# Learning to Label Stack Exchange Questions

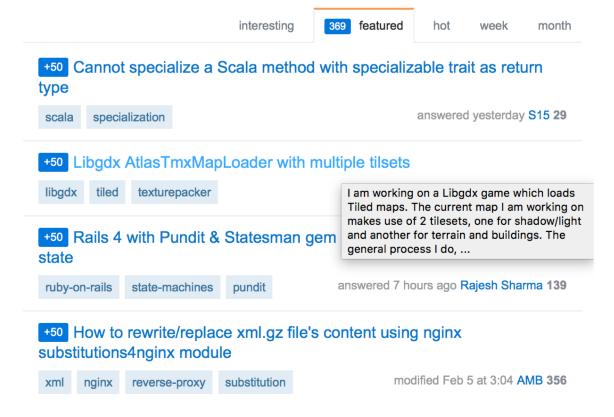
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#### Problem

- Q/A sites get thousands of questions daily
- Questions must be labeled for easy retrieval



### Task

 Automatically assign labels to questions using deep learning and natural language processing

#### How to efficiently iterate over each Entry in a Map?



If I have an object implementing the Map interface in Java and I wish to iterate over every pair contained within it, what is the most efficient way of going through the map?

1282



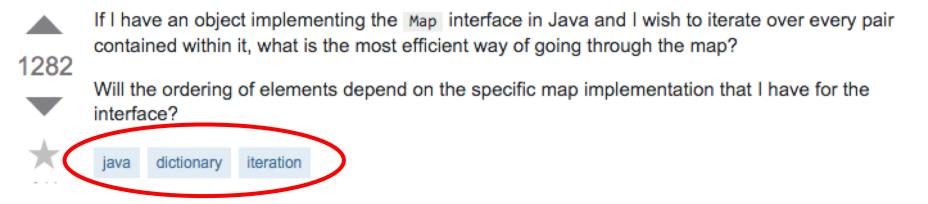
Will the ordering of elements depend on the specific map implementation that I have for the interface?



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How to efficiently iterate over each Entry in a Map?

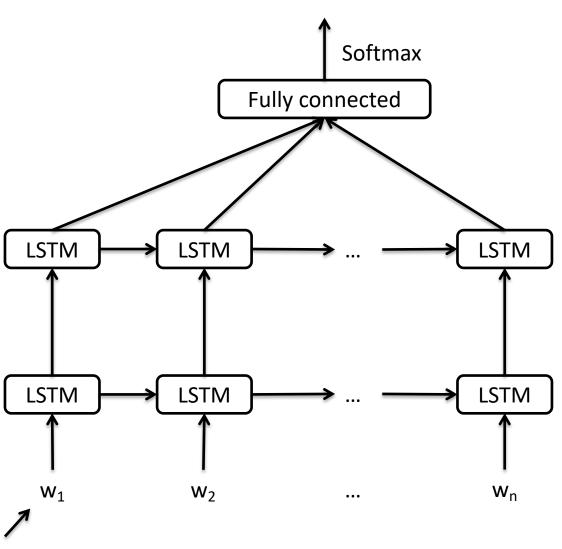


## Approach

- 1. Convert title and questions to word vectors
  - GloVe: Global Vectors for Word Representation\*
  - Pre-trained word vectors on Wikipedia 2014 + Gigaword 5
  - Dimension of vectors: 50
- 2. Train a long-short term memory (LSTM) model
- 3. Predict multiple labels for each example

<sup>\*</sup>http://nlp.stanford.edu/projects/glove/

### Architecture



word embeddings

#### Data

- Stack Exchange data\* (~7GB):
  - Id, title, body, tags
- Statistics:
  - Questions: 6,034,195
  - Unique tags: 42,048
  - Tags/question: 1-5
  - Average no. of tags/question: 2.89

<sup>\*</sup>https://www.kaggle.com/c/facebook-recruiting-iii-keyword-extraction/data

#### Reduced data

- Reduce number of tags:
  - Keep top 10 tags
  - Keep top 100 co-occurring tags with each top tag
  - Use only questions containing these tags
  - Remove stopwords, numbers and punctuation
- Statistics:
  - Questions: 1,220,004
  - Unique tags: 573
  - Maximum question length: 2694

## First approach

- Input data:
  - -1,220,004\*2694\*50 = 164,334,538,800
- Process in batches, reduce question size:
  - Batch size: 10,000 questions
  - Maximum question length: 512
- No improvement after each batch

#### Even more reduced data

- Reduce number of tags:
  - Keep top 10 tags
  - Keep top 10 co-occurring tags with each top tag
  - Use only questions containing these tags
- Statistics:
  - Questions: 582,331
  - Unique tags: 72
  - Maximum question length: 2694

## Data used in experiments

Questions: 200,000

Unique tags: 72

Maximum question length: 400

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## Baseline: N-grams + Logistic Regression

- Extract n-grams from title and body (cleaned text)
- Train a one-vs-rest classifier using Logistic Regression
- Implemented using scikit-learn (Python)

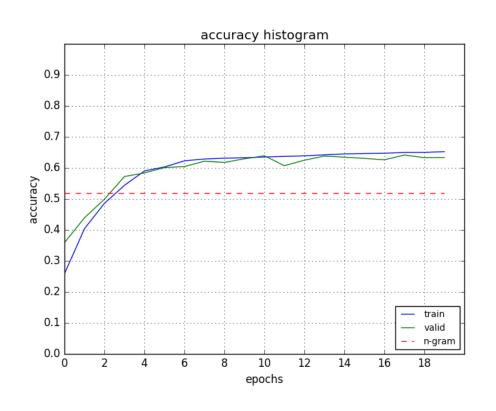
N	Number of features	
1-gram	1,171,964	
{1,2}-gram	5,551,226	

