## **Problem 3**

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```
In [9]: import numpy as np
    from P3CHelpers import *
    from keras.models import Sequential
    import sys
```

## 3D:

Fill in the generate\_traindata and find\_most\_similar\_pairs functions

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```
In [10]: def get_word_repr(word_to_index, word):
             Returns one-hot-encoded feature representation of the specified word given
             a dictionary mapping words to their one-hot-encoded index.
             Arguments:
                 word_to_index: Dictionary mapping words to their corresponding index
                                 in a one-hot-encoded representation of our corpus.
                                String containing word whose feature representation we wish
                 word:
         to compute.
             Returns:
                                              Feature representation of the passed-in word.
                feature_representation:
             unique_words = word_to_index.keys()
             # Return a vector that's zero everywhere besides the index corresponding to <w
         ord>
             feature representation = np.zeros(len(unique words))
             feature representation[word to index[word]] = 1
             return feature representation
         def generate_traindata(word_list, word_to_index, window_size=4):
             Generates training data for Skipgram model.
             Arguments:
                                Sequential list of words (strings).
                 word list:
                 word to index: Dictionary mapping words to their corresponding index
                                in a one-hot-encoded representation of our corpus.
                 window size:
                                Size of Skipgram window.
                                 (use the default value when running your code).
             Returns:
                 (trainX, trainY):
                                       A pair of matrices (trainX, trainY) containing train
         ing
                                       points (one-hot-encoded vectors representing individ
         ual words) and
                                        their corresponding labels (also one-hot-encoded vec
         tors representing words).
                                       For each index i, trainX[i] should correspond to a w
         ord in
                                        <word list>, and trainY[i] should correspond to one
         of the words within
                                       a window of size <window_size> of trainX[i].
             trainX = []
             trainY = []
             numWords = len(word_to_index)
             trainX = []
             trainY = []
             allZeroes = [0 for i in range(numWords)]
             for i in range(len(word_list)):
                 for j in range(-window_size, window_size + 1):
                     if i + j \ge 0 and i + j < len(word_list) and j != 0:
                         trainXVector = get_word_repr(word_to_index, word_list[i])
                         trainX.append(trainXVector)
                         trainYVector = get_word_repr(word_to_index, word_list[i+j])
                         trainY.append(trainYVector)
             return (np.array(trainX), np.array(trainY))
```

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```
In [12]: def find_most_similar_pairs(filename, num_latent_factors):
             Find the most similar pairs from the word embeddings computed from
             a body of text
             Arguments:
                                     Text file to read and train embeddings from
                 filename:
                 num_latent_factors: The number of latent factors / the size of the embeddi
         ng
             # Load in a list of words from the specified file; remove non-alphanumeric cha
         racters
             # and make all chars lowercase.
             sample_text = load_word_list(filename)
             # Create dictionary mapping unique words to their one-hot-encoded index
             word_to_index = generate_onehot_dict(sample_text)
             # Create training data using default window size
             trainX, trainY = generate_traindata(sample_text, word_to_index)
             # TODO: 1) Create and train model in Keras.
             # vocab size = number of unique words in our text file. Will be useful when ad
         ding layers
             # to your neural network
             vocab_size = len(word_to_index)
             model = Sequential()
             model.add(Dense(num latent factors, input dim=(vocab size)))
             model.add(Dense(vocab size))
             model.add(Activation('softmax'))
             model.compile(loss='categorical crossentropy', optimizer='rmsprop',
                           metrics=['accuracy'])
             fit = model.fit(trainX, trainY)
             print("Hidden layer shape" + str(model.layers[0].get_weights()[0].shape))
             print("Output layer shape" + str(model.layers[1].get_weights()[0].shape))
             # TODO: 2) Extract weights for hidden layer, set <weights> variable below
             weights = model.layers[0].get weights()[0]
             # Find and print most similar pairs
             similar pairs = most similar pairs(weights, word to index)
             for pair in similar pairs[:30]:
                 print(pair)
```

```
In [6]:

In [ ]:
```

## 3G:

Run the function below and report your results for dr\_seuss.txt.

```
In [ ]: find_most_similar_pairs('data/dr_seuss.txt', 10)
In [ ]:
```

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