

# **Matching Job Description to Job Titles: Machine Learning Mechanism**

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The input dataset here contains two parts: a) cleaned job posting as X, which has only the extracted information of required technical skills, job responsibilities, job qualifications; b) cleaned job titles as Y, whose stopwords, irrelevant description and job-level components are all removed (but the descriptive component and objective component haven't been seperated).

**X:** Job descriptions are encoded to squence, and post-padded with 0s, then serialize with time step of 15.

**Y:** The job labels are encoded with [fastText](https://github.com/facebookresearch/fastText) (<https://github.com/facebookresearch/fastText>).

I randomly sampled 80% of original dataset to form training dataset and leave the left to be testing dataset.

**training:** I chose to minimize cosine proximity loss between vectorized job description (by a trainable LSTM) and vectorized job titles (by a un-trainable fastText).

I tested 21 LSTM variants to find out optimal structures to vectorize job description, among them I chose the LSTM-19, which has the following structures:

```
def create_LSTM(input_dim,output_dim,time_steps=10,embedding_matrix=[]):
    batch_size = 1
    # inputs.shape = (batch_size, time_steps, input_dim)
    inputs = Input(shape=(batch_size,time_steps, input_dim))
    if embedding_matrix != []:
        embedding_layer = Embedding(embedding_matrix.shape[0],
                                    embedding_matrix.shape[1],
                                    weights=[embedding_matrix],
                                    input_shape=(input_dim,),
                                    trainable=False)
        x = embedding_layer(inputs)
        x = Reshape([embedding_matrix.shape[1],input_dim])(x)
    else:
        x = Reshape([time_steps,input_dim])(inputs)

    x = Bidirectional(LSTM(100, return_sequences=True))(x)
    x = BatchNormalization()(x)
    x = Activation('relu')(x)

    x = attention_3d_block(x,input_dim=200)
    x = BatchNormalization()(x)
    x = Activation('relu')(x)

    #LSTM OUT
    x = Bidirectional(LSTM(150))(x)
    x = BatchNormalization()(x)
    x = Activation('relu')(x)

    #NN OUT
    x = Dense(output_dim, activation='tanh')(x)
    model = Model(input=inputs, output=x)
    print(model.summary())
    return model
```

**evaluation:** Except for looking at the predicted titles by human intuition, I made a score called *Ranking Percentage Score*, which is the percentage of correct label located in the prediction sequence, which is ordered by predicted cosine proximity.

## 1. Load the packages

```

In [1]: import numpy as np
        from numpy.random import seed
        seed(1)
        from tensorflow import set_random_seed
        set_random_seed(2)
        from keras.preprocessing import sequence
        from keras.preprocessing.text import Tokenizer
        import pandas as pd
        import json
        import os
        import datetime
        try:
            CWDIR = os.path.abspath(os.path.dirname(__file__))
        except:
            CWDIR = os.getcwd()

        from keras import metrics
        from keras.optimizers import SGD, Adam, RMSprop
        pd.options.mode.chained_assignment = None # default='warn'

        def import_local_package(addr_pkg,function_list=[]):
            #import local package by address
            #it has to be imported directly in the file that contains functions
            required the package, i.e. it cannot be imported by from ../utils import i
            import_local_package
            import importlib.util
            spec = importlib.util.spec_from_file_location('pkg', addr_pkg)
            myModule = importlib.util.module_from_spec(spec)
            spec.loader.exec_module(myModule)
            if len(function_list)==0:
                import re
                function_list = [re.search('[a-zA-Z]*.*',x).group() for x
in dir(myModule) if re.search('[a-zA-Z]',x) != None]

                for _f in function_list:
                    try:
                        eval(_f)
                    except NameError:
                        exec("global {}; {} = getattr(myModule,'{}'.format(
_f,_f,_f)) #exec in function has to use global in 1 line
                        print("{} imported".format(_f))

                return
        import_local_package(os.path.join(CWDIR,'./lib/utils.py'),['train_model','l
oad_model','get_rank_df'])
        import_local_package(os.path.join(CWDIR,'./data/lib/prepare_data.py'),[])

```

```
/home/nyartsgnaw/anaconda3/lib/python3.6/site-packages/h5py/__init__.py:36:
FutureWarning: Conversion of the second argument of issubdtype from `float`
to `np.floating` is deprecated. In future, it will be treated as `np.float6
4 == np.dtype(float).type`.
    from ._conv import register_converters as _register_converters
Using TensorFlow backend.

train_model imported
load_model imported
get_rank_df imported
Word2Vec imported
WordNetLemmatizer imported
get_similar_words imported
get_time_series imported
nltk imported
prepare_data imported
remove_stopwords imported
sent_tokenize imported
stopwords imported
wn imported
word_tokenize imported
```

