In [1]:

```
import matplotlib.pyplot as plt
import seaborn as sns
import matplotlib as mpl
import matplotlib.pylab 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. Computatio
evolve, processes manipulate other abstract things called data. The evolution of a p
create programs to direct processes. In effect, we conjure the spirits of the comput
```

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 = []
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])
```

```
[(['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]:

```
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

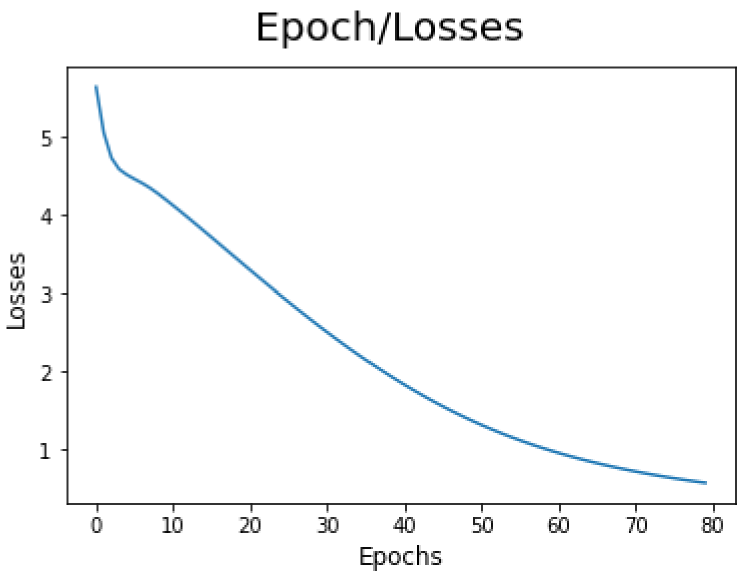
Plot loss/epoch

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')
```



Predict function

In [20]:

```
def predict(words):
    context_idxs = np.array([word_to_ix[w] for w in words])
    preds = forward(context_idxs, 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]:

```
'other'
```

In [ ]:

```

```