

# Subject Review

COMP90042

Natural Language Processing

Lecture 22



THE UNIVERSITY OF  
MELBOURNE

# Preprocessing

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

BPE  
max matches for word  
without space.

↓  
more aggressive.

lemma.

# N-gram Language Models

- Derivation
  - Smoothing techniques
    - ▶ Add-k
    - ▶ Absolute discounting
    - ▶ Katz Backoff → *re distribute based on this lower model*
    - ▶ Kneser-Ney smoothing
    - ▶ Interpolation
  - Evaluation
- take away mass, redistribute evenly*
- count based on versatile continuation counts*
- PP*
- $p_{-1} + \lambda_1 p_1 + \lambda_2 p_2 + \dots$*

# \* Text Classification

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

features

# Part-of-Speech Tagging

- English POS *manually annotate sentences*
  - ▶ 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

$P(\dots) P(\dots)$

- Training

- Viterbi algorithm  $\rightarrow$  decode. greedy decode.

- Generative vs. discriminative models

$\mathcal{G}$

work through

add features - performance

# DL: Feed-forward Networks

- Formulation
- Designing FF networks for NLP tasks
  - ▶ Topic classification
  - ▶ Language model
  - ▶ POS tagging
- Word embeddings *sparsity. how to get WE. from NN.*
- Convolutional networks *slide window / filter.*



# DL: Recurrent Networks

- Formulation
- RNN language models *why.*
- 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

*RST, bunch of relations between words*  
*Nucleus, satellite*

# Formal Language Theory & FST

- Formal language theory as a framework for defining language
- Regular languages
  - ▶ Closure properties
- Finite state acceptors *N-gram - FSA.*
  - ▶ Word morphology, weighted variant
  - ▶ *N*-gram language model as WFSA
- 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: ReVERB
- 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
  - ▶ Neural models: copy mechanism
- Evaluation: ROUGE

# Exam

# Exam Structure

- 40 marks
- Gradescope
- 2 hours in total:
  - ▶ 1 hour 45 minutes of writing
  - ▶ 15 minutes to upload answers
- 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
  - ▶ Outline or analyse an algorithm
  - ▶ 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