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Project title:

Tag suggestion for Stack Overflow

Dataset:

Step 1: We obtained question IDs from the StackLite dataset on kaggle.

Step 2: To obtain question body and tags corresponding to each question id we used the StackAPI. The code to fetch the complete dataset is in fetch dataset.py

We have attached the raw dataset with question id, body and tags. Please find it in file dataset.txt

Following is a snippet from the raw dataset:

```
193896
;;;;;
c linux reverse-engineering decompiling assembly
;;;;

cpread as searching for a decompiler for a C program. The binary is a 32-bit x86 Linux executable. Objdump works
fine, so basically I am searching for something which attempts to reconstruct the C source from the asm source.
;;;;

|||||
194121
;;;;

pph deprecated
;;;;

communicate with database. We created a wrapper to it that would do something else beside just calling the
original function and I did a mass search-replace in the entire code to make sure that wrapper is used instead.

cp>Now, how do we prevent usage of ibase_* functions in the future?

cp>Preferably, I'd like to still have them available, but make it throw a NOTICE or WARNING when it is used.

cp>A solution in pure PHP (not needing to compile a custom version of PHP) is preferred.
;;;;;

|||||
```

We have used ;;;; to separate categories (id, tags, body) within a post and $|\cdot|\cdot|\cdot|$ to separate posts.

We have implemented two methods to predict tags for Stack Overflow posts:

Deep model approach

The training part of this approach is done using train_model.py

To prepare data for training, we need to separate documents and ground truth labels in two files - processed docs.txt and labels.txt

The code for this is available in preprocess.py

How do we process the raw posts?

- 1. Select top 118 tags based on their occurrence count in the tags dictionary and only consider posts which has at least one tag belonging to the set. Furthermore, for each post we only consider the top tags and discard the rest.
- 2. Use BeautifulSoup to filter out all elements containing code.

```
question_body = categories[2]
soup = BeautifulSoup(question_body, 'html.parser')

# Remove all tags with a class or id containing the word snippet
# Later use these snippets to predict the programming language ^_^
for snippet_tag in soup.find_all(attrs={'class': re.compile('snippet')}):
    snippet_tag.decompose()

for snippet_tag in soup.find_all(attrs={'id': re.compile('snippet')}):
    snippet_tag.decompose()

# Remove all the  ...  tags
for extra in soup('pre'):
    extra.extract()
```

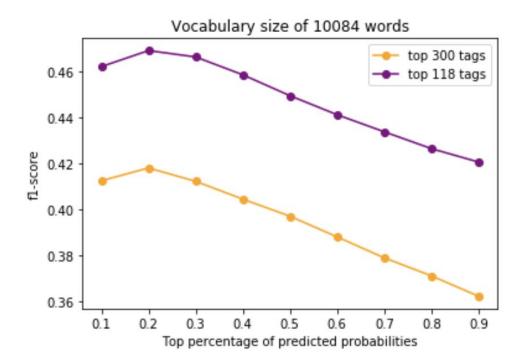
3. Filter out numbers, non-alphanumeric characters, punctuations appended to tokens and stopwords.

In train model.py we convert processed docs to tf-idf vectors:

```
data = open('./processed_docs.txt', 'r').read().split('\n')[:-1]
vectorizer = TfidfVectorizer(input='content', token_pattern=r'\S+', analyzer='word', min_df=0.0002)
X = vectorizer.fit_transform(data).todense()
```

We input the obtained tf-idf vectors to a deep model which outputs a binary vector of size number of total tags (118 in our case). We use softmax activation in the last layer and categorical_crossentropy loss to train our model. More details can be found in train model.py

Results from the deep model approach:



We get a maximum F1-score of 0.48 when we consider top 118 tags and threshold the top 20% of the values in the last layer of the neural network to 1.

Supervised LDA approach

Here we approach the problem as one of Topic Modelling. We preprocess our data in a similar manner as above but we store it in the following format:

[label1, label2, ...] document tokens

Sample processed dataset looks like:

[sql,database,version-control,oracle]often run following problem work changes project require new tables columns database make database modifications continue work usually remember write changes replicated live system however don't always remember i've changed don't always remember write make push live system get big obvious error newcolumnx ugh regardless fact may best practice situation version control system databases don't care specific database technology want know one exists happens work ms sql server great [sql,asp.net,xml]anyone got experience creating sql-based asp.net site-map providers i've got default xml file web.sitemap working properly menu sitemappath controls i'll need way users site create modify pages dynamically need tie page viewing permissions standard asp.net membership system well

We ran the Supervised LDA code on 800 posts since it was not able to scale to the full dataset of 35k posts. The processed dataset can be found in 800_llda_dataset.txt

The code is present in 11da.py

Results:

Here we are showing the top words corresponding to certain topics.

```
label 28 : vim
  label 0 : common
                    -- label 8 : mysql
                                                                        multithreading
like: 0.0128
                                              vim: 0.1598
                                                                      static: 0.0468
                    mysql: 0.1427
use: 0.0123
                                              emacs: 0.0959
                                                                      reproduce: 0.0312
                    database: 0.0897
would: 0.0115
                                                                      threads: 0.0234
                                              version: 0.0959
way: 0.0113
                    data: 0.0815
                                                                      queue: 0.0234
                                              prime: 0.0320
code: 0.0084
                    server: 0.0326
                                                                      thread: 0.0234
know: 0.0076
                                              editors: 0.0320
                    administrator: 0.0245
want: 0.0072
                                                                      variables: 0.0234
                    compatibility: 0.0204
                                              graphical: 0.0320
need: 0.0066
                                                                      safe: 0.0234
                    backups: 0.0204
                                              macvim: 0.0320
work: 0.0064
                                                                      ram: 0.0156
                    log: 0.0163
                                              cocoa: 0.0320
file: 0.0058
                                                                      mutex: 0.0156
                                              carbonemacs: 0.0320
net: 0.0052
                    binary: 0.0163
                                                                      watson: 0.0156
server: 0.0049
                    replicate: 0.0122
                                              xemacs: 0.0320
                                                                      alloc: 0.0156
best: 0.0048
                    matter: 0.0122
                                              aquamacs: 0.0320
                                                                      cpu: 0.0156
anyone: 0.0048
                    dump: 0.0122
                                              tough: 0.0320
                                                                      threaded: 0.0156
user: 0.0047
```

Instructions to run code:

Deep model

```
a. python3 preprocess.py
This will run on dataset.txt and create the processed_docs.txt
and labels.txt
```

- b. python3 train model.py
- c. python test model.py

[Extra: plot.py to visualise results]

Supervised LDA model

```
a. python llda.py -f 800_llda_dataset.txt -k 171
    -f dataset filename
    -k number of tags
(For the given dataset file, number of tags is 171)
```