```
In [1]: from keras.preprocessing import text
    from keras.preprocessing import sequence
    from keras.utils import pad_sequences
    from keras.utils import to_categorical
    import numpy as np
    import pandas as pd
```

```
In [3]: 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
    vocab_size = len(word2id)
    embed_size = 100
    window_size = 2
    print('Vocabulary Size:', vocab_size)
    print('Vocabulary Sample:', list(word2id.items())[:10])
```

Vocabulary Size: 81

Vocabulary Sample: [('learning', 1), ('deep', 2), ('networks', 3), ('an d', 4), ('as', 5), ('of', 6), ('neural_', 7), ('→networks', 8), ('supervi sed', 9), ('_', 10)]

```
In [5]: 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</pre>
                                 and i != index])
                     label_word.append(word)
                     x = pad_sequences(context_words, maxlen=context_length)
                     y = to_categorical(label_word, vocab_size)
                     yield (x, y)
        i = 0
        for x, y in generate_context_word_pairs(corpus=wids, window_size=window_siz
            if 0 not in x[0]:
        # print('Context (X):', [id2word[w] for w in x[0]], '-> Target (Y):',id2wor
                if i == 10:
                     break
                i += 1
```

```
In [6]: 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
    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=
```

Model: "sequential"

Layer (type)	Output Shape	Param #						
embedding (Embedding)	(None, 4, 100)	8100						
lambda (Lambda)	(None, 100)	0						
dense (Dense)	(None, 81)	8181						
Total params: 16281 (63.60 KB) Trainable params: 16281 (63.60 KB) Non-trainable params: 0 (0.00 Byte)								

None

```
for epoch in range(1, 6):
In [10]:
              loss = 0.
              i = 0
              for x, y in generate_context_word_pairs(corpus=wids,window_size=window_
                  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: 451.10684871673584
          Epoch: 2
                          Loss: 447.54382729530334
          Epoch: 3
                          Loss: 445.5001447200775
          Epoch: 4
                           Loss: 443.5994474887848
          Epoch: 5
                           Loss: 441.92649722099304
In [11]:
         weights = cbow.get_weights()[0]
          weights = weights[1:]
          print(weights.shape)
          pd.DataFrame(weights, index=list(id2word.values())[1:]).head()
          (80, 100)
Out[11]:
                          0
                                   1
                                            2
                                                     3
                                                                       5
                                                                                 6
                             0.031762 0.032208
                                               0.005768  0.053792  0.011835  -0.031391
              deep
                    0.052731
                                                                                   0.0449
           networks
                    0.038828 -0.025172 -0.046274 -0.015622 -0.009437 -0.014444 -0.031176 -0.0058
               and -0.001400
                             0.003003
                                      0.041346
                                               0.005582 -0.020057
                                                                -0.040284 -0.014791 -0.0142
                as
                    0.023122
                            0.041794 -0.039984
                                               0.024638
                                                        0.037334
                                                                 0.041932 -0.038312 -0.0002
                of -0.023991 0.034275 0.003980
                                               0.014363 -0.029694
                                                                 0.042253
                                                                          0.024232 -0.0135
          5 rows × 100 columns
         from sklearn.metrics.pairwise import euclidean distances
          distance_matrix = euclidean_distances(weights)
          print(distance_matrix.shape)
          similar_words = {search_term: [id2word[idx] for idx in_distance_matrix[word
          for search term in ['deep']}
              similar_words
            Cell In[13], line 4
              similar_words = {search_term: [id2word[idx] for idx in_distance_matrix
          [word2id[search_term]-1].argsort()[1:6]+1]
                                                                       Λ
          SyntaxError: invalid character '_' (U+2423)
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In []: