

**Natural Language Processing
Final Exam**

July 10th, 2025

1. **[2 points]** Introduce the general problem of ambiguity for natural language processing. Provide examples of ambiguity at several layers of linguistic descriptions, such as morphology, syntax, lexical semantics, etc.
2. **[6 points]** An English corpus T has been tokenized based on white spaces. The resulting dictionary and word frequencies are reported in the following table

word	hot	hottest	host	shot	shortest
freq	17	9	7	11	5

Apply the byte-pair encoding (BPE) algorithm to derive subword tokens for T , using the character ‘_’ to mark the end of each word. Report and comment each of the first eleven iterations (merge operations) in a run of the algorithm, including the frequency updates.

3. **[6 points]** In the context of N-gram techniques for statistical language modeling, answer the following questions.
 - (a) Introduce the general idea underlying backoff techniques. Define and explain the so-called stupid backoff.
 - (b) Define the technique of simple linear interpolation, and discuss possible strategies for the choice of the interpolation values.
4. **[2 points]** Introduce and discuss the basic ideas underlying concatenative morphology and template morphology, and provide some examples.

(see next page)

5. **[3 points]** In the context of large language models, introduce the relationships known as scaling laws. Discuss how these laws can be practically used to improve pre-training strategies.
6. **[5 points]** Consider the projective dependency tree consisting of the following unlabeled dependency relations

head	w_3	w_1	$\langle \text{ROOT} \rangle$	w_3	w_7	w_5	w_3	w_7
dependent	w_1	w_2	w_3	w_4	w_5	w_6	w_7	w_8

Answer the following questions.

- (a) Draw a graphical representation of the dependency tree above, with arcs directed from the head to the dependent.
 - (b) Apply to the above tree the oracle presented in syntactic parsing lecture, to construct a sequence of training instances for the arc-standard parser.
7. **[7 points]** In the course lectures we have presented two deep learning models that derive an attention distribution by computing a vector of so-called bilinear products. Describe the main idea of bilinear attention and the two applications in which we have exploited it, and report the equations of the two models.
8. **[2 points]** Discuss some of the good practices in prompt engineering we have presented in the LLM's lectures, and provide some examples.