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def lossFun(inputs, targets, hprev):
 inputs, targets are both list of integers.
 hprev is Hx1 array of initial hidden state
 returns the loss, gradients on model parameters, and last hidden state
 xs, hs, ys, ps = \{\}, \{\}, \{\}, \{\}
 hs[-1] = np.copy(hprev)
 loss = 0
# forward pass
 for t in xrange(len(inputs)):
   xs[t] = np.zeros((vocab\_size,1)) # encode in 1-of-k representation
   xs[t][inputs[t]] = 1
  hs[t] = np.tanh(np.dot(Wxh, xs[t]) + np.dot(Whh, hs[t-1]) + bh) # hidden state
  ys[t] = np.dot(Why, hs[t]) + by \# unnormalized log probabilities for next chars
  ps[t] = np.exp(ys[t]) / np.sum(np.exp(ys[t])) # probabilities for next chars
   loss += -np.log(ps[t][targets[t],0]) # softmax (cross-entropy loss)
# backward pass: compute gradients going backwards
dWxh, dWhh, dWhy = np.zeros_like(Wxh), np.zeros_like(Whh), np.zeros_like(Why)
 dbh, dby = np.zeros_like(bh), np.zeros_like(by)
 dhnext = np. zeros_like (hs[0])
 for t in reversed (xrange (len (inputs))):
  dy = np.copy(ps[t])
  dy[targets[t]] -= 1 # backprop into y
  dWhy += np. dot(dy, hs[t].T)
  dby += dy
  dh = np.dot(Why.T, dy) + dhnext # backprop into h
  dhraw = (1 - hs[t] * hs[t]) * dh # backprop through tanh nonlinearity
  dbh += dhraw
  dWxh += np. dot(dhraw, xs[t].T)
  dWhh += np. dot (dhraw, hs[t-1].T)
  dhnext = np.dot(Whh.T, dhraw)
 for dparam in [dWxh, dWhh, dWhy, dbh, dby]:
  np. clip (dparam, -5, 5, out=dparam) # clip to mitigate exploding gradients
 return loss, dWxh, dWhh, dWhy, dbh, dby, hs[len(inputs)-1]
```