# **FoLT Tutorial 3 Summary**

# N-Gram Language Models:

# Part 1: Text Generation:

- we are gonna make a function has 3 parameters:
  - nltk.ConditionalFreqDist that takes a word and generate the next word that has the highest probability
  - start\_word that we are gonna use as the root of the phrase
  - number\_of\_words which gives us how many words will be generated after start\_word
- Code generate\_sentence:

```
def generate_sentence(
    bigram_model: nltk.ConditionalFreqDist,
    start_word: str,
    number_of_words=20
) → str:

    prev_word = start_word
    words = [prev_word]
    for i in range(number_of_words):
        # select the next word with the highest probability
        word = bigram_model[prev_word].max()
        prev_word = word
        words.append(word)
    return " ".join(words)
```

• Example of usage:

```
generate_sentence(cfd, "living", 4)
```

• Result:

'living creature that he said'

# Part 2: Build bigram model from a corpus

- In this part we are gonna use the austen-persuasion.tx fom the gutenberg corpus
- We are gonna make a function that does the following:
  - take all the words from austen-persuasion.txt
  - make a bigram out of those words using nltk.bigrams()
  - initiate bigram\_model for **generate\_sentence** using nltk.ConditionalFreqDist
- Code build\_bigram\_models\_from\_austen\_persuasion:

```
def build_bigram_models_from_austen_persuasion():
    gutenberg_words = gutenberg.words('austen-persuasion.txt')
    gutenberg_bigrams = nltk.bigrams(gutenberg_words)
    bigram_model = nltk.ConditionalFreqDist(gutenberg_bigrams)
    return bigram_model

bigram_model = build_bigram_models_from_austen_persuasion()
```

• Example:

```
start_word = "We"
generated_text_w_highest_prob = generate_sentence(bigram_model, start_word)
print("Generated sentence by selecting highest probability: \n{}".format(
    generated_text_w_highest_prob))
```

• Result:

```
Generated sentence by selecting highest probability:
We are not be a very much to be a very much to be a very
```

#### Part 3: Generate sentence from most common words

- This part is is similar to **PART 2**, but the only difference, is that we are gonna use randomness to pick 1 of the top 5 words
- Things you should know:
  - ConditionalFreqDist.most\_common(n: int) takes a number n as input and returns a list of n most common words and their counts:
    - ->[(word\_1, n\_1), (word\_2, n\_2), ...]
  - random.choice takes input as a list of tokens (top 5 possible tokens given the previous token) and randomly selects one.
- Code generate\_sentence\_from\_most\_common\_words:

• Example:

```
start_word = "We"
generated_text_w_most_common = generate_sentence_from_most_common_words(bigram_model,
start_word)
print("Generated sentence by randomly selecting most common next words: \n{}".format(
    generated_text_w_most_common))
```

• Result:

```
Generated sentence by randomly selecting most common next words:
We have done in his being the very little boy ," said he could do . I have no , she
```

### Part 4: Documents and spans with spaCy (We are gonna use concepts from Lecture 1)

- In this section we are gonna use spacy to get a token's :
  - Part Of Speech (Assigning word types to tokens, like verb or noun.)

- Dependency label (Assigning syntactic dependency labels, describing the relations between individual tokens, like subject or object.)
- Lemmatization (Assigning the base forms of words.)
- Code + Example:

```
raw = "Hard to judge whether these sides were good. We were grossed " \
        "out by the melted styrofoam and didn't want to eat it for fear of getting sick."

nlp = spacy.load("en_core_web_sm")

doc = nlp(raw)

span = doc[4:9]

print(span)

for token in span:
    print(token.text, token.pos_, token.dep_, token.lemma_)
```

# • Result:

these sides were good.
these DET det these
sides NOUN nsubj side
were AUX ccomp be
good ADJ acomp good
. PUNCT punct .