## CS B551, Fall 2016, In-class individual activity #2

Write your name and IU user id:

A certain little-known human language consists of only nouns and verbs. A sentence begins with a noun with probability 75% and with a verb with probability 25%. If a given word in a sentence is a noun, there is a 75% probability that the next word is a verb and a 25% probability that the next word is a noun. If a given word is a verb, there is a 75% probability that the next word in the sentence is a noun, and a 25% probability that is a verb.
We can model this language using a Markov chain with two states: noun and verb.
1. Draw the Markov chain described above.

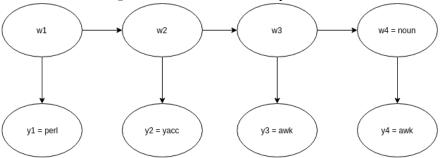
2. What is the probability that the first three words of a sentence are all nouns?

$$(0.75)(0.25)(0.25) \approx 0.0469$$

3. What is the probability that the third word of a sentence is a noun? For the third word to be a noun, the first three words must be NNN, NVN, VNN, or VVN. Thus the probability that the third word is a noun is:

$$(0.75)(0.25)(0.25) + (0.75)(0.75)(0.75) + (0.25)(0.75)(0.25) + (0.25)(0.25)(0.75) = .5625$$

- 4. There are only four words in the language: awk, yacc, grep, and perl. Each of these words can be either a noun or a verb. Of all noun occurrences, 10% are the word awk, 20% are the word yacc, 40% are the word grep, and 30% are the word perl. Of all verb occurrences, 20% are the word awk, 30% are the word yacc, 45% are the word grep, and 5% are the word perl.
  - (a) Using this model and the Variable Elimination algorithm, find the probability that the fourth word of the following sentence is a noun: Perl yacc awk awk.



We'd like to calculate  $p(s_4 = noun|y_1 = perl, y_2 = yacc, y_3 = awk, y_4 = awk)$ , which is equal to:

$$\frac{p(s_4 = noun, y_1 = perl, y_2 = yacc, y_3 = awk, y_4 = awk)}{p(y_1 = perl, y_2 = yacc, y_3 = awk, y_4 = awk)}$$

First we compute the numerator using Variable elimination. Let's use an elimination ordering of  $s_1$ ,  $s_2$ ,  $s_3$ . The initial set of factors F involved in our problem is:  $\{P(s_4 = noun|s_3), P(s_3|s_2), P(s_2|s_1), P(s_1), P(y_4 = awk|s_4 = noun), P(y_3 = awk|s_3), P(y_2 = yacc|s_2), P(y_1 = perl|s_1)\}.$ 

First we eliminate  $s_1$ . The subset of factors in F involving  $s_1$  is  $\{P(s_2|s_1), P(s_1), P(y_1 = perl|s_1)\}$ . We want to sum over all possible values of  $s_1$  of the product of these three factors to eliminate  $s_1$ ; the resulting expression will still involve  $s_2$ , so we'll get a new factor  $\tau_1(s_2)$ :

$$\tau_{1}(s_{2}) = \sum_{s_{1} \in \{noun, verb\}} P(s_{2}|s_{1})P(s_{1})P(y_{1} = perl|s_{1})$$

$$= P(s_{2}|s_{1} = noun)P(s_{1} = noun)P(y_{1} = perl|s_{1} = noun)$$

$$+P(s_{2}|s_{1} = verb)P(s_{1} = verb)P(y_{1} = perl|s_{1} = verb)$$

Plugging in numbers, we get:

$$\tau_1(s_2 = noun) = (0.25)(0.75)(0.3) + (0.75)(0.25)(0.05) \approx 0.065625$$

$$\tau_1(s_2 = verb) = (0.75)(0.75)(0.3) + (0.25)(0.25)(0.05) \approx 0.171875$$

Our new set F of factors is now  $\{P(y_4 = awk | s_4 = noun), P(s_4 = noun | s_3), P(s_3 | s_2), P(y_3 = awk | s_3), P(y_2 = yacc | s_2), \tau_1(s_2)\}$ . We've eliminated  $s_1$ !

Now we eliminate  $s_2$ :

$$\tau_2(s_3) = \sum_{s_2 \in \{noun, verb\}} P(s_3|s_2)\tau(s_2)P(y_2 = yacc|s_2)$$

so:

$$\tau_2(s_3 = noun) = (0.25)(0.065625)(0.2) + (0.75)(0.171875)(0.3) \approx 0.041953$$

$$\tau_2(s_3 = verb) = (0.75)(0.065625)(0.2) + (0.25)(0.171875)(0.3) \approx 0.022734$$

Our new set F of factors is now  $\{P(y_4 = awk|s_4 = noun), P(s_4 = noun|s_3), P(y_3 = awk|s_3), \tau_2(s_3)\}$ . We've eliminated  $s_2$ !

