## assignment\_1\_B

April 6, 2025

## 1 EΠΕΞΕΡΓΑΣΙΑ $\Phi \Upsilon \Sigma IKH \Sigma \Gamma \Lambda \Omega \Sigma \Sigma \Lambda \Sigma - E$ 1

```
B. N-gram Language Models
                         Γ
    \mathbf{M}
         : E
                   Φ
    \mathbf{\Sigma}
           : I
                  K
    \mathbf{E}
          : 2025-03-22 | v.0.0.1
    1.1.1
             \mathbf{B}
                1: Φ
                                        Corpus
                                                                            Wall Street Journal.
    Χ
                                      NLTK,
                                                         199
                 corpus treebank
    \Delta
       • 170
                            (training set).
       29
                           (test set).
    T corpus
                            tokens
[3]: import nltk
     from nltk.corpus import treebank
     from collections import Counter
     # K
                corpus treebank
     nltk.download('treebank')
     # Λ
                  199
     files = treebank.fileids()
                    train
                             test
     train_files = files[:170]
     test_files = files[170:]
     train_sents = [sent for file in train_files for sent in treebank.sents(file)]
     test_sents = [sent for file in test_files for sent in treebank.sents(file)]
```

: {len(train\_sents)}")
: {len(test\_sents)}")

print(f"∏

print(f"∏

```
[nltk_data] Downloading package treebank to
    [nltk_data]
                    /Users/ioanniskoutsoukis/nltk_data...
    [nltk_data]
                  Unzipping corpora/treebank.zip.
                     : 3509
    П
    П
                     : 405
    1.1.2 B 2: Δ
                         Λ
                                  Α
                                             <UNK>
    Υ
                                                                      L:
                               tokens
       • T tokens
                                   3
                                                   <UNK>.
       • T
                                     \mathbf{L}.
[4]: \# \Sigma
                    tokens
     train_tokens = [token for sent in train_sents for token in sent]
     # M
     freqs = Counter(train_tokens)
                   :
     vocab = {word for word, count in freqs.items() if count >= 3}
     # Σ
                        <UNK>
     def replace_with_unk(sent, vocab):
         return [token if token in vocab else "<UNK>" for token in sent]
    1.1.3 В 3: П
                          П
                                   <B0S>
                                           <EOS>
    Κ
                            tokens:
       <BOS>
       • <EOS>
    Α
                                                   . H
                                                                 <UNK>
        tokens.
[5]: def prepare_sentences(sents, vocab):
         result = []
         for sent in sents:
             replaced = replace_with_unk(sent, vocab)
             result.append(['<BOS>'] + replaced + ['<EOS>'])
         return result
                                 n-qrams
     train_prepared = prepare_sentences(train_sents, vocab)
     test_prepared = prepare_sentences(test_sents, vocab)
```

[11]: print(train sents[0])

```
print(train_prepared[0])
                        :", freqs["Pierre"])
      print("Pierre
      print("Vinken
                          :", freqs["Vinken"])
     ['Pierre', 'Vinken', ',', '61', 'years', 'old', ',', 'will', 'join', 'the',
     'board', 'as', 'a', 'nonexecutive', 'director', 'Nov.', '29', '.']
     ['<BOS>', '<UNK>', '<UNK>', ',', '61', 'years', 'old', ',', 'will', 'join',
     'the', 'board', 'as', 'a', 'nonexecutive', 'director', 'Nov.', '29', '.',
     '<EOS>'l
     Pierre
                  : 1
     Vinken
                  : 2
     1.1.5 В 5: П
                             \mathbf{\Sigma}
                                     П
                                          (\mathbf{E}
                                                 1)
[15]: from collections import Counter
      import math
      import re
      # 0
                                     n-qrams
      def extract_ngrams(sentences, n):
          Δ
                         (
                                                tokens)
                        n-grams
          ngrams = []
          for sent in sentences:
              ngrams.extend([tuple(sent[i:i+n]) for i in range(len(sent)-n+1)])
          return ngrams
      # Lowercase
      def to lowercase(sents):
          return [[token.lower() for token in sent] for sent in sents]
      # Abstract digits
      def abstract_digits(sents):
          def abstract token(token):
              return re.sub(r'\d', '#', token)
          return [[abstract_token(token) for token in sent] for sent in sents]
      # E
               n-gram
                             add-k smoothing
      class NgramModel:
          def __init__(self, n, k, train_data):
```

 $\mathbf{E}$ 

```
11 11 11
                  n-grams (2 bigram, 3 trigram)
        n:
        k:
                        (add-k)
                                          <UNK>, <BOS>, <EOS>)
        train\_data:
                                tokens (
        n n n
        self.n = n
        self.k = k
        self.train_data = train_data
                          context ( . . bigram: (w1, w2), context: w1)
        self.ngrams = Counter(extract ngrams(train data, n))
        self.context = Counter(extract_ngrams(train_data, n - 1))
        # 1
                V (
                               smoothing)
        self.vocab = set(token for sent in train data for token in sent)
    def prob(self, ngram):
        n n n
        r
                        add-k smoothing
        n n n
        prefix = ngram[:-1]
        return (self.ngrams[ngram] + self.k) / (self.context[prefix] + self.k *_
 →len(self.vocab))
    def perplexity(self, test_data):
        n n n
               perplexity
        11 11 11
        ngrams_test = extract_ngrams(test_data, self.n)
        N = len(ngrams_test)
        log_sum = 0
        for ngram in ngrams_test:
            p = self.prob(ngram)
            log sum += math.log(p)
        return math.exp(-log_sum / N)
# Σ
                            (original, lowercase, digits)
def evaluate_all_versions(train_raw, test_raw):
    versions = {
        "Original text": (train_raw, test_raw),
        "Lowercase": (to_lowercase(train_raw), to_lowercase(test_raw)),
        "Abstract digits": (abstract_digits(train_raw), ___
 ⇔abstract_digits(test_raw))
    settings = [(2, 1), (2, 0.01), (3, 1), (3, 0.01)]
    results = {v: {} for v in versions}
```

```
for version, (train, test) in versions.items():
    for n, k in settings:
        model = NgramModel(n, k, train)
        ppl = model.perplexity(test)
        results[version][f'{n}-gram (k={k})'] = round(ppl, 2)

return results

evaluate_all_versions(train_prepared, test_prepared)
```

#### 1.1.6 $\Pi$ & $\Sigma$ Perplexity

Language model	Original text	Lowercase	Abstract digits
$\overline{\text{Bigrams } (k=1)}$	383.74	349.46	336.26
Bigrams $(k = 0.01)$	137.84	130.43	122.12
Trigrams $(k = 1)$	1505.81	1374.44	1319.79
Trigrams $(k = 0.01)$	464.06	427.07	386.26

```
• II
                 : T bigrams (2-grams) k = 0.01
                                                                  perplexity
                k = 0.01
• T
        k: H
                                              k = 1,
              n-grams,
                                   perplexity.
• E
        lowercase: H
                                            (lowercase)
                                                                               . A
                                          ( . . The
                                                    the),
        abstract digits: H
• E
                                                        (#)
                                                                        perplexity.
                         (hapax legomena),
```

#### 1.1.7 B 6: $\Pi$ $\Pi$ (E 2)

```
[29]: import random
      def generate_sentence(model, max_length=30, verbose=False):
          Δ
                                      n-gram.
          \boldsymbol{E}
          - model:
                          NgramModel
          - max_length:
          - verbose:
                           True,
          sentence = ['<BOS>']
          while len(sentence) < max_length:</pre>
              context = tuple(sentence[-(model.n - 1):]) if model.n > 1 else tuple()
              # T
                                           (
                                                <UNK>)
                           tokens
              candidates = [w for w in model.vocab if w != '<UNK>']
              # T
              probs = []
              for word in candidates:
                  ngram = context + (word,)
                  prob = model.prob(ngram)
                  probs.append(prob)
              # K
              total = sum(probs)
              probs = [p / total for p in probs]
              # T
              next_word = random.choices(candidates, weights=probs, k=1)[0]
              if verbose:
                  print(f"Context: {context} -> E : '{next_word}'\_
       ⇔(Prob={round(max(probs), 4)})")
              sentence.append(next_word)
              if next_word == '<EOS>':
                  break
          return ' '.join(sentence[1:-1]) # <BOS>, <EOS>
        Σ
      combinations = [
```

#### M 2-gram k=1:

-----

- \*  $\Pi$  1: <BOS> heart stopped medical Carla Carnival Old total start On globe Co plastic foreign number On Says Entertainment Sen. surprisingly Canada compound nearly Mercantile complex directly 51 equipment chemicals <EOS>
- \* II 2: <BOS> 8.50 N.J eager where remove plunged employees national gives Two suffer gains seeing Several Markey basic roof-crush sense Price disciplinary acne providing century as looming 13 conference few <EOS>
- \*  $\Pi$  3: <BOS> `` agreement widget people flat section 41 retired 30 So bridge covered multi-crystal turns In wait estimated household won standard expire Reupke becoming Money Co large front Computer <EOS>

### M 2-gram k=0.01:

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- \*  $\Pi$  1: <BOS> In her , a share . <EOS>
- \*  $\Pi$  2: <BOS> By earned fueled Trudeau refund card business appears US\$ margins approach it to help \*-1 always held media withdrawal year ending resolution of two funded premium 9 1\/2 <EOS>
- \*  $\Pi$  3: <BOS> And 47 employer looming received letters A. commodity markets Dorrance 2029 200 represent quarterly courts underwriters General 22\/32 Freeport-McMoRan Lake working Fairless mature fell agency live unlike stand <EOS>

#### M 3-gram k=1:

-----

- \*  $\Pi$  1: <BOS> 70 court where expressed federal 40 figures \* Klauser Law % made Group editorial Morgan simply Ms. court ship financings Although ones himself 12 St. 33 French During <EOS>
- \*  $\Pi$  2: <BOS> great produce understand Profit Poor pretax as run scheduled ballplayers suggests effective lesson index-arbitrage effects classroom family bulk compliance remained U.S.A. prevent Johnson rules trying men Nasdaq able <EOS>
- \* II 3: <BOS> lucrative Senate generally markets required respond

bankruptcy market respond themselves chairman Guinea damage staff treatment not asks Index elected Moody industry Illuminating Singapore defense promotion \*-3 fall impose  $\langle EOS \rangle$ 

# M 3-gram k=0.01:

- \* II 1: <BOS> accommodate hard Dodge race step parents less 47 compliance disgorge Chevrolet profitable smoking activity smoking designers Many took crisis Terms referred Richard branch larger According techniques Hahn opinion <EOS>
- \*  $\Pi$  2: <BOS> 30-year clients credit plastic senior district bells risks local three decided early Article \*-3 squeeze Smith publicly knew Chicago harder Instead church liquidity disciplinary Acquisition effect big specified <EOS>
- \* II 3: <BOS> nine dismissed statute rating heads debts total collected commodity required one-third beaten political widespread large Donald 1 noted hours 2009 hundred entered strategy quick N.C York-based charge France <EOS>

1.1.8 П	& Y	П	п п			
K		n-gram	$({\rm Bigram}/{\rm Trigram}$	k=1	k=0.01),	:
<ul><li>Ο</li><li>Π</li><li>Σ</li></ul>			, , "smoking ac	tivity sm	oking designers"	,
П		:				
• 1	n-gram		( .		)	
• E ,	•		<b0s></b0s>			
• H <b>k</b> :	<b>=1</b> s.		,		corpus,	
• A	$\operatorname{trigram}$		training corpus.			
$\Sigma$ ,			n-gram ,		,	