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Title: Assignment 5: Implement the Continuous Bag of Words (CBOW) Model

```
#importing libraries
from keras.preprocessing import text
from keras.utils import np utils
from keras.preprocessing import sequence
from keras.utils import pad sequences
import numpy as np
import pandas as pd
#taking random sentences as data
data = """Deep learning (also known as deep structured learning) is part of a broader family
Deep-learning architectures such as deep neural networks, deep belief networks, deep reinforc
dl data = data.split()
#tokenization
tokenizer = text.Tokenizer()
tokenizer.fit on texts(dl data)
word2id = tokenizer.word index
word2id['PAD'] = 0
id2word = {v:k for k, v in word2id.items()}
wids = [[word2id[w] for w in text.text to word sequence(doc)] for doc in dl data]
vocab size = len(word2id)
embed size = 100
window size = 2
print('Vocabulary Size:', vocab size)
print('Vocabulary Sample:', list(word2id.items())[:10])
    Vocabulary Size: 75
     Vocabulary Sample: [('learning', 1), ('deep', 2), ('networks', 3), ('neural', 4), ('and
    <
#generating (context word, target/label word) pairs
def generate context word pairs(corpus, window size, vocab size):
    context length = window size*2
    for words in corpus:
        sentence length = len(words)
        for index, word in enumerate(words):
            context words = []
```

```
label word = []
            start = index - window size
            end = index + window size + 1
            context_words.append([words[i]
                                 for i in range(start, end)
                                 if 0 <= i < sentence length
                                 and i != index])
            label word.append(word)
            x = pad sequences(context words, maxlen=context length)
            y = np utils.to categorical(label word, vocab size)
            yield (x, y)
i = 0
for x, y in generate context word pairs(corpus=wids, window size=window size, vocab size=voca
   if 0 not in x[0]:
        # print('Context (X):', [id2word[w] for w in x[0]], '-> Target (Y):', id2word[np.argw
       if i == 10:
            break
        i += 1
#model building
import keras.backend as K
from keras.models import Sequential
from keras.layers import Dense, Embedding, Lambda
cbow = Sequential()
cbow.add(Embedding(input dim=vocab size, output dim=embed size, input length=window size*2))
cbow.add(Lambda(lambda x: K.mean(x, axis=1), output shape=(embed size,)))
cbow.add(Dense(vocab size, activation='softmax'))
cbow.compile(loss='categorical crossentropy', optimizer='rmsprop')
print(cbow.summary())
# from IPython.display import SVG
# from keras.utils.vis utils import model to dot
# SVG(model to dot(cbow, show shapes=True, show layer names=False, rankdir='TB').create(prog=
     Model: "sequential"
```

Layer (type)	Output Shape	Param #
		==========
embedding (Embedding)	(None, 4, 100)	7500
lambda (Lambda)	(None, 100)	0
dense (Dense)	(None, 75)	7575

Total params: 15,075 Trainable params: 15,075 Non-trainable params: 0

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None

```
for epoch in range(1, 6):
   loss = 0.
   i = 0
   for x, y in generate_context_word_pairs(corpus=wids, window_size=window_size, vocab_size=
       i += 1
       loss += cbow.train on batch(x, y)
       if i % 100000 == 0:
            print('Processed {} (context, word) pairs'.format(i))
   print('Epoch:', epoch, '\tLoss:', loss)
   print()
    Epoch: 1
                   Loss: 433.88288593292236
    Epoch: 2
                    Loss: 428.80262994766235
    Epoch: 3
                    Loss: 425.27664613723755
    Epoch: 4
                    Loss: 422.0698399543762
```

```
weights = cbow.get_weights()[0]
weights = weights[1:]
print(weights.shape)
```

pd.DataFrame(weights, index=list(id2word.values())[1:]).head()

Loss: 419.70696926116943

(74, 100)

Epoch: 5

	0	1	2	3	4	5	6
deep	-0.023805	0.014639	0.007089	0.012630	-0.019672	-0.024277	0.032819
networks	0.037381	0.012409	-0.048280	0.010494	0.014020	-0.050136	-0.056002
neural	0.010798	-0.013446	-0.036526	0.040533	-0.028618	-0.038640	0.031019
and	-0.042606	-0.042891	0.007063	-0.033965	-0.027141	0.030406	-0.049688
as	-0.010727	-0.016468	-0.048425	-0.025175	-0.020269	0.023518	-0.036965

5 rows × 100 columns

from sklearn.metrics.pairwise import euclidean_distances

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