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.
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):
    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', 'ide
```

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]: #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_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
```

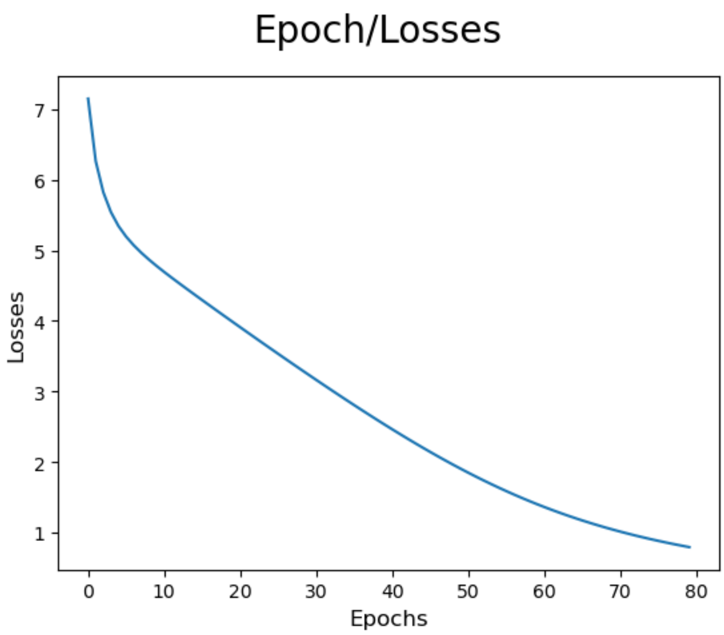
Analyze

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'`