## 2. Setup the experiment

```

In [2]: # inputs
path_exp = os.path.join(CWDIR, './experiments/exp_logs.xlsx')
df_exp = pd.read_excel(path_exp)
with open(os.path.join(CWDIR, './experiments/.idx'), 'r') as f:
    idx = int(f.read())
idx = 35

exp = df_exp.iloc[idx]

# setup model parameters
start_time = datetime.datetime.now()
EXP_ID = exp['EXP_ID'] #the name for this experiment
MODEL_ID = exp['MODEL_ID'] #model framework
OUTPUT_DIM = int(exp['OUTPUT_DIM']) # LSTM output vector dimension, should
match that of Word2Vec of labels
INPUT_DIM = int(exp['INPUT_DIM']) # LSTM input vector dimension, length of
tokens cut from original data texts for each record
TIME_STEPS = int(exp['TIME_STEPS']) #for LSTM sequential
N_EPOCH = int(exp['N_EPOCH']) #for LSTM
PATIENCE = int(exp['PATIENCE']) #for LSTM
TRAIN_MODEL = int(exp['IS_TRAIN'])
LOSS=exp['LOSS_FUNC']
print(exp)
import_local_package(os.path.join(CWDIR, './experiments/models/{}.py'.format(
(MODEL_ID))), [])

# setup logging address
path_vectors = os.path.join(CWDIR, './logs/models/vectors_JT-{}.csv'.format(
INPUT_DIM))
path_model = os.path.join(CWDIR, './logs/models/LSTM_{}.model'.format(EXP_ID
))
path_eval = os.path.join(CWDIR, './logs/eval/LSTM_eval_{}.csv'.format(EXP_ID
))
path_training_model = os.path.join(CWDIR, './logs/models/LSTM_train_{}.model
'.format(EXP_ID))
path_training_log = os.path.join(CWDIR, './logs/train_logs/LSTM_logs{}.csv'.
format(EXP_ID))
#     if os.path.isfile(path_training_log):
#         os.remove(path_training_log)
os.system('mkdir -p {}'.format(os.path.join(CWDIR, './logs/eval/')))
os.system('mkdir -p {}'.format(os.path.join(CWDIR, './logs/models/')))
os.system('mkdir -p {}'.format(os.path.join(CWDIR, './logs/train_logs/')))
# fix random seed for reproducibility
np.random.seed(7)

```

```

EXP_ID          35
N_EPOCH         200
PATIENCE        100
IS_TRAIN        1
LOSS_FUNC       cosine_proximity
OUTPUT_DIM      200
INPUT_DIM       200
TIME_STEPS      15
MODEL_ID        LSTM_19
IS_RUN          0
RANK_SCORE      0.506717
QUIT_LOSS       -0.876086
QUIT_MSE        0.00585117
QUIT_EPOCH      199
N_PARAMS        765200
start_time      2018-08-06 20:03:05
end_time        2018-08-06 21:32:46
note            Bi-LSTM
Name: 35, dtype: object
Activation imported
BatchNormalization imported
Bidirectional imported
Conv1D imported
Conv2D imported
Conv2DTranspose imported
Convolution3D imported
Dense imported
Dropout imported
Embedding imported
Flatten imported
GaussianDropout imported
GaussianNoise imported
Input imported
K imported
LSTM imported
Lambda imported
LeakyReLU imported
MaxPooling2D imported
Model imported
Permute imported
RepeatVector imported
Reshape imported
Sequential imported
TimeDistributed imported
UpSampling1D imported
UpSampling2D imported
UpSampling3D imported
attention_3d_block imported
create_LSTM imported
initializers imported
merge imported
regularizers imported

```

### 3. Load/prepare the data

```
In [3]: # read the label Ys
path_vectors = os.path.join(CWDIR, './logs/models/vectors_JT-{}.csv'.format(
INPUT_DIM))
if not os.path.isfile(path_vectors):
    os.system('python {}'.format(os.path.join(CWDIR, './lib/train_fasttext.p
y'))))

labels = pd.read_csv(path_vectors).values

# read the data Xs
path_data = os.path.join(CWDIR, './data/df_all.csv')
df_all = pd.read_csv(path_data)

# encode the data into sequence
tokenizer = Tokenizer()
tokenizer.fit_on_texts([' '.join(df_all['texts'])])
data = tokenizer.texts_to_sequences(df_all['texts'])
data = sequence.pad_sequences(data, padding='post', truncating='post', maxlen
=INPUT_DIM) # truncate and pad input sequences

# prepare trainig/testing data/labels
judge = (df_all['split']=='train').values
train_data = data[judge]
test_data = data[~judge]
train_labels = labels[judge]
test_labels = labels[~judge]

# serialize the data/labels to model input/output format
X_train, Y_train = get_time_series(train_data, train_labels, TIME_STEPS, 0)
X_test, Y_test = get_time_series(test_data, test_labels, TIME_STEPS, 0)
```

