

Eidgenössische Technische Hochschule Zürich Swiss Federal Institute of Technology Zurich

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Course Assignment Episode 2

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Question 1: Distributions over Palindromes (15 pts)

In this problem, we will consider finding palindromic subsequences of a string. Let Σ be an alphabet¹ of characters. Suppose $\mathbf{x} \in \Sigma^*$ (where * denotes the Kleene closure) is a string. We denote a subsequence as $\mathbf{y} = \mathbf{x}_{k(|\mathbf{x}|)}$ where $k(|\mathbf{x}|)$ is an ordered (strictly increasing) subset of $\{1, \ldots, |\mathbf{x}|\}$ and $\mathbf{x}_{k(|\mathbf{x}|)}$ is the (ordered) sequence $\{x_i\}_{i \in k(|\mathbf{x}|)}$. We denote the set of all palindromic subsequences of \mathbf{x} by $\mathcal{P}(\mathbf{x})$.² In contrast to a substring, a subsequence need not be contiguous. We call a subsequence palindromic, or, simply, a palindrome if $\mathbf{y} = \mathbf{y}^{\mathrm{R}}$, i.e. if the subsequence \mathbf{y} is the same when read forwards and backwards.

- (a) Give a dynamic-programming algorithm that returns the maximum-length palindromic subsequence **x**. Analyze its time and space complexity **A Word to the Wise:** Yes, this is a very common coding interview question. And, yes, you can just Google the answer. Please don't—you're only cheating yourself out of an education.
- (b) Reformulate the dynamic program you derived in (a) to use semiring notation. Revisit the slides from Lecture 6 and determine which semiring is used in the special case you found in (a).
- (c) Now, exploit your semiring generalization to give an algorithm that counts the number of palindromes in **x**. Which semiring gives this result? Also, explain which semiring gives you a "recognition" algorithm, i.e. an algorithm returns "yes" if there is a palindrome in **x** and "no" otherwise.
- (d) Give an algorithm with the same runtime complexity as the one in (a) that computes the average length of all palindromes in **x**. Which semiring(s) did you use?
- (e) Give pseudocode for a prioritized version of your algorithm in (a). **Hint**: refer to Dijkstra's algorithm for inspiration.³ Analyze the time and space complexity of your algorithm.
- (f) This class is about probabilistic modeling over structures! So, let's model some palindromes—they are language after all. For any palindromic subsequence $\mathbf{y} = \mathbf{x}_{k(|\mathbf{x}|)}$, suppose you are given a score function $\operatorname{score}(\mathbf{y}, \mathbf{x}) : \Sigma^* \times \Sigma^* \to \mathbb{R}$. Now, consider a probabilistic model over all palindromes in \mathbf{x} defined as follows:

$$p(\mathbf{y} \mid \mathbf{x}) = \frac{1}{Z} \exp \{ \operatorname{score}(\mathbf{y}, \mathbf{x}) \}$$
 (1)

¹An alphabet is a non-empty, finite set.

²Note that this set may be exponentially large (in the size of $|\mathbf{x}|$).

³We do expect competence in undergraduate algorithms as this is a graduate-level computer science course. However, if you do not know Dijkstra's algorithm, please feel free to consult a friend in the course with a computer science background or a an algorithms textbook.

where we have the normalizer:

Consider the following grammar:

$$Z = \sum_{\mathbf{y}' \in \mathcal{P}(\mathbf{x})} \exp\{\operatorname{score}(\mathbf{y}', \mathbf{x})\}$$
 (2)

Give a slow, naïve algorithm to compute Z; analyze its time and space complexity. Now, explain how to construct a decomposable score(\mathbf{y}, \mathbf{x}) such that you can compute Z in the same runtime as your algorithm given in (a). **Hint:** There is more than one way to construct a decomposable score function that will work out—try to be creative and think about how to design a fun decomposable function.

