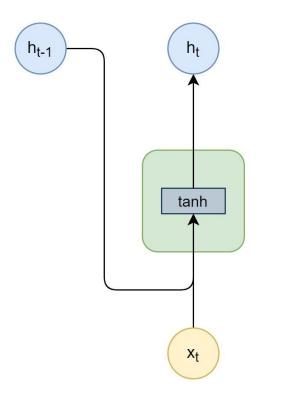
HW3P1 Bootcamp

RNN, GRU, CTC, and Greedy/Beam Search (Fall 2022)

Aparajith Srinivasan

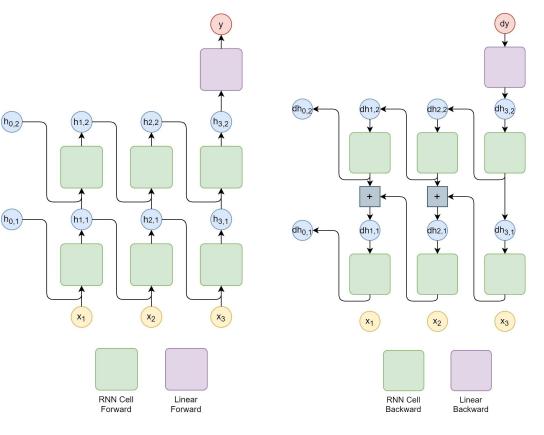
RNN Cell Forward / Backward



$$h_t = tanh(W_{ih}x_t + b_{ih} + W_{hh}h_{t-1} + b_{hh})$$

Tip: Very similar to how you did *linear.py* in hw1p1.

RNN Phoneme Classifier

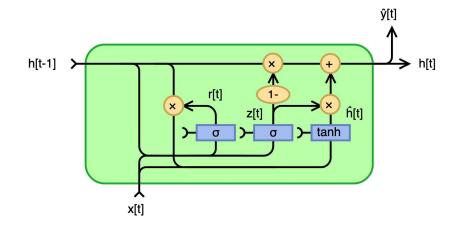


- Forward is straight forward (XD)
- Backward is tricky

.

- 2 diagrams in the write up for understanding the data flow
- Then follow the pseudocode exactly

GRU Cell Forward/Backward



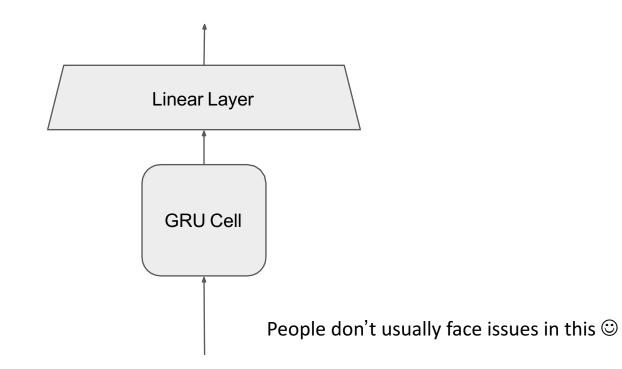
$$\begin{split} \mathbf{r}_t &= \sigma(\mathbf{W}_{ir}\mathbf{x}_t + \mathbf{b}_{ir} + \mathbf{W}_{hr}\mathbf{h}_{t-1} + \mathbf{b}_{hr}) \\ \mathbf{z}_t &= \sigma(\mathbf{W}_{iz}\mathbf{x}_t + \mathbf{b}_{iz} + \mathbf{W}_{hz}\mathbf{h}_{t-1} + \mathbf{b}_{hz}) \\ \mathbf{n}_t &= \tanh(\mathbf{W}_{in}\mathbf{x}_t + \mathbf{b}_{in} + \mathbf{r}_t \otimes (\mathbf{W}_{hn}\mathbf{h}_{t-1} + \mathbf{b}_{hn})) \\ \mathbf{h}_t &= (1 - \mathbf{z}_t) \otimes \mathbf{n}_t + \mathbf{z}_t \otimes \mathbf{h}_{t-1} \end{split}$$

https://colah.github.io/posts/2015-08-Backprop

GRU Cell Forward / Backward

- GRU backward be the longest question in HW3P1
- Tips:
 - Modify the *test_gru.py* code accordingly all dWs and dbs should correct to make sure that your dx and dh are correct
 - Can try to decompose eqns in forward (That's how I did :'))
 - $A = Tanh(Wx^*x + bx + Wh^*h + bh)$
 - $Z1 = Wx^*x + bx$
 - Z2 = Wh*h + bh
 - Z = Z1 + Z2
 - A = Tanh(Z)
 - Backward is easy now. Need to compute the gradients in this order.
 Given dA (actually dLdA ignoring for simplicity)
 - dZ -> dZ1, dZ2 -> dWh, dh, dbh -> ...

GRU Inference



CTC based questions

Lecture slides have everything needed to complete all the CTC sections and also decoding

CTC based questions

• We have given example questions for you to understand the math behind it

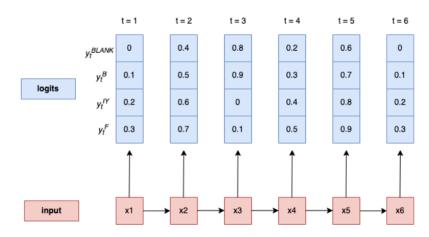
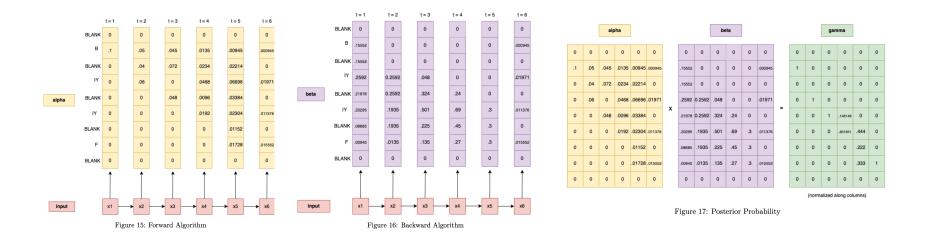


Figure 12: An overall CTC setup example

CTC based questions

• We have given example questions for you to understand the math behind it



Greedy Search

• Taking the most probably output at each time step

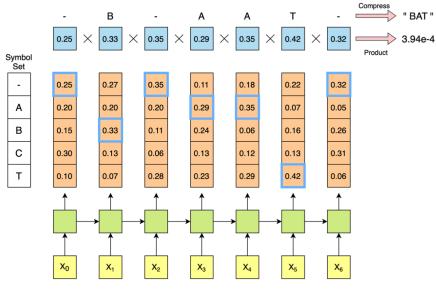


Figure 18: Greedy Search

Greedy Search

Taking the most probably output at each time step ٠

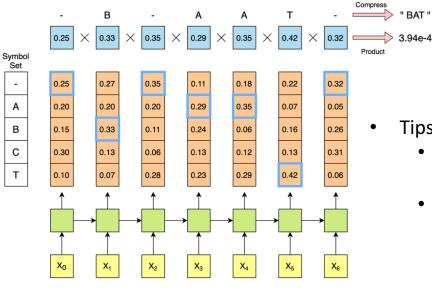
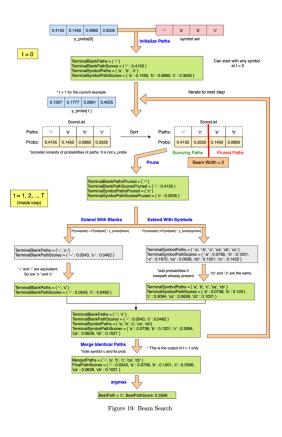


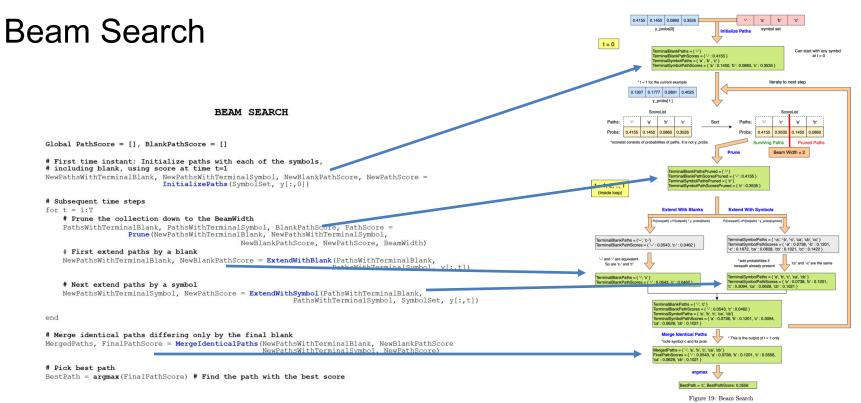
Figure 18: Greedy Search

- Tips
 - Write your compress function separately
 - Can complete without a for loop but a ٠ for loop wont cause autolab to time out \odot

Beam Search

- Another hard question in this part
- Tips to complete this question fast
 - Understand beam search from the lecture videos and slides
 - Beware of the definition of *set()* (python { }) and *list()* (python []) from the code given in lecture slides. There is a difference in the python implementation
 - Complete each function *InitializePaths, Prune, ExtendWithBlank, ExtendWithSymbol, MergeIdenticalPaths* individually and then check your outputs with the flow chart given in the write up





- Green boxes show the output for the 1st test case in the local autograder for just 1 time step
- You can break the flow in-between and check your answers

Thank you! Q & A