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In [1]:
         import matplotlib.pyplot as plt
         import seaborn as sns
         import matplotlib as mpl
         import matplotlib.pylab as pylab
         import numpy as np
 In [2]: 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."""
 In [4]: | sentences = re.sub('[^A-Za-z0-9]+', ' ', sentences)
 In [5]: | sentences = re.sub(r'(?:^| )\w(?:$| )', ' ', sentences).strip()
 In [6]: | sentences = sentences.lower()
 In [7]: |words = sentences.split()
         vocab = set(words)
 In [8]: vocab size = len(vocab)
         embed dim = 10
         context size = 2
 In [9]: | word to ix = {word: i for i, word in enumerate(vocab)}
         ix to word = {i: word for i, word in enumerate(vocab)}
In [10]: 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'], 't
         o'), (['about', 'to', 'the', 'idea'], 'study'), (['to', 'study', 'idea', 'of'],
         'the'), (['study', 'the', 'of', 'computational'], 'idea')]
In [11]: embeddings = np.random.random_sample((vocab_size, embed_dim))
In [12]: def linear(m, theta):
             w = theta
             return m.dot(w)
```

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In [13]: def log_softmax(x):
             e_x = np.exp(x - np.max(x))
             return np.log(e_x / e_x.sum())
In [14]: def NLLLoss(logs, targets):
             out = logs[range(len(targets)), targets]
             return -out.sum()/len(out)
In [15]: 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 [16]: def forward(context_idxs, theta):
             m = embeddings[context_idxs].reshape(1, -1)
             n = linear(m, theta)
             o = log_softmax(n)
             return m, n, o
In [17]: 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
In [18]: def optimize(theta, grad, lr=0.03):
             theta -= grad * lr
             return theta
In [19]: | theta = np.random.uniform(-1, 1, (2 * context_size * embed_dim, vocab_size))
```

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In [20]: 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)

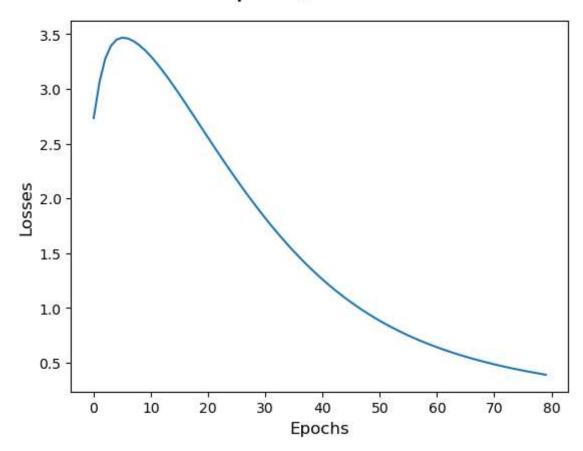
        epoch_losses[epoch] = losses
```

```
In [21]: 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[21]: Text(0, 0.5, 'Losses')

## Epoch/Losses



```
In [23]: predict(['we', 'are', 'to', 'study'])
```

Out[23]: 'about'