Now eliminate  $s_3$ :

$$\tau_2(s_4) = \sum_{s_3 \in \{noun, verb\}} P(y_4 = awk | s_4 = noun) P(s_4 = noun | s_3) \tau(s_3) P(y_3 = awk | s_3)$$

so:

$$\tau_3(s_4 = noun) = (0.1)\{(0.25)(0.041953)(0.1) + (0.75)(0.022734)(0.2)\} \approx 0.0004458925$$

$$\tau_3(s_4 = verb) = (0.2)\{(0.75)(0.041953)(0.1) + (0.25)(0.022734)(0.2)\} \approx 0.000856635$$

Now  $\tau_3$  is the numerator we wanted to calculate  $(p(s_4, y_1 = perl, y_2 = yacc, y_3 = awk, y_4 = awk))$ . To compute the denominator, we can just marginalize out over the two possible values of  $s_4$  (noun or verb):

$$p(s_4 = noun|y_1 = perl, y_2 = yacc, y_3 = awk, y_4 = awk) = \frac{0.0004458925}{0.0004458925 + 0.000856635} \approx 0.3423286$$

(b) Using this model and the Viterbi algorithm, find the most likely part-of-speech state sequence for the following sentence: Perl yacc awk awk.

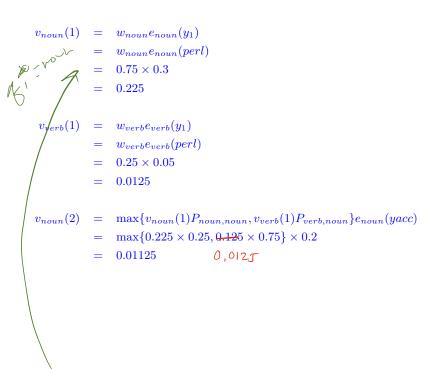
To simplify the notation, let  $p_{ij}$  be the transition probabilities,  $e_i(y)$  be the emission probability of outputing y given state i, and  $w_i$  be the initial probability for state i. As we saw in class, we'll define  $v_i(t)$  to be the probability of the most possible path ending in state i at position t, i.e.:

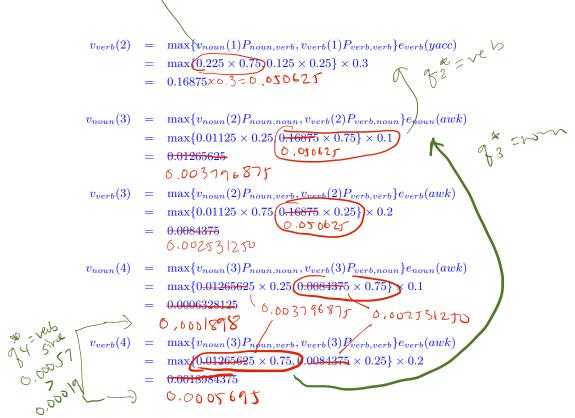
$$v_i(t) = \max_{q_1, q_2, q_{t-1}} = P(q_1, q_2, \dots, q_{t-1}, q_t = i | y_1, y_2, \dots, y_t)$$

As we saw, Viterbi can be implemented via dynamic programming. In other words, we can calculate  $v_i(t)$  in an iterative way:

for 
$$t = 1$$
:  $v_j(1) = w_j e_j(y_t)$   
for  $t > 1$ :  $v_j(t) = \max_{i \in \{n, v\}} [v_i(t-1)p_{ij}]e_j(y_t)$ 

After finishing calculating all the  $v_j(t)$ , then the state sequence that outputs the largest probability is what we want. Specifically,





Now we know that the optimal state sequence ends with verb, i.e.  $q_4^* = verb$ , since  $v_{verb}(4) > v_{noun}(4)$ .  $\checkmark$  We now "trace backwards" from there. In the maximization inside  $v_{verb}(4)$ , we chose the first of the two possibilities (that the prior state was a noun), so  $q_3^* = noun$ . In the maximization where we computed  $v_{noun}(3)$ , we chose the second possibility (that prior state was a verb), so  $q_2^* = verb$ , and so on. So the best state sequence is noun, verb, noun verb.