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In [1]: #Implement the Continuous Bag of Words (CBOW) Model. Stages can be:
# a. Data preparation
# b. Generate training data
# c. Train model
# d. Output
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import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib as mpl
import matplotlib.pylab as pylab
import numpy as np
%matplotlib inline
```

C:\Users\Suraj\anaconda3\lib\site-packages\scipy\\_\_init\_\_.py:146: UserWarning: A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy (detected version 1.26.1  
warnings.warn(f"A NumPy version >={np\_minversion} and <{np\_maxversion}")

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In [2]: #Data Preperation
import re
```

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In [3]: sentences = """We are about to study the idea of a computational process.
Computational processes are abstract beings that inhabit computers.
As they evolve, processes manipulate other abstract things called data.
The evolution of a process is directed by a pattern of rules
called a program. People create programs to direct processes. In effect,
we conjure the spirits of the computer with our spells."""
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In [4]: # remove special characters
sentences = re.sub('[^A-Za-z0-9]+', '', sentences)

# remove 1 letter words
sentences = re.sub(r'(?<^|)\w(?:$|)', '', sentences).strip()

# lower all characters
sentences = sentences.lower()
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In [5]: #Vocabulary
words = sentences.split()
vocab = set(words)
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In [6]: vocab_size = len(vocab)
embed_dim = 10
context_size = 2
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In [7]: #Implementation
word_to_ix = {word: i for i, word in enumerate(vocab)}
ix_to_word = {i: word for i, word in enumerate(vocab)}
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In [8]: #Data bag
# data - [(context), target]

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])
```

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[[('we', 'are', 'to', 'study'], 'about'), ('are', 'about', 'study', 'the'], 'to'), ('about', 'to', 'the', 'idea'], 'study'),
(['to', 'study', 'idea', 'of'], 'the'), ('study', 'the', 'of', 'computational'], 'idea')]
```

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In [9]: #embedding
embeddings = np.random.random_sample((vocab_size, embed_dim))

In [10]: #Linear Model
def linear(m, theta):
    w = theta
    return m.dot(w)

In [11]: #Log softmax + NLLloss = Cross Entropy
def log_softmax(x):
    e_x = np.exp(x - np.max(x))
    return np.log(e_x / e_x.sum())

In [12]: def NLLLoss(logs, targets):
    out = logs[range(len(targets)), targets]
    return -out.sum()/len(out)

In [13]: 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 [14]: #Forward Function
def forward(context_idx, theta):
    m = embeddings[context_idx].reshape(1, -1)
    n = linear(m, theta)
    o = log_softmax(n)

    return m, n, o

In [15]: #Backward function
def backward(preds, theta, target_idx):
    m, n, o = preds

    dlog = log_softmax_crossentropy_with_logits(n, target_idx)
    dw = m.T.dot(dlog)

    return dw

In [16]: #Optimize function
def optimize(theta, grad, lr=0.03):
    theta -= grad * lr
    return theta

In [17]: #Genrate training data

theta = np.random.uniform(-1, 1, (2 * context_size * embed_dim, vocab_size))

In [18]: epoch_losses = {}

for epoch in range(80):

    losses = []

    for context, target in data:
        context_idx = np.array([word_to_ix[w] for w in context])

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preds = forward(context_idxxs, theta)

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

losses.append(loss)

grad = backward(preds, theta, target_idxxs)
theta = optimize(theta, grad, lr=0.03)

epoch_losses[epoch] = losses

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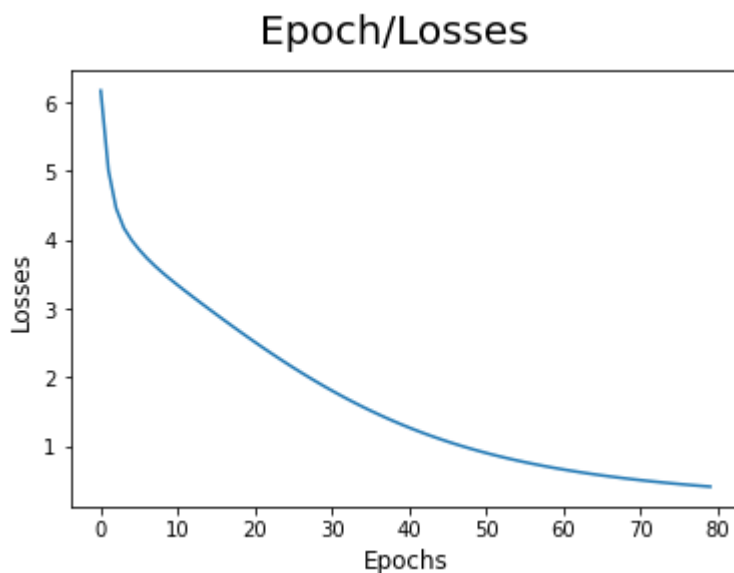
```

In [19]: #Analyze
#plot loss / epochs
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[19]: Text(0, 0.5, 'Losses')



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In [20]: #Predict Function
def predict(words):
    context_idxxs = np.array([word_to_ix[w] for w in words])
    preds = forward(context_idxxs, theta)
    word = ix_to_word[np.argmax(preds[-1])]

    return word

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In [21]: #(['we', 'are', 'to', 'study'], 'about')
predict(['we', 'are', 'to', 'study'])

```

Out[21]: 'about'

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In [22]: def accuracy():
    wrong = 0

    for context, target in data:
        if predict(context) != target:
            wrong += 1

```

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return (1 - (wrong / len(data)))
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In [23]: accuracy()

Out[23]: 1.0

In [24]: *#Output*  
predict(['processes', 'manipulate', 'things', 'study'])

Out[24]: 'they'

In [ ]: