

# Lecture 1a: administration

William Webber ([william@williamwebber.com](mailto:william@williamwebber.com))

COMP90042, 2014, Semester 1, Lecture 1a

# COMP90042: Web Search and Text Analysis

# Course outline

- ▶ Text-document matrix (Week 1)
- ▶ Geometric models (Weeks 2 to 5)
- ▶ Probabilistic models (Weeks 6 to 9)
- ▶ Beyond the TDM (Weeks 10 to 12)

## Course outline: text-document matrix

- ▶ Parsing, stemming, stopping, and other transformations
- ▶ The “bag-of-words” representation of text
- ▶ The text-document matrix
- ▶ Term frequency scoring

## Course outline: geometric models

- ▶ TDM as vector space; cosine distance; document similarity
- ▶ Search and information retrieval using cosine distance
- ▶ Text clustering
- ▶ Text classification using support vector machines
- ▶ SVD methods and LSA

# Course outline: probabilistic models

- ▶ Probabilistic document similarity
- ▶ Language modelling for search and retrieval
- ▶ Text classification using:
  - ▶ Naive Bayes
  - ▶ Maxent (logistic regression)
- ▶ Topic modelling

## Course outline: beyond bag-of-words

- ▶ POS tagging, phrase and named entity identification
- ▶ Anchor text, page rank, and other graphical models
- ▶ Cross-lingual information retrieval
- ▶ Perhaps other things ... (e.g. CJK segmentation)

# What we will not be covering

- ▶ Machine translation
- ▶ Sentiment analysis
- ▶ Deep parsing or other advanced NLP techniques
- ▶ Information extraction
- ▶ Automatic text summarization
- ▶ Text compression
- ▶ Engineering, optimization, and efficiency issues
- ▶ ...



## About myself

- ▶ Industry consultant in text analytics (not an academic, though active as independent researcher)
- ▶ Masters by thesis in distribution information retrieval (UMelb)
- ▶ PhD in information retrieval evaluation (UMelb)
- ▶ Postdoc in e-discovery (University of Maryland)

# Pre-requisites and outcomes

## Pre-requisites

- ▶ Python programming skills
- ▶ COMP90049 / COMP30018, “Knowledge Technologies”

## Outcomes

- ▶ Practical familiarity with range of text analysis technologies
- ▶ Understanding of theoretical models underlying these tools
- ▶ Competence (and courage!) in reading research literature (including maths!)

## Contact hours

Tuesday 9:45 - 10:45am	Th:Doug McDonell-309 (Denis Driscoll Theatrette)
Wednesday 2:15 - 3:15pm	Th:Chemical & Biomolecular Engineering-Theatre
Wednesday 4:15 - 5:15pm	Alice Hoy-236 (Comp Lab)
Thursday 2:15 - 3:15pm	Alice Hoy-222 (Comp Lab)

# Consultation

- ▶ No office on campus!
- ▶ Regular consultation, Wednesday 3:15pm to 4:15pm; room to be announced
- ▶ Email me at [william@williamwebber.com](mailto:william@williamwebber.com)
- ▶ I will curate a mailing list

# Expectations on students

You are CS masters students at Australia's top university. I expect competence accordingly.

- ▶ You will be assessed to this standard!
- ▶ Code to be neat, (comparatively) bug-free, documented
- ▶ Expected to be able to install software, packages, tools
- ▶ Expect fluency in written English

I will use all tools available to me to detect plagiarism, and will be *very strict in prosecuting it*.

## Student weekly work

- ▶ Readings will be provided in advance of lectures (not this week though). Expect c. 2 hours of reading.
- ▶ Weekly worksheets need to be finished in full. Workshops are *only to begin* these worksheets.
- ▶ All workshop work (and all project work) to be committed to shared subversion repository (details at first workshop).
- ▶ We are over-full on workshops. Hopefully students will drop out.
- ▶ I am taking both workshops.

# Assessment

- ▶ 10% on workshop work, as committed to Subversion repository. (Option: you may instead give 20 minute research presentation at workshop, but you need to email me by the end of next week to take this option).
- ▶ 40% on projects. There will be two projects:
  - ▶ First project due end of Easter non-teaching break. Set work.
  - ▶ Second project due at end of semester. Free choice research-y project.
- ▶ 50% on final exam. This will be closed book. Everything on the course (in readings, lectures, workshops, projects) is examinable!

# Learning resources

- ▶ Lecture notes are primary resources.
- ▶ No text book as such, but following texts are useful:
  - ▶ Christopher D. Manning, Prabhakar Raghavan and Hinrich Schtze, *Introduction to Information Retrieval*, Cambridge University Press. 2008. (Available free online)
  - ▶ Charu C. Aggarwal and ChengXiang Zhai (ed.), *Mining Text Data*, Springer, 2012. (c. \$US 150)
- ▶ Citations to other readings will be given as required.
- ▶ Wikipedia is a very good place to start!