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In[]: ## Name : Tanmay Gujar

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```

```
In[1]: import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib as mpl
import matplotlib.pyplot as pylab
import numpy as np
%matplotlib inline
```

```
In[2]: #DataPreparation
import re
```

```
In[3]: sentences="""We are about to study the idea of a computational process. Computation
al 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."""
```

Clean Data

```
In[4]: #removespecialcharacters
sentences=re.sub('[^A-Za-z0-9]+',' ',sentences)

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

#lower all characters
sentences=sentences.lower()
```

Vocabulary

```
In[5]: words=sentences.split() vocab=set(words)
```

```
In[6]: vocab_size=len(vocab)
vocab_dim=10
context_size=2
```

Implementation

```
In[7]: word_to_ix={word:i for i,word in enumerate(vocab)} ix_to_word={i:word for i,word in enumerate(vocab)}
```

Data bags

```
In[8]: #data-[(context),target]

data=[]
for i in range(2, len(words)-2):
    [words[i-2], words[i-1], words[i+1], words[i+2]]
```

```
target=words[i]
data.append((context,target))print(data[:5])
```

```
[(['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')]
```

Embeddings

In[9]:

```
embeddings=np.random.random_sample((vocab_size,embed_dim))
```

Linear Model

In[10]:

```
def linear(m,theta):w
    =theta
    return m.dot(w)
```

Log softmax + NLLloss = Cross Entropy

In[11]:

```
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]
```

Forward function

In[14]:

```
def forward(context_idxs,theta):
    m=embeddings[context_idxs].reshape(1,-1)
    n=linear(m,theta)
    o=log_softmax(n)

    return m,n,o
```

Backward function

In[15]:

```
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
```

Optimize function

```
In[16]:  
  
def optimize(theta, grad, lr=0.03):  
    eta -= grad * lr  
    return theta
```

Training

```
In[17]:  
  
# Generate training data  
  
theta = np.random.uniform(-1, 1, (2 * context_size * embed_dim, vocab_size))
```

```
In[18]:  
  
epoch_losses = {}  
  
for epoch in range(80):  
  
    sses = []  
  
    for context, target in data:  
        context_idx = np.array([word_to_ix[w] for w in context])  
        preds = forward(context_idx, theta)  
  
        target_idx = np.array([word_to_ix[target]])  
        loss = NLLLoss(preds[-1], target_idx)  
  
        losses.append(loss)  
  
        grad = backward(preds, theta, target_idx)  
        theta = optimize(theta, grad, lr=0.03)  
  
    epoch_losses[epoch] = losses
```

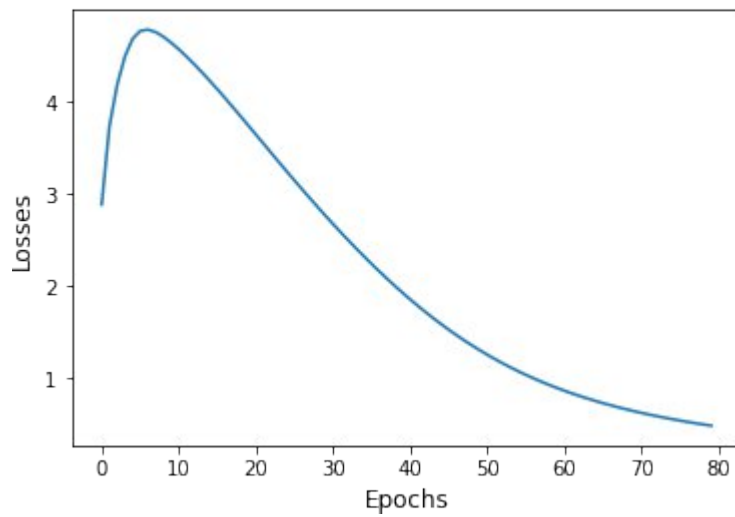
Analyze

Plot loss/epoch

```
In[19]:  
  
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')
```

Epoch/Losses



Predict function

In[20]:

```
def predict(words):  
    context_idx = np.array([word_to_ix[w] for w in words])  
    preds = forward(context_idx, theta)  
    word = ix_to_word[np.argmax(preds[-1])]  
  
    return word
```

In[21]:

```
#(['we', 'are', 'to', 'study'], 'about')  
predict(['we', 'are', 'to', 'study'])
```

Out[21]: 'about'

Accuracy

In[22]:

```
def accuracy():  
    wrong = 0  
  
    for context, target in data:  
        if predict(context) != target:  
            wrong += 1  
  
    return (1 - (wrong / len(data)))
```

In[23]:

```
accuracy()
```

Out[23]: 1.0

In[24]:

```
predict(['processes', 'manipulate', 'things', 'study'])
```

Out[24]: 'effect'

In[]: