

Subject Review

COMP90042

Natural Language Processing

Lecture 23

Semester 1 2021 Week 12

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Preprocessing

- Sentence segmentation
- Tokenisation
 - ▶ Subword tokenisation
- Word normalisation
 - ▶ Derivational vs. inflectional morphology
 - ▶ Lemmatisation vs. stemming
- Stop words

N-gram Language Models

- Derivation
- Smoothing techniques
 - ▶ Add- k
 - ▶ Absolute discounting
 - ▶ Katz Backoff
 - ▶ Kneser-Ney smoothing
 - ▶ Interpolation

Text Classification

- Building a classification system
- Text classification tasks
 - ▶ Topic classification
 - ▶ Sentiment analysis
 - ▶ Native language identification
- Algorithms
 - ▶ Naive-Bayes, logistic regression, SVM
 - ▶ kNN, neural networks
- Bias vs. variance
- Evaluation metrics: Precision, recall, F1

Part-of-Speech Tagging

- English POS
 - ▶ Open vs. closed POS classes
- Tagsets
 - ▶ Penn Treebank tags
- Automatic taggers
 - ▶ Rule-based
 - ▶ Statistical
 - Unigram, classifier-based, HMM

Hidden Markov Models

- Probabilistic formulation
 - ▶ Parameters: emission and transition probabilities
- Training
- Viterbi algorithm
- Generative vs. discriminative models

DL: Feed-forward Networks

- Formulation
- Designing FF networks for NLP tasks
 - ▶ Topic classification
 - ▶ Language model
 - ▶ POS tagging
- Word embeddings
- Convolutional networks

DL: Recurrent Networks

- Formulation
- RNN language models
- LSTM
 - ▶ Functions of gates
 - ▶ Variants
- Designing RNN for NLP tasks
 - ▶ Text classification: sentiment analysis
 - ▶ POS tagging

Lexical Semantics

- Definition of word senses, glosses
- Lexical relationships
 - ▶ Synonymy, antonymy, hypernymy, meronymy
- Structure of WordNet
- Word similarity
 - ▶ Path length, depth information, information content
- Word sense disambiguation
 - ▶ Supervised vs. unsupervised

Distributional Semantics

- Matrices for distributional semantics
 - ▶ VSM, TF-IDF, word-word co-occurrence
- Association measures: PMI, PPMI
- Count-based methods: SVD
- Neural methods: skip-gram, CBOW
- Evaluation
 - ▶ Word similarity, analogy

Contextual Representation

- Formulation with RNN
- ELMo
- BERT
 - ▶ Objectives
 - ▶ Fine-tuning for downstream tasks
- Transformers
 - ▶ Multi-head attention

Discourse

- Motivation for modelling beyond words
- Discourse segmentation
 - ▶ Text Tiling
- Discourse parsing
 - ▶ Rhetorical structure theory
- Anaphora resolution
 - ▶ Centering
 - ▶ Supervised models

Formal Language Theory & FSA

- Formal language theory as a framework for defining language
- Regular languages
 - ▶ Closure properties
- Finite state acceptors
 - ▶ Word morphology, weighted variant
- Finite state transducers
 - ▶ Weighted variant, edit distance, morphological analysis

Context-Free Grammar

- Center embedding
- Basics of CFG
- Syntactic constituent and its properties
- CFG parsing
 - ▶ Chomsky normal form
 - ▶ CYK
- English sentence structure (Penn Treebank)

Probabilistic Context-Free Grammar

- Ambiguity in grammars
- Basics of probabilistic CFGs
- Probability of a CFG tree
- Parsing
 - ▶ Probabilistic CYK
- Improvements
 - ▶ Parent conditioning
 - ▶ Head lexicalisation

Dependency Grammar

- Notion of dependency between words
- Universal dependency
- Properties of dependency trees
 - ▶ Projectivity
- Parsing
 - ▶ Transition-based
 - ▶ Graph-based

Machine Translation

- Statistical MT
 - ▶ Language + translation model
 - ▶ Alignments
- Neural MT
 - ▶ Encoder-decoder
 - ▶ Beam search decoding
 - ▶ Attention mechanism
- Evaluation: BLEU

Information Extraction

- Named entity recognition
 - ▶ NER tags, IOB tagging, models
- Relation extraction
 - ▶ Rule-based, supervised, semi-supervised, distant supervision
 - ▶ Unsupervised
- Temporal expression extraction
- Event extraction

Question Answering

- IR-based QA
 - ▶ Question processing, answer type prediction
 - ▶ Passage retrieval, answer extraction
- Reading comprehension
 - ▶ Models: LSTM-based, BERT
- Knowledge-based QA
- Hybrid QA: IBM Watson

Topic Modelling

- Evolution of topic models
- LDA
 - ▶ Sampling-based learning
 - ▶ Hyper-parameters
- Evaluation:
 - ▶ Word intrusion
 - ▶ Topic coherence

Summarisation

- Extractive summarisation
 - ▶ Single-document
 - Unsupervised content selection
 - ▶ Multi-document
 - Maximum marginal relevance
- Abstractive summarisation
 - ▶ Encoder-decoder models: copy mechanism
- Evaluation: ROUGE

Ethics

- Learning outcomes
- Arguments against ethical checks
- Core NLP ethics concepts:
 - ▶ bias, dual use, privacy
- Discussion of applications/use cases

Exam

Exam Structure

- Open book
- 120 points (40% for the subject)
- Gradescope
- Time: 120 minutes writing time +15 minutes reading time
- 3 parts:
 - ▶ A: short answer questions
 - ▶ B: method questions
 - ▶ C: algorithm questions

Short Answer Questions

- Several short questions
 - ▶ 1-2 sentence answers for each
 - ▶ Definitional, e.g. what is X?
 - ▶ Conceptual, e.g. relate X and Y, purpose of Z?
 - ▶ May call for an example illustrating a technique/problem

Method Questions

- Longer answer
- Focus on analysis and understanding
 - ▶ Contrast different methods
 - ▶ Analyse an algorithm/application
 - ▶ Motivate a modelling technique
 - ▶ Explain or derive mathematical equation

Algorithmic Questions

- Perform algorithmic computations
 - ▶ Numerical computations for algorithm on some given example data
 - ▶ Present an outline of an algorithm on your own example
- Not required to simplify maths (e.g. leaving fractions as $\log(5/4)$ is fine)

What to Expect

- Even coverage of topic from the semester
- Be prepared for concepts that have not yet been assessed by homework / project
- Prescribed reading is *fair game* for topics mentioned in the lectures and workshops
- Mock exam

Questions?

- Final survey: <https://forms.gle/CYBfYuEh46mjGqG86>