

```
In [ ]: ## Name : Satyam kumar  
## Roll no: I4250  
## Subject: LP-IV(DL)
```

```
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]: #Data Prepration  
import re
```

```
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."""
```

Clean Data

```
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()
```

Vocabulary

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

```
In [6]: vocab_size = len(vocab)  
embed_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):  
        theta -= 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):  
  
            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
```

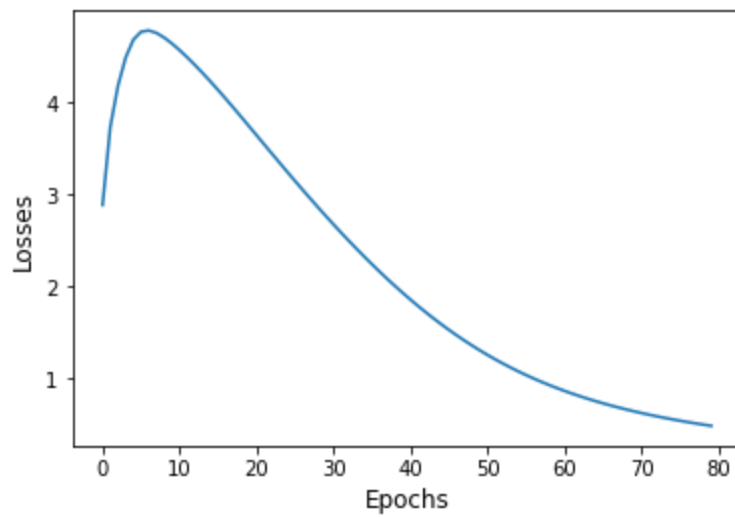
Analyze

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 [ ]:
```