```
In [1]: import tensorflow
 In [2]: import pandas as pd
          import numpy as np
          import collections
          import re
In [20]: f=open(r"C:\Users\Admin\Desktop\text.txt")
          document=f.read()
          f.close()
In [21]: document
Out[21]: 'If you want to associate environment variables with an environment, you
          can use the config API. '
In [25]: l_document = re.sub(r"[^a-zA-Z0-9]", " ", document.lower()).split()
In [26]: l document
Out[26]: ['if',
           'you',
           'want',
           'to',
           'associate',
           'environment',
           'variables',
           'with',
           'an',
           'environment',
           'you',
           'can',
           'use',
           'the',
           'config',
           'api']
In [27]: wordset=set(l_document)
In [28]: wordset
Out[28]: {'an',
           'api',
           'associate',
           'can',
           'config',
           'environment',
           'if',
           'the',
           'to',
           'use',
           'variables',
           'want',
           'with',
           'you'}
```

```
In [29]: def calculateBOW(wordset, l doc):
           tf diz = dict.fromkeys(wordset,0)
           for word in l doc:
               tf diz[word]=l doc.count(word)
           return tf diz
In [30]: bow1 = calculateBOW(wordset, l document)
         df bow = pd.DataFrame([bow1])
         df bow.head()
            if with can config to an use want variables you api the environment associate
Out[30]:
                                      1
                                                        2
                           1 1
                                 1
                                           1
                                                           1
                                                               1
In [31]: from sklearn.feature extraction.text import CountVectorizer
         vectorizer = CountVectorizer()
In [33]: X = vectorizer.fit transform([document])
         df bow sklearn = pd.DataFrame(X.toarray(),columns=vectorizer.get feature
         df bow sklearn.head()
            an api associate can config environment if the to use variables want with you
Out[33]:
                         1
                              1
                                    1
                                               2 1
                                                                               1
                                                                                   2
            - 1
                                                     1 1
                                                             1
In [34]: print(vectorizer.get_feature_names_out())
         ['an' 'api' 'associate' 'can' 'config' 'environment' 'if' 'the' 'to' 'use
           'variables' 'want' 'with' 'you']
In [37]: import nltk
         nltk.download('punkt')
         import re
         import numpy as np
         f= open(r"C:\Users\Admin\Desktop\text.txt")
         text=f.read()
         f.close()
         dataset = nltk.sent_tokenize(text)
         for i in range(len(dataset)):
             dataset[i] = dataset[i].lower()
             dataset[i] = re.sub(r'\W', ' ', dataset[i])
             dataset[i] = re.sub(r'\s+', ' ', dataset[i])
         [nltk data] Downloading package punkt to
         [nltk data]
                      C:\Users\Admin\AppData\Roaming\nltk_data...
         [nltk_data]
                       Package punkt is already up-to-date!
In [38]: print(dataset)
         ['if you want to associate environment variables with an environment you
         can use the config api ']
```

```
In [39]: word2count = {}
         for data in dataset:
              words = nltk.word_tokenize(data)
              for word in words:
                  if word not in word2count.keys():
                      word2count[word] = 1
                  else:
                      word2count[word] += 1
In [40]: word2count
Out[40]: {'if': 1,
           'you': 2,
           'want': 1,
           'to': 1,
           'associate': 1,
           'environment': 2,
           'variables': 1,
           'with': 1,
           'an': 1,
           'can': 1,
           'use': 1,
           'the': 1,
           'config': 1,
           'api': 1}
In [41]: vocab size = len(wordset)
          embed_dim = 10
          context_size = 4
In [42]: word_to_ix = {word: i for i, word in enumerate(wordset)}
          ix_to_word = {i: word for i, word in enumerate(wordset)}
In [43]: word_to_ix
Out[43]: {'if': 0,
           'with': 1,
           'can': 2,
           'config': 3,
           'to': 4,
           'an': 5,
           'use': 6,
           'want': 7,
           'variables': 8,
           'you': 9,
           'api': 10,
           'the': 11,
           'environment': 12,
           'associate': 13}
In [44]: ix to word
```

```
Out[44]: {0: 'if',
          1: 'with',
          2: 'can',
          3: 'config',
          4: 'to',
          5: 'an',
          6: 'use',
          7: 'want',
          8: 'variables',
          9: 'you',
          10: 'api',
