# CS6320, Fall 2021 Dr. Mithun Balakrishna Homework 3 Due Friday, October 15<sup>th</sup>, 2021 11:59pm

#### **A. Submission Instructions:**

- Submit your solutions via eLearning.
- Please submit a single zip file containing **ALL** the relevant homework solution files. The zip filename should follow the pattern "HW#\_FirstnameLastname.zip" (Example: HW3\_Claire Underwood.zip)
  - o **Penalty of 5 points** if not followed
- For all non-programming questions:
  - o Please include **ALL** the solutions in a **single** PDF/Doc/PS/Image file
  - The filename should follow the pattern "HW#\_FirstnameLastname.FileExtension" (Example: HW3\_Claire Underwood.pdf)
  - o **Penalty of 5 points** if not followed
- For programming questions:
  - Write the programming solutions in C/C++, Java, or Python. For using any other programming language, please get prior approval from the TA.
  - Include a Readme file with instructions on how to build and run your programming question solution
    - Instructions should be very simple:
      - python bigram.py input\_arguments

OR

python bigram.py (if the input arguments are hard coded)

- Hard coding the input arguments to your program is fine unless the TA cannot run your code directly. Do NOT include instructions such as: "Please modify the path in my main function. Then copy the training data in the same folder."
- Provide your training data together unless the dataset is too large.
- Penalty of 10 points if not followed
- Submit ALL your source code files
  - Do not write your solutions in the readme file
  - Penalty of 10 points if not followed
- Late Submission Penalty:
  - o up to 2 hours late 10% deduction
  - o 2 4 hours late 20% deduction
  - o 4 12 hours late 35% deduction
  - o 12 24 hours late 50% deduction
  - o 24 48 hours late 75% deduction
  - o more than 48 hours late 100% deduction (zero credit)

## **B. Problems:**

## 1. POS Tagging (60 points)

For this question, you have been given a POS-tagged training file, NLP6320\_POSTaggedTrainingSet.txt (provided as Addendum to this homework on eLearning), that has been tagged with POS tags from the Penn Treebank POS tagset (Figure 1).

Tag	Description	Example	Tag	Description	Example
CC	coordin. conjunction	and, but, or	SYM	symbol	+,%, &
CD	cardinal number	one, two, three	TO	"to"	to
DT	determiner	a, the	UH	interjection	ah, oops
EX	existential 'there'	there	VB	verb, base form	eat
FW	foreign word	mea culpa	VBD	verb, past tense	ate
IN	preposition/sub-conj	of, in, by	VBG	verb, gerund	eating
JJ	adjective	yellow	VBN	verb, past participle	eaten
JJR	adj., comparative	bigger	VBP	verb, non-3sg pres	eat
JJS	adj., superlative	wildest	VBZ	verb, 3sg pres	eats
LS	list item marker	1, 2, One	WDT	wh-determiner	which, that
MD	modal	can, should	WP	wh-pronoun	what, who
NN	noun, sing. or mass	llama	WP\$	possessive wh-	whose
NNS	noun, plural	llamas	WRB	wh-adverb	how, where
NNP	proper noun, singular	IBM	\$	dollar sign	\$
NNPS	proper noun, plural	Carolinas	#	pound sign	#
PDT	predeterminer	all, both	66	left quote	' or "
POS	possessive ending	's	"	right quote	or "
PRP	personal pronoun	I, you, he	(	left parenthesis	[, (, {, <
PRP\$	possessive pronoun	your, one's	)	right parenthesis	], ), }, >
RB	adverb	quickly, never	,	comma	,
RBR	adverb, comparative	faster		sentence-final punc	.!?
RBS	adverb, superlative	fastest	:	mid-sentence punc	:;
RP	particle	up, off		, 578 	

Figure 1. Penn Treebank POS tagset

#### Other Instructions:

- **1.** Use each line (ending with newline character) in the corpus as a single text sentence.
- **2.** Use whitespace (i.e. space, tab, and newline) to tokenize each text sentence into words/tokens.

- 3. Use the WORD\_POS pattern to extract the actual word and part-of-speech tag (i.e. the WORD part in the WORD\_POS pattern is the actual word and POS part in the WORD\_POS pattern is the part-of-speech tag) from the tokenized word.
  - For example, in the tokenized word "Brainpower\_NNP", "Brainpower" is the actual word and "NNP" is the part-of-speech tag.
- **4.** Do **NOT** perform any type of word/token normalization (i.e. case-normalization, stemming, lemmatization, etc.). Proper nouns etc
- **5.** Creation and matching of words and part-of-speech tags should be exact and case-sensitive.
- **6.** Smoothing is NOT required.
- 7. Please consider special tag "<s>" at the start of a text sentence and "</s>" at the end of a text sentence.

  Position of the word also matters in POS

Use the POS tagged file to perform:

- **a. Transformation-based POS Tagging**: Implement Brill's transformation-based POS tagging algorithm using the following template ONLY:
  - Change tag a to tag b when the previous word is tagged z Slide 23 & 24 in POS
  - 1. Write a software program to learn rules using the above specified template.
  - 2. Write a computer program to compute the POS tags for any input sentence using the learnt rules. Your program should assign a POS tag for every whitespace separated word in an input sentence.

### b. Naïve Bayesian Classification (Bigram) based POS Tagging:

$$\hat{t}_1^n = \underset{t_1^n}{\operatorname{argmax}} P(t_1^n | w_1^n) \approx \underset{t_1^n}{\operatorname{argmax}} \prod_{i=1}^n P(w_i | t_i) P(t_i | t_{i-1})$$

- Using the given corpus, write a computer program to compute the bigram models (counts and probabilities) required by the above Naïve Bayesian Classification formula. Please do not submit the bigram models with the homework solution submission. The TA can run your program to produce and check them.
- Using the created bigram models, write a computer program to compute the POS tags for any input sentence using the above Naïve Bayesian Classification formula.
   Your program should assign a POS tag for every space separated word in an input sentence.
- Following the instructions to consider a special tag "<s>" at the start of a text sentence and "</s>" at the end of a text sentence, the formula to compute the POS tag sequence for the input sentence "John went to work." will be:

$$\begin{split} \hat{t}_{1}^{5} &= \operatorname*{argmax}_{\hat{t}_{1}^{5}} P(John|tag_{1}) * P(tag_{1}| < s >) * P(went|tag_{2}) * P(tag_{2}|tag_{1}) \\ &* P(to|tag_{3}) * P(tag_{3}|tag_{2}) * P(work|tag_{4}) * P(tag_{4}|tag_{3}) \\ &* P(.|tag_{5}) * P(tag_{5}|tag_{4}) * P( |tag_{5}) \end{split}$$

# 2. HMM Decoding: Viterbi Algorithm (40 points):

		NNP	MD	VB	JJ	NN	RB	DT
$\Pi \rightarrow$	< <i>s</i> >	0.2767	0.0006	0.0031	0.0453	0.0449	0.0510	0.2026
	NNP	0.3777	0.0110	0.0009	0.0084	0.0584	0.0090	0.0025
	MD	0.0008	0.0002	0.7968	0.0005	0.0008	0.1698	0.0041
	VB	0.0322	0.0005	0.0050	0.0837	0.0615	0.0514	0.2231
	JJ	0.0366	0.0004	0.0001	0.0733	0.4509	0.0036	0.0036
	NN	0.0096	0.0176	0.0014	0.0086	0.1216	0.0177	0.0068
	RB	0.0068	0.0102	0.1011	0.1012	0.0120	0.0728	0.0479
	DT	0.1147	0.0021	0.0002	0.2157	0.4744	0.0102	0.0017

**Table 1. HMM Transition Probability** 

	Janet	will	back	the	bill
NNP	0.000032	0	0	0.000048	0
MD	0	0.308431	0	0	0
VB	0	0.000028	0.000672	0	0.000028
JJ	0	0	0.000340	0	0
NN	0	0.000200	0.000223	0	0.002337
RB	0	0	0.010446	0	0
DT	0	0	0	0.506099	0

**Table 2. HMM Observation Likelihood** 

For the HMM shown above, please perform the following:

**Programmatically** implement the Viterbi algorithm to compute the most likely tag sequence and its probability for any given observation sequence. Example observation sequences:

- i. Janet will back the bill
- ii. will Janet back the bill
- iii. back the bill Janet will