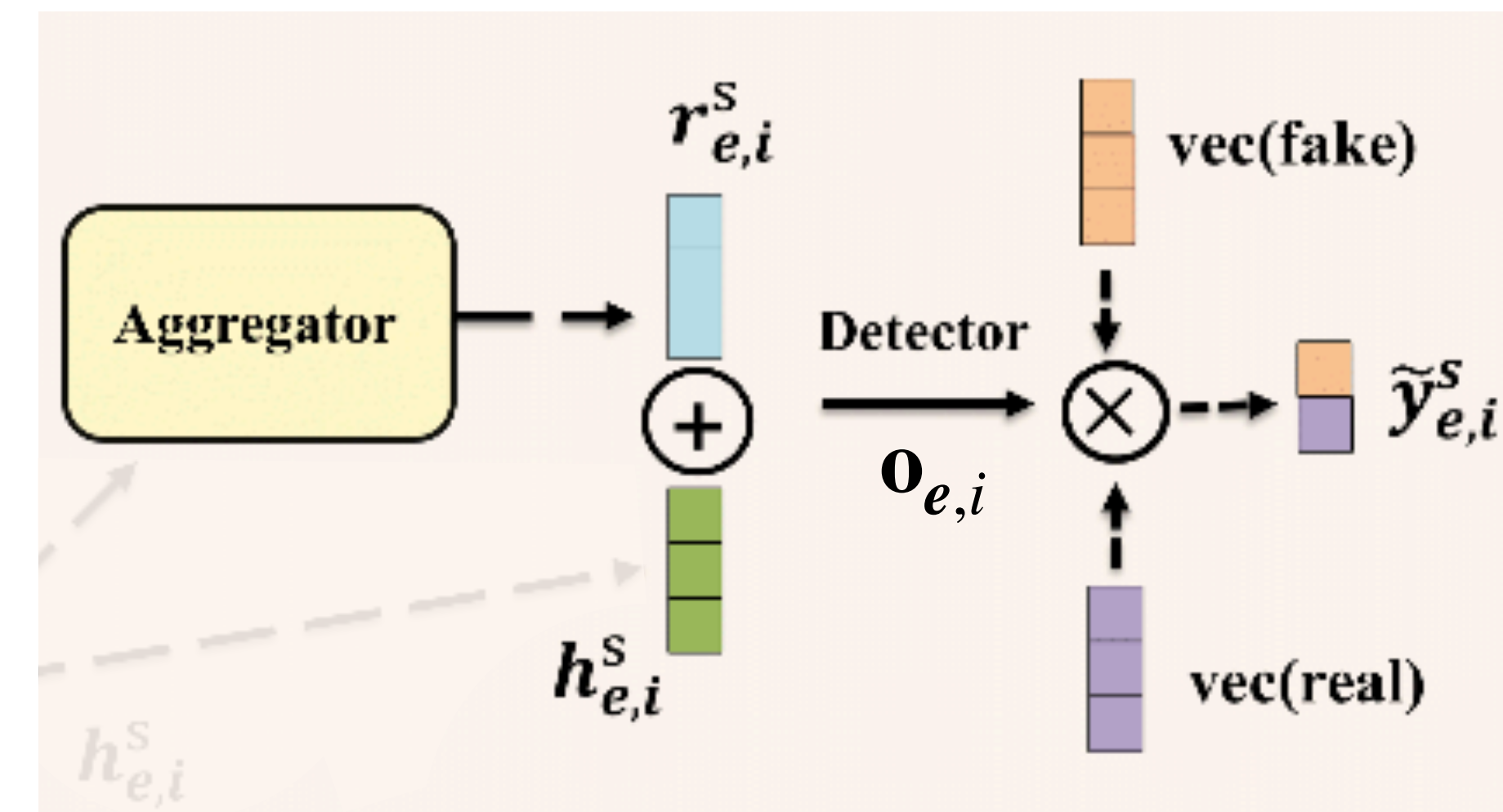


Methodology

Detector based on Label Embedding

- Define two embeddings $\text{vec}(\text{fake})$ and $\text{vec}(\text{real})$.
- To ensure that the label embedding can capture the semantic meanings of corresponding labels, propose to use embeddings $\text{vec}(\text{fake})$ and $\text{vec}(\text{real})$ in the detector as metrics and output prediction are determined based on metric matching.
- The detector is fully connected layer output vector $\mathbf{o}_{e,i}$.
- $\text{similarity}(\mathbf{o}_{e,i}, \text{vec}(\text{fake})) = \|\mathbf{o}_{e,i} \circ \text{vec}(\text{fake})\|$,
 $\text{similarity}(\mathbf{o}_{e,i}, \text{vec}(\text{real})) = \|\mathbf{o}_{e,i} \circ \text{vec}(\text{real})\|$
- The two scores then mapped into $[0,1]$ as probabilities via softmax.



Experiments

Datasets

	Twitter	Weibo
# of fake News	6,934	4,050
# of real News	5,683	3,558
# of images	514	7,606

- Twitter, Weibo datasets
- The news events are included in the Twitter dataset, obtain events on Weibo dataset via single-pass clustering method.
- Only keep the events which associated with **more than 20 posts** and randomly split the posts on same event into support and query data.
- Training and testing set **do not contain any common event**.