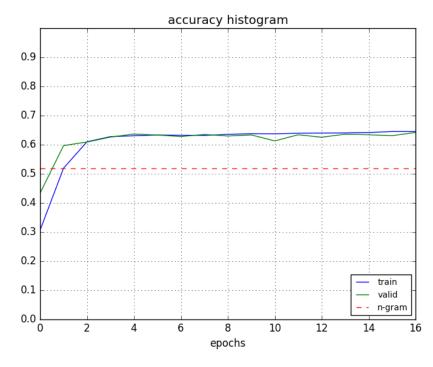
#### Results

- Training set: 180,000 questions (90%)
- Validation set: 20,000 questions (10%)
- Deep models:
  - 128 batch size
  - 18 epochs
  - Rmsprop
- Subset accuracy:
  - the set of predicted labels must exactly match the true set of labels

Method	Validation Accuracy	Categorical Loss Entropy
1-gram + Logistic Regression	0.487	16.66
{1,2}-gram + Logistic Regression	0.518	16.09
2 layer LSTM	0.642	3.32
3 layer LSTM	0.641	3.48

## Results (accuracy)



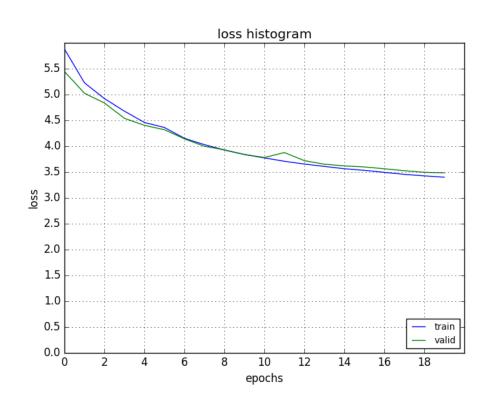


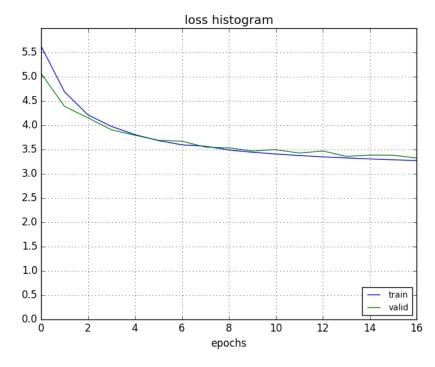
Questions: 100,000

3-layer LSTM

Questions: 200,000

## Results (categorical loss)



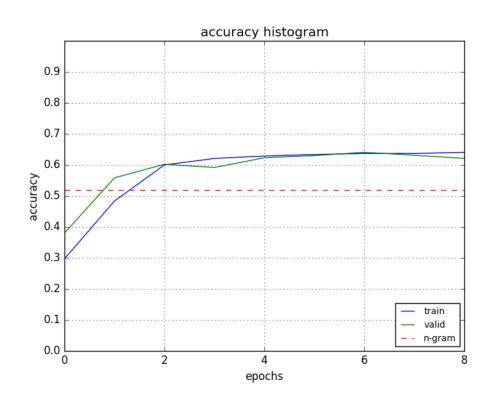


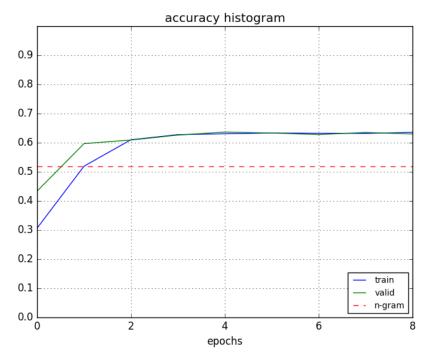
Questions: 100,000

3-layer LSTM

Questions: 200,000

## Results II (accuracy)



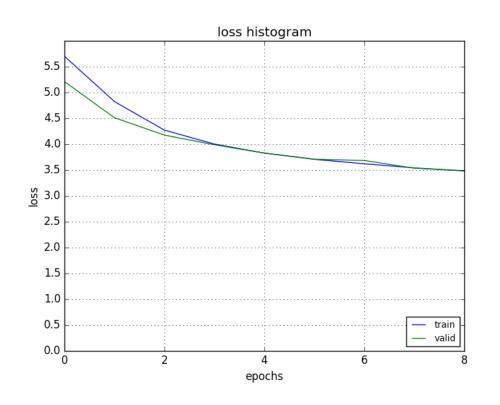


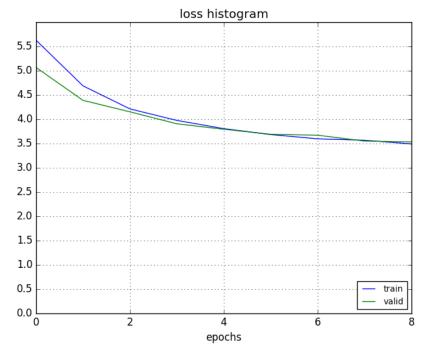
Questions: 200,000

3-layer LSTM

Questions: 200,000

## Results II (categorical loss)





Questions: 200,000

3-layer LSTM

Questions: 200,000

#### References

- S. Hochreiter, J. Schmidhuber. Long short-term memory. Neural computation, 2007.
- J. Pennington, R. Socher, C. Manning. GloVe: Global Vectors for Word Representation, 2014.
- X. Zhang, J. Zhao, Y. LeCun. Character-level Convolutional Networks for Text Classification. ArXiv: 1509.01626v2, 2015.
- LSTM Networks for Sentiment Analysis. deeplearning.net.