

2. Multilingual word embeddings

$$\min \|WX - Y\|_F^2 = \min \text{Tr}[(WX - Y)^T (WX - Y)]$$
$$= \min \text{Tr}(X^T X)$$

$$= \min [\text{Tr}(X^T X) + \text{Tr}(Y^T Y) - 2\text{Tr}(X^T W^T Y)]$$

$$= \min [\text{Tr}(X^T X) + \text{Tr}(Y^T Y) - 2\text{Tr}(W^T Y W^T)]$$

$$= \min [\text{Tr}(X^T X) + \text{Tr}(Y^T Y) - 2\text{Tr}(\Sigma U^T W V)]$$

so we need to minimize $-2\text{Tr}(\Sigma U^T W V)$

For a positive diagonal matrix Σ ,

we always know $\text{Tr}(\Sigma) \geq \text{Tr}(\Sigma O)$ for $O \in \text{Od}(\mathbb{R})$

Only when $O = \text{Id}$ we can get " $=$ ".

Such being the case, $\text{Tr}(\Sigma U^T W V) \leq \text{Tr}(\Sigma)$

To minimize $-2\text{Tr}(\Sigma U^T W V)$

we can set $W = UV^T$

3. Sentence classification with BoV

the weighted model train mean-squared-error

is 1.16783707865

the weighted model dev mean-squared-error

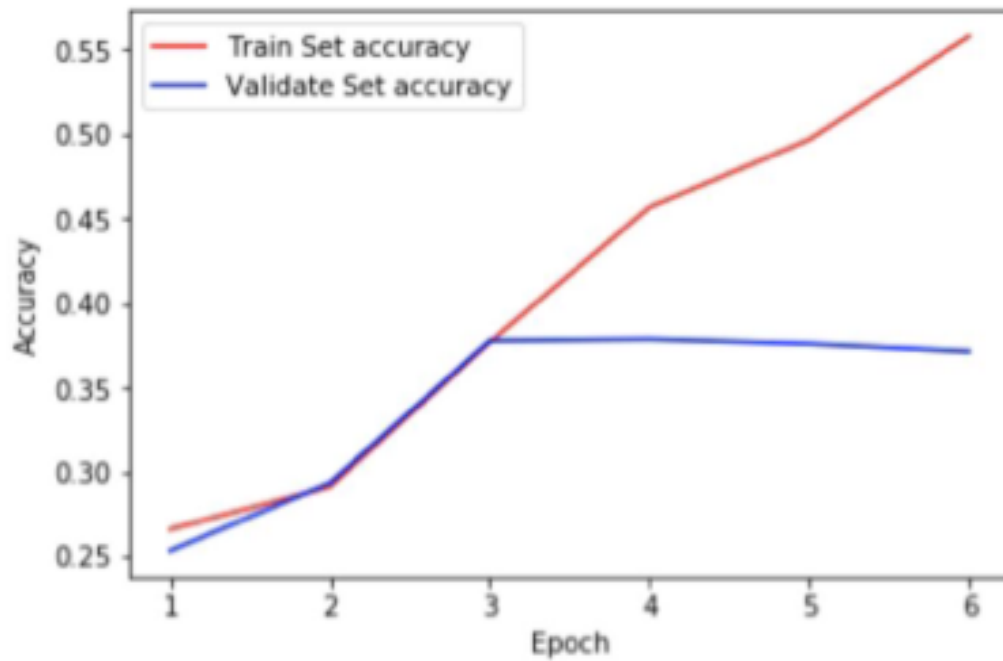
is 1.29427792916

4. Deep learning models for classification

Loss function: categorical-crossentropy

$$L = -\frac{1}{n} \sum_{i=1}^n [y_i \log(\hat{y}_i) + (1-y_i) \log(1-\hat{y}_i)]$$

where y_i is the ground truth and \hat{y}_i is the predicted prob for the truth



creative:

Instead of word vectors, I use the sentence vectors and implement it in LSTM model.

Compared with traditional embedding method, LSTM or RNN can extract more information.

The validate accuracy is higher than previous method.