Question 2: Am I a Semiring? (10 pts)

- (a) Define the set $A = \{a+b\sqrt{3} \mid a,b \in \mathbb{Z}\}$. Can you find a **0** and **1** such that $\langle A,+,\times,\mathbf{0},\mathbf{1}\rangle$ is a semiring? Prove your answer.
- (b) Is $\langle \mathbb{R}, \max, \times, 0, 1 \rangle$ a semiring? Prove your answer.
- (c) Let A be a finite set. Is $(2^A, \cup, \cap, \emptyset, \emptyset)$ a semiring? Prove your answer.

Question 3: CRFs as Context-Free Grammars (15 pts)

In this question, we ask you to show how to encode a conditional random field (CRF) for part-of-speech tagging as a weighted context-free grammar (CFG). Then, we ask you to manipulate the CKY algorithm to recover the forward algorithm presented in Lecture #6.

$$S \to (A \mid B) S$$

$$S \to (A \mid B)$$

$$A \to a$$

$$B \to b$$

- (a) Describe as concisely as possible the set of strings this grammar can generate.
- (b) For any string $\mathbf{s} = s_1, \dots, s_N$ in this set, what is the maximum number of parse trees under this grammar that yield \mathbf{s} ?
- (c) In this question, we denote spans as [i,j]; the length of [i,j] is j-i. We say a span is admissible under the grammar if we can build a complete parse tree with a constituent from i to j. In this setting, a complete tree means we have built a span [1, N+1] rooted at non-terminal S. Suppose we are parsing a string of length N, what are the spans admissible by the grammar under consideration? In your answer, please enumerate all spans of length ℓ for every $\ell \in \{1, \ldots, N\}$.
- (d) Modifying the pseudocode for CKY given in class, show that CKY on this grammar runs in $o(n^3)$ time. Give a tight worst-case runtime.
- (e) Given a set of part-of-speech tags $\mathcal{T} = \{T_1, \dots, T_{|\mathcal{T}|}\}$ and a set of words $\mathcal{W} = \{w_1, \dots, w_{|\mathcal{W}|}\}$, construct a grammar (without weights) of a similar structure to the grammar under consideration that encodes a first-order conditional random field (CRF) for part-of-speech tagging, i.e, a CRF that scores part-of-speech bigrams. You should assume any word in \mathcal{W} may take on any part of speech. **Hint**: The answers as you put down for (b) and (c) above should also be true for your grammar.

(f) Now show how to construct a score function $score(\cdot)$ for your grammar such that the unnormalized probability of a tree

$$p(\text{tree} \mid s_1, \dots, s_N) \propto \prod_{\pi \in \text{tree}} \text{score}(\pi)$$
 (3)

where π is a production in the grammar you designed, is the same as the unnormalized probability under a CRF:

$$p(t_1, \dots, t_N \mid s_1, \dots, s_N) \propto \prod_{n=1}^N \psi(t_n, t_{n-1}) \cdot \phi(t_n, s_n)$$
 (4)

Note that we define $t_0 := \mathsf{BOS}$ as a beginning-of-sequence tag. Your answer for score (\cdot) should be a function of ψ and ϕ .

(g) Your answer to (f) allows us to use CKY to compute the normalizer Z. Give a tight worst-case runtime bound for CKY as a function of number of part-of-speech tags $|\mathcal{T}|$ and the length of the sentence N under this specific grammar.

Question 4: Minimum Edit Distance as a Finite-State Machine (10 pts)

- (a) For this problem, we define minimum edit distance as the minimum number of insertions, substitutions or deletions required to convert one string into another.
 - (i) Design a weighted finite-state acceptor (WFSA) that accepts (only) all strings of edit distance ≤ 1 from the string "ilovenlp."
 - (ii) Now consider a WFSA that accepts only strings of edit distance $\leq d$ from some target string $\mathbf{s} = s_1, \ldots, s_N$. What is the minimal number of states required in this WFSA as a function of d and N?
- (b) Consider an n-gram language model, as in Lecture 5.
 - (i) Design a WFSA for a trigram language model that accepts only strings with probability > 0 under the model. A transition function δ , an initial weight function λ and a final weight function ρ should be defined. We expect edge cases at the beginning and end of input to be properly handled.
 - (ii) How many states does the n-gram language model have as a function of n and |V| where V is our vocabulary?