04 bonus translation

September 29, 2021

```
[]: from notebook.services.config import ConfigManager
cm = ConfigManager()
cm.update('livereveal', {
        'width': 1280,
        'height': 800,
        'scroll': True
})
```

0.1 Imports & Config

```
[39]: import tensorflow as tf
      from sklearn.model_selection import train_test_split
      from pathlib import Path
      import sys
      from os import environ
      from inspect import getfile, currentframe
      from os.path import join, dirname, abspath
      from gensim.models import Word2Vec, KeyedVectors
      import pandas as pd
      import numpy as np
      from numpy.linalg import norm
      from numpy.random import permutation
      from sklearn.manifold import TSNE
      import matplotlib.pyplot as plt
      from matplotlib.ticker import FuncFormatter
      plt.style.use('ggplot')
```

```
[2]: def batches(1, n):
    """ Yield successive n-sized chunks from l."""
    for i in range(0, len(1), n):
        yield l[i:i + n]
```

```
[41]: LANGUAGES = ['en', 'es']
translation_path = Path('translation')
if not translation_path.exists():
    translation_path.mkdir(parents=True, exist_ok=True)
```

0.2 Load word2vec Models

[11]: 'Missing: 13.17%'

 \rightarrow sort_index(1)

matches.head()

matches.es_id = matches.es_id.astype(int)

```
[6]: model, words, vectors = {}, {}, {}
      for language in LANGUAGES:
          file_name = 'word2vec/word_vectors/vectors {}.bin'.format(language)
          model[language] = KeyedVectors.load_word2vec_format(file_name,
                                                              binary=True,
                                                              unicode_errors='ignore')
          words[language] = model[language].index2word
          vectors[language] = model[language].vectors
          vectors[language] /= norm(vectors[language], axis=1).reshape(-1, 1)
     0.3 Translate Top 10k English Terms
 [7]: to_translate = words['en'][:10000]
      pd.Series(to_translate).to_csv(translation_path / 'words_en.txt', index=False)
 [8]: translated = pd.Series((translation_path / 'words_es.txt').read_text().

¬split('\n'))
      translated.count()
 [8]: 10000
 [9]: matches = pd.DataFrame({'en': to_translate, 'es': translated})
      matches.head()
 [9]:
           en
                  es
      0 </s> </s>
      1
         the
                  el
      2
          of
                  de
      3
         and
                   У
           in
                  en
[10]: word2id = {}
      for language in LANGUAGES:
          word2id[language] = {w: i for i, w in enumerate(words[language])}
[11]: | matches['es_id'] = matches.es.map(word2id['es'])
      'Missing: {:.2%}'.format(matches.es_id.isnull().sum()/len(matches))
```

```
2
```

[12]: matches = matches.dropna().reset_index().rename(columns={'index': 'en_id'}).

```
[12]:
         en en_id es es_id
     0
       the
                 1
                    el
                            2
     1
         of
                 2
                            1
                    de
     2 and
                 3
                            5
                     У
                            4
     3
         in
                 4 en
                            6
         to
                 5
```

```
[13]: matches.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8683 entries, 0 to 8682
Data columns (total 4 columns):
en         8683 non-null object
en_id         8683 non-null int64
es         8683 non-null object
es_id         8683 non-null int64
dtypes: int64(2), object(2)
memory usage: 271.4+ KB
```

0.4 Learn Translation Matrix

0.4.1 Model Settings

```
[14]: BATCH_SIZE = 100
    TRAIN_SIZE = 5000
    EARLY_STOP = 5
    SOURCE_DIM = vectors['en'].shape[1]
    TARGET_DIM = vectors['es'].shape[1]
```

0.4.2 Train & Test Sets for Source & Target Embeddings

```
[15]: source_id = matches.en_id.values
source_train = vectors['en'][source_id[:TRAIN_SIZE]]
source_test = vectors['en'][source_id[TRAIN_SIZE:]]

target_id = matches.es_id.values
target_train = vectors['es'][target_id[:TRAIN_SIZE]]
target_test = vectors['es'][target_id[TRAIN_SIZE:]]
```

0.4.3 TensorFlow Graph

```
[16]: tf.logging.set_verbosity(tf.logging.INFO)

with tf.name_scope('inputs'):
    s = tf.placeholder(tf.float32, shape=[None, SOURCE_DIM], name='source')
    t_ = tf.placeholder(tf.float32, shape=[None, TARGET_DIM], name='target')
```

```
with tf.name_scope('parameters'):
    T = tf.Variable(
        tf.random_normal([SOURCE_DIM, TARGET_DIM]),
        name='translation_matrix'
)
    tf.summary.histogram('histogram', T)

with tf.name_scope('inference'):
    t = tf.matmul(s, T, name='projection')
    # Regularized L2 Loss
    loss = tf.nn.12_loss(t - t_, name='12_loss') + .01 * tf.nn.12_loss(T)

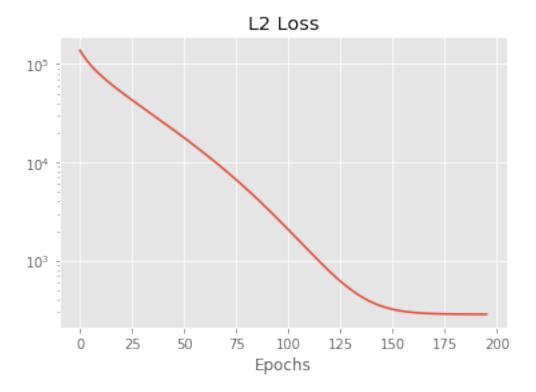
with tf.name_scope('training'):
    optimizer = tf.train.AdamOptimizer()
    train = optimizer.minimize(loss)
```

