# Week 1: The Syllabus (CMPT 419 / 980)

Fall 2025.

Course webpage is linked from Canvas (Canvas is very minimal for now.)

Press Space for next slide

#### **Course Structure**

#### Lecture Schedule

#### Monday (1hr)

Discuss previous week's readings

Introduce new materials

Start lecture content

#### • Thursdays (2hr)

Finish lecture (if needed)

Discussion

Class activity and/or lab time (1hr)

Occasionally: Quizzes

## **Communication Policy**

- Questions at start/end of each session
- Email response: 2-3 business days
- Prefer in-class questions for:
  - Faster responses
  - Benefit to all students

# **Reading Components**

## **Reading Structure**

- Mandatory Readings: Core course material, weekly responses required
- Optional Readings: Theme-organized, abstracts/intros, based on interests

Note: Al assistance discouraged for reading responses. I prefer bullet points over LLM-generated text!

# **Reading Schedule**

#### **Timeline**

- Finalized Monday of Week X-1
- Complete by Monday of Week X
- Responses due before class

#### **Course Modules**

# **Course Organization**

Module	Weeks	Focus
1	1-4	Admin & Human-centred / Data-centred Al Frameworks
2	5–7	Technical work in data valuation, data scaling, and algorithmic collective action
3	8–10	Online platforms, content ecosystems, and data
4	11–13	Frontiers in Data Governance

#### **Grading Structure**

- "Reading Responses": 10
- "Coding Assignments": 20
- "Quizzes": 20
- "Final Project": 50

#### **Detailed Version**

- 10% reading responses (12 total, drop lowest 2 using Canvas, so each of your top 10 responses effectively is worth 1%)
- 20% coding assignments (4 total; 5/5/5/5, drop lowest 1 using Canvas, so each of your top 3 assignments effectively is worth 6.6.67%)
- 20% quizzes (2 total; 10/10; may adjust scoresfor difficulty)
- 50% final project (5% project proposal, 45% actual project; must submit a written document and a presentation for both)

# **Course FAQs**

#### **Attendance**

- Not directly graded
- Participation expected
- (Maybe) Cold-calling for reading responses
- Stay home when sick
- No need to email for absences

## **Coding & Tools**

#### FLEXIBLE.

- Python for ML/Data Science
- JavaScript for web programming
- LLM assistance allowed (with caveats)
- Free tools always available
- 4 assignments + 1 project

## **Group Work Policy**

- Some opportunities available. Generally optional
- Contribution statements required
- Code/writing review mandatory
- Guidelines per assignment

#### **Assessment Structure**

- Likely TWO quizzes
- No midterm/final
- In-class format
- Make-up options available
- Based on lectures & readings

# Al Tool Usage

Allowed with: Logs, attribution, documentation

#### **Example citations:**

- Produced by model XYZ
- Generated with ChatGPT
- Al-assisted, heavily edited

## Example

I generated this deck by first manually writing a draft of the syllabus as syllabus.md, then asked Claude to style it.

# Agenda (if time)

- More details about modules
- Loose prerequisites & crash course
- Finding research papers
- Managing references & notes

#### **Module 1: Intro**

#### Goals:

- Exposure to human/data-centric Al frameworks
- Learn frameworks aimed at researchers/designers
- Key question:
  - When would a human- or data-centric approach change your Al product?

#### Frameworks We'll See

- Human-Centered ML (Chancellor)
- DataPerf (data-centric AI)
- HCAI (Schneiderman)
- Value Sensitive Algorithm Design (Zhu)
- FairML
- "Public AI" (some of my work!)
- More to come!

# **Module 1 Learning Goals**

- Identify similarities/differences between approaches
- Apply concepts to scenarios (e.g., product design)

# Module 2: Technical work in data valuation, data scaling, and algorithmic collective action

- Data influence: Effect of each observation
- Scaling: Predict performance vs. dataset size