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In [ ]: 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 [ ]: #Data Prepration
        import re
In [ ]: 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 [ ]: # 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()
In [ ]: #Vocabulary
        words = sentences.split()
        vocab = set(words)
In [ ]: |vocab size = len(vocab)
        embed dim = 10
        context size = 2
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In [ ]: #Implementation
        word_to_ix = {word: i for i, word in enumerate(vocab)}
        ix_to_word = {i: word for i, word in enumerate(vocab)}
In [ ]: #Data bag
        # 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])
In [ ]: #embedding
        embeddings = np.random.random_sample((vocab_size, embed_dim))
In [ ]: #Linear Model
        def linear(m, theta):
            w = theta
            return m.dot(w)
In [ ]: #Log softmax + NLLLoss = Cross Entropy
        def log softmax(x):
            e x = np.exp(x - np.max(x))
            return np.log(e x / e x.sum())
In [ ]: def NLLLoss(logs, targets):
            out = logs[range(len(targets)), targets]
            return -out.sum()/len(out)
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In [ ]: 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 [ ]: #Forward Function
        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 [ ]: #Backward function
        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 [ ]: #Optimize function
        def optimize(theta, grad, lr=0.03):
            theta -= grad * lr
            return theta
In [ ]: #Genrate training data
        theta = np.random.uniform(-1, 1, (2 * context_size * embed_dim, vocab_size))
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In [ ]: 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 [ ]: #Analyze
        #plot loss / epochs
        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)
In [ ]: #Predict Function
        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
```