#### 4. Train the model

```
In [4]: # create/load the model
from keras.models import load_model
if os.path.isfile(path_model):
    model = load_model(path_model)
else:
    embedding_matrix = []
    # embedding_matrix = load_embedding_fasttext(path_JD)
    # model = create_LSTM(input_dim=INPUT_DIM,output_dim=OUTPUT_DIM,embedding_matrix=embedding_matrix)
    model = create_LSTM(input_dim=INPUT_DIM,output_dim=OUTPUT_DIM,time_steps=TIME_STEPS,embedding_matrix=embedding_matrix)
    # train the model
    N_EPOCH = 1
    if TRAIN_MODEL == True:
        adam=Adam(lr=0.005, beta_1=0.9, decay=0.001)
        model.compile(loss=LOSS, optimizer=adam, metrics=['mse', 'cosine_proximity'])
        model = train_model(model,X_train=X_train.reshape([-1,1,TIME_STEPS,INPUT_DIM]),\
                                Y_train=Y_train,\
                                verbose=1,n_epoch=N_EPOCH,validation_split=0.1,patience=PATIENCE,\
                                model_path=path_training_model,\
                                log_path=path_training_log)
        model.save(path_model)
```

```
/home/nyartsgnaw/anaconda3/lib/python3.6/site-packages/keras/engine/topology.py:1271: UserWarning: The `Merge` layer is deprecated and will be removed after 08/2017. Use instead layers from `keras.layers.merge`, e.g. `add`, `concatenate`, etc.
```

```
    return cls(**config)
```

```
WARNING:tensorflow:Variable *= will be deprecated. Use variable.assign_mul if you want assignment to the variable value or 'x = x * y' if you want a new python Tensor object.
```

```
1
```

```
Train on 12436 samples, validate on 1382 samples
```

```
Epoch 1/1
```

```
- 28s - loss: -7.0686e-01 - mean_squared_error: 0.0069 - cosine_proximity: -7.0686e-01 - val_loss: -6.5758e-01 - val_mean_squared_error: 0.4877 - val_cosine_proximity: -6.5758e-01
```

## 5. Evaluate the model



```
In [9]: # evaluate the model by ranking percentage score
yhat = model.predict(X_test.reshape([-1,1,TIME_STEPS,INPUT_DIM]))[:5]
# prepare the original testing labels
titles_all = df_all.titles.values
titles_test = titles_all[~judge][:5]
Y = np.concatenate([Y_test,Y_train])

df = get_rank_df(yhat,titles_test,Y,titles_all)
#      df = get_rank_df(yhat,titles_test,Y_test,titles_test)
df.to_csv(path_eval,index=False)
```

Job: full-time community connections intern paid internship  
software engineer  
data administrator 1 it division financial monitoring center  
radio optimization senior engineer  
head financial department  
safety manager  
procurement specialist  
sales senior specialist commercial directorate  
web developer  
domestic expert international exposure accounting  
human resources senior specialist  
Percentage\_Rank of 0: 0.7865214116010895

Job: bcc specialist  
software engineer  
data administrator 1 it division financial monitoring center  
radio optimization senior engineer  
head financial department  
safety manager  
procurement specialist  
sales senior specialist commercial directorate  
domestic expert international exposure accounting  
human resources senior specialist  
education officer  
Percentage\_Rank of 1: 0.7081184446891117

Job: chauffeur fsn-3 fp-bb\*  
software engineer  
data administrator 1 it division financial monitoring center  
radio optimization senior engineer  
head financial department  
safety manager  
procurement specialist  
sales senior specialist commercial directorate  
domestic expert international exposure accounting  
human resources senior specialist  
education officer  
Percentage\_Rank of 2: 0.9626818102798864

Job: demographic analysis workshop  
software engineer  
data administrator 1 it division financial monitoring center  
radio optimization senior engineer  
head financial department  
safety manager  
procurement specialist  
sales senior specialist commercial directorate  
human resources senior specialist  
education officer  
administrative assistant  
Percentage\_Rank of 3: 0.8447586486643102

Job: receptionist  
software engineer  
data administrator 1 it division financial monitoring center  
radio optimization senior engineer  
head financial department  
safety manager  
procurement specialist  
sales senior specialist commercial directorate  
technical writer  
domestic expert international exposure accounting  
web developer

## 6. Log the experiment

```
In [11]: # log the results
exp['start_time'] = str(start_time.replace(microsecond=0))
exp['end_time'] = str(datetime.datetime.now().replace(microsecond=0))

try:
    exp['RANK_SCORE'] = df['rank'].mean()
except Exception as e:
    print(e)

try:
    df_log = pd.read_csv(path_training_log)
    exp['QUIT_LOSS'] = df_log.iloc[-1]['loss']
    exp['QUIT_EPOCH'] = df_log.iloc[-1]['epoch']
    exp['QUIT_MSE'] = df_log.iloc[-1]['mean_squared_error']
except Exception as e:
    print(e)

try:
    exp['N_PARAMS'] = model.count_params()
except Exception as e:
    print(e)
```

Error tokenizing data. C error: Expected 5 fields in line 202, saw 7