          11: 'the',
          12: 'environment',
          13: 'associate'}
In [45]: data = []
         for i in range(2, len(words) - 2):
             context = [words[i - 2], words[i - 1], words[i + 1], words[i + 2]]
             target = words[i]
             data.append((context, target))
         print(data[:5])
         [(['if', 'you', 'to', 'associate'], 'want'), (['you', 'want', 'associate
          , 'environment'], 'to'), (['want', 'to', 'environment', 'variables'], 'a
         ssociate'), (['to', 'associate', 'variables', 'with'], 'environment'),
         (['associate', 'environment', 'with', 'an'], 'variables')]
In [46]: embeddings = np.random.random_sample((vocab_size, embed_dim))
In [47]: embeddings
Out[47]: array([[0.08267304, 0.06872026, 0.52314585, 0.85719784, 0.23772
                 0.21990182, 0.59516288, 0.71610117, 0.0135521, 0.32786658
                [0.87422636, 0.9516628, 0.58702059, 0.58114781, 0.06213979,
                 0.66809174, 0.40835397, 0.77462323, 0.08932546, 0.99915129],
                [0.92544957, 0.01498296, 0.66945095, 0.6775853, 0.32878235,
                 0.83584151, 0.71668359, 0.82645497, 0.24184188, 0.28974657],
                [0.93778769, 0.7786902, 0.64712119, 0.53514385, 0.1035955]
                 0.08273537, 0.31713938, 0.34834821, 0.34856258, 0.04982951],
                [0.99234292, 0.93822048, 0.28073592, 0.22362329, 0.96376676,
                 0.09481795, 0.7022882 , 0.34656426, 0.56779161, 0.50963159],
                [0.83133302, 0.63362814, 0.05718258, 0.42295399, 0.31136499,
                 0.75578362, 0.69763447, 0.74176033, 0.45508205, 0.40099483],
                [0.91259908, 0.8091109 , 0.71821172, 0.48200706, 0.71107511,
                 0.83958432, 0.18230773, 0.85929422, 0.88451157, 0.38777645],
                [0.60251991, 0.58005213, 0.51924271, 0.09766742, 0.09339876,
                 0.99921754, 0.13523909, 0.24603776, 0.04676803, 0.68127441],
                [0.48273284, 0.31194804, 0.33253094, 0.28536348, 0.95858413,
                 0.73051657, 0.12943171, 0.4085084 , 0.23834409, 0.18177285],
                [0.96775123, 0.07719052, 0.78566105, 0.68862173, 0.95716544,
                 0.25867813, 0.73095074, 0.91189725, 0.8519008 , 0.47168835],
                [0.15087434, 0.61855211, 0.9900737 , 0.21864929, 0.5205878
                 0.5260361 , 0.06047527, 0.02868031, 0.84643083, 0.9322089 ],
                [0.69422868, 0.75207946, 0.00739601, 0.0782964 , 0.39192214,
                 0.43202907, 0.93559772, 0.29486897, 0.60429403, 0.63154731],
                [0.27478837, 0.03938594, 0.83430168, 0.00952585, 0.76060092,
                 0.20769285, 0.77576561, 0.39348018, 0.52399601, 0.92591978],
                [0.88870875, 0.70144542, 0.79982161, 0.32783298, 0.72976906,
                 0.83529162, 0.56352137, 0.58888341, 0.35290347, 0.63938508]])
```

```
In [48]: def linear(m, theta):
             w = theta
             return m.dot(w)
In [49]: def log softmax(x):
             e_x = np.exp(x - np.max(x))
             return np.log(e x / e x.sum())
         def NLLLoss(logs, targets):
             out = logs[range(len(targets)), targets]
             return -out.sum()/len(out)
In [50]: import tensorflow as tf
         import keras.backend as k
         from keras.models import Sequential
         from keras.layers import Dense, Embedding, Lambda
In [51]: def log_softmax_crossentropy_with_logits(logits, target):
             out = np.zeros_like(logits)
             out[np.arange(len(logits)),target] = 1
             softmax = np.exp(logits) / np.exp(logits).sum(axis=-1,keepdims=True)
             return (- out + softmax) / logits.shape[0]
In [52]:
          def forward(context_idxs, theta):
             m = embeddings[context idxs].reshape(1, -1)
             n = linear(m, theta)
             o = log softmax(n)
             return m, n, o
In [53]: def backward(preds, theta, target_idxs):
             m, n, o = preds
             dlog = log softmax crossentropy with logits(n, target idxs)
             dw = m.T.dot(dlog)
             return dw
In [54]: def optimize(theta, grad, lr=0.03):
             theta -= grad * lr
             return theta
In [55]: theta = np.random.uniform(-1, 1, ( context_size * embed_dim, vocab_size))
```

```
In [56]: epoch_losses = {}
for epoch in range(80):
    losses = []
    for context, target in data:
        context_idxs = np.array([word_to_ix[w] for w in context])
        preds = forward(context_idxs, theta)

        target_idxs = np.array([word_to_ix[target]])
        loss = NLLLoss(preds[-1], target_idxs)

        losses.append(loss)

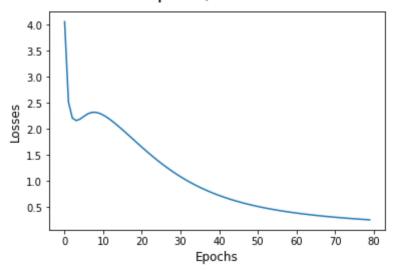
        grad = backward(preds, theta, target_idxs)
        theta = optimize(theta, grad, lr=0.03)
        epoch_losses[epoch] = losses
```

```
import matplotlib.pyplot as plt
ix = np.arange(0,80)

fig = plt.figure()
fig.suptitle('Epoch/Losses', fontsize=20)
plt.plot(ix,[epoch_losses[i][0] for i in ix])
plt.xlabel('Epochs', fontsize=12)
plt.ylabel('Losses', fontsize=12)
```

Out[57]: Text(0, 0.5, 'Losses')

Epoch/Losses



Out[59]: 'variables'