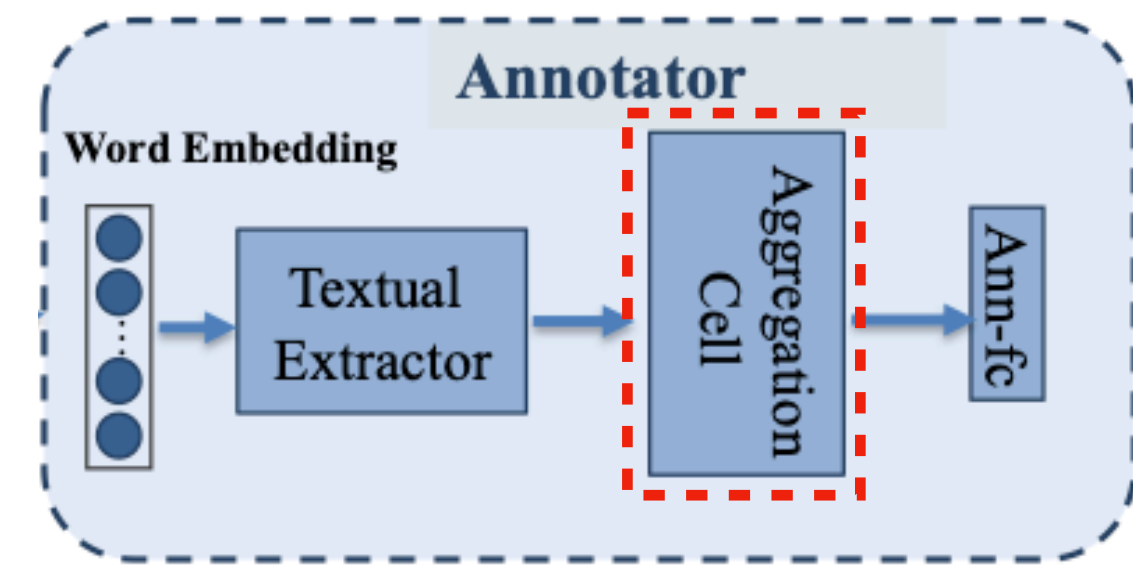


# Methodology

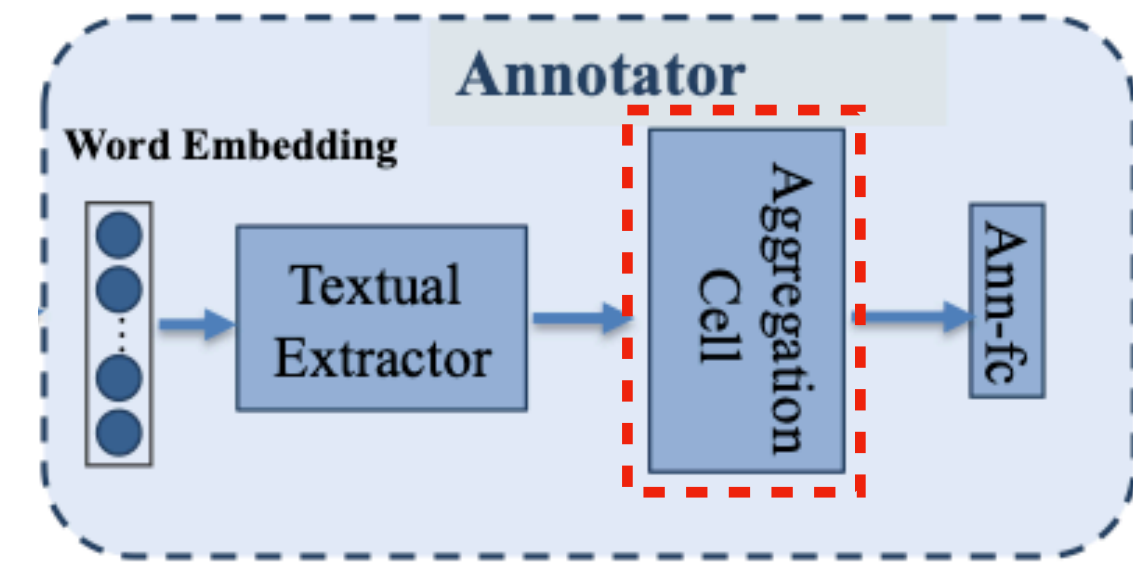
## Automatic Annotation based on Reports



- One news article may have reports from multiple users
- Propose to aggregate features obtained from different reports for one sample.
- Design an aggregation cell consisting of a commutative aggregation function and a fully-connected layer
- The commutative aggregation function, like sum, mean and max-pooling, can combine the permutation invariant input set

# Methodology

## Automatic Annotation based on Reports



- Take the  $i$ -th sample as an example, and the  $j$ -th report message:  $r_j^{(i)}$
- The corresponding report message set is denoted as  $R^{(i)} = \{r_1^{(i)}, r_2^{(i)}, \dots, r_{|R^{(i)}|}^{(i)}\}$ 
  - $|R^{(i)}|$ : number of report messages of  $i$ -th sample
- $r_j^{(i)} \in R^{(i)}$  is first fed into the textual feature extractor to obtain  $\mathbf{h}_j^{(i)}$
- Use the aggregation cell to combine  $R^{(i)}$  to learn the hidden feature representation  $\mathbf{h}^{(i)}$
- Procedure of aggregation cell:  $\mathbf{h}^{(i)} = \text{ReLU} \left( \mathbf{w}_r \cdot \sum_{j=1}^{|R^{(i)}|} \frac{\mathbf{h}_j^{(i)}}{|R^{(i)}|} \right)$ ,  $\mathbf{w}_r$ : weight of the fully-connected layer