0.4.4 Run TensorFlow Model

```
[17]: with tf.Session().as_default() as sess:
          tf.global_variables_initializer().run()
          test_loss = np.inf
          early_stop = 0
          losses = []
          while True:
              s_train, s_test, t_train, t_test = train_test_split(source_train,
                                                                    target_train,
                                                                    test size=0.2,
                                                                    random_state=42)
              for idx in batches(permutation(s_train.shape[0]), BATCH_SIZE):
                  train.run(feed_dict={s: s_train[idx], t_: t_train[idx]})
              loss_ = sess.run(loss, feed_dict={s: s_test, t_: t_test})
              losses.append(loss_)
              if loss_ < test_loss:</pre>
                  test_loss = loss_
                  early_stop = 0
              else:
                  early_stop += 1
              if early_stop == EARLY_STOP:
                  break
          translation_matrix = T.eval(sess)
```

0.4.5 Review Results

```
[18]: ax = pd.Series(losses).plot(title='L2 Loss', logy=True)
ax.set_xlabel('Epochs');
```



```
[19]: translation_matrix.shape
```

[19]: (300, 300)

0.4.6 Evaluate Translation Accuracy

Train Accuracy

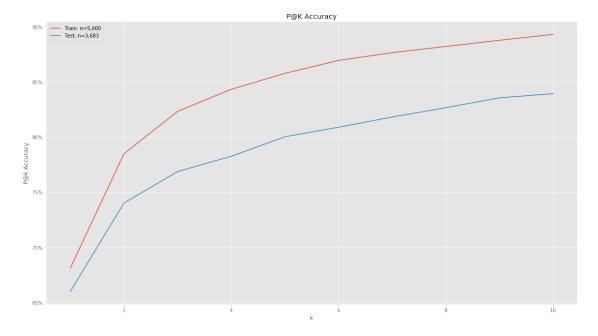
Test Accuracy

```
[32]: ranking, nn_test = get_accuracy(source_test, target_test, translation_matrix)
test_accuracy = ranking.value_counts().sort_index()
test_precision = test_accuracy.cumsum().div(len(ranking)).iloc[:10].iloc[:10]
test_precision = test_precision.to_frame('Test: n={:,d}'.format(len(ranking)))
```

Plot Results

```
[34]: result = pd.concat([train_precision, test_precision], axis=1)
    ax = result.plot(title='P@K Accuracy', figsize=(16,9))
    ax.set_xlabel('k')
    ax.set_ylabel('P@k Accuracy')
    ax.yaxis.set_major_formatter(FuncFormatter(lambda y, _: '{:.0%}'.format(y)))

plt.tight_layout()
    plt.gcf().savefig('assets/translation_accuracy.png', dpi=300);
```



View Translations

```
[35]: top_5_preds = pd.DataFrame(nn_train[:, :5], index=matches.en.iloc[:TRAIN_SIZE])
      for col in top_5_preds.columns:
          top_5_preds[col] = top_5_preds[col].map(matches.es.to_dict())
      print(top_5_preds.head(10))
             0
                     1
                                      3
                                                    4
     en
                            la
     the
            el
                    el
                                     de
                                                   de
     of
            de
                    de
                            el
                                     el
                                                   la
     and
                            de
                                     de
                                                  con
             У
                    У
     in
                            de
                                     de
            en
                    en
     to
                                    que directamente
             a
                          para
                     a
                            el
                                     el
     а
            ıın
                    un
                                                  que
                             es
     was
           era estaba
                                     es
                                                   es
     is
            es
                             es existe
                                               existe
     for para además
                        además
                                 además
                                               además
            en
[36]: with pd.HDFStore(translation_path / 'results.h5') as store:
          store.put('translation_matrix', pd.DataFrame(translation_matrix))
          store.put('top_5_preds', top_5_preds)
          store.put('accuracy', result)
     0.4.7 Visualize Bilingual Vector Space in tensorboard
     Embeddings
[37]: LIMIT = 2500
      source_projection = np.dot(source_test, translation_matrix)
      projector_data = pd.DataFrame(np.vstack((source_projection[:LIMIT],
                                                target_test[:LIMIT])))
      projector_data.to_csv(translation_path / 'embeddings.tsv', index=False,__
       →sep='\t', header=None)
      projector_data.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 5000 entries, 0 to 4999
     Columns: 300 entries, 0 to 299
     dtypes: float32(300)
     memory usage: 5.7 MB
     Meta Data
[38]: labels = pd.melt(matches.iloc[TRAIN_SIZE:TRAIN_SIZE+LIMIT].loc[:, ['en', 'es']],
                       var_name='language',
                       value name='word')
      labels.to_csv(translation_path / 'meta_data.tsv', index=False, sep='\t')
      labels.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5000 entries, 0 to 4999
Data columns (total 2 columns):
language 5000 non-null object
word 5000 non-null object

dtypes: object(2)
memory usage: 78.2+ KB

0.5 Resources

• Exploiting similarities among languages for machine translation