

This handout includes space for every question that requires a written response. Please feel free to use it to handwrite your solutions (legibly, please). If you choose to typeset your solutions, the `README.md` for this assignment includes instructions to regenerate this handout with your typeset L^AT_EX solutions.

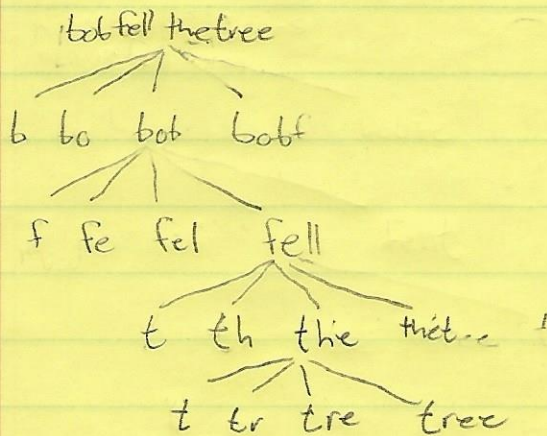
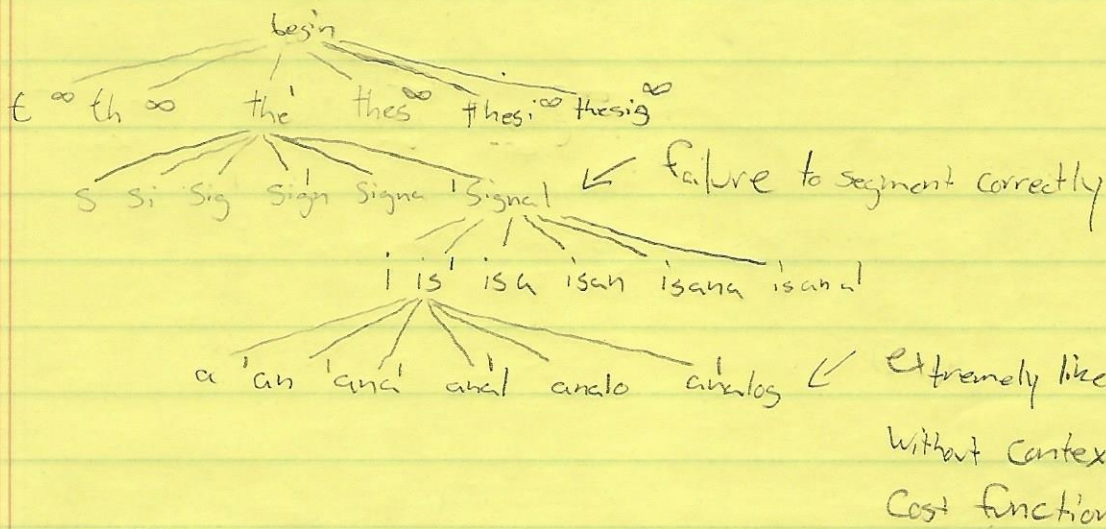
1.a

See image next page

1(a) Word Segmentation

Unigram model with a binary cost function $c(w_i) = \begin{cases} w_i \text{ in dictionary} = 0 \\ w_i \text{ not in dictionary} = \infty \end{cases}$

Query = "thesignalanalog"



= bob fell the tree

2.a

A good example sentence where a greedy vowel insertion implementation would be challenging is in the sentence "the mood is bright" = "th md s brght".

A reasonable set of possibleFills here are [(the),(mad,mood,mud,meed,mood),(as,is,us),(bright, brought)]

The real challenge here is selecting mood with a greedy implementation. The bigram with "the" would likely be somewhat equally weighted across the 5 options listed.

3.a

State space

(Location in the string we are presently looking at, Prior word selected)

Actions

Move from the previously selected word to lowest cost possible fill of the shortest segment of the string assuming that it is the only option

Costs

Bigram correlation score

End test

Reaching the end of the string

3.c

State space

Location in the string we are presently looking at

Actions

Move from the previously selected word to lowest cost possible fill of the shortest segment of the string assuming that it is the only option

Cost

Bigram correlation score plus the difference between the heuristic of the end state and initial state

End state

Reached the end of the string

 $h(s)$

A singular weight vector that matches the dimensionality of the embedded version of the current state

For the Heuristic to work $h(s)$ must not exceed the total cost of the inserted and segmented string to ensure we don't go negative for UCS. To achieve this the heuristic of the desired end state needs to be 0. This is achieved during training of the weight vector.

4.a

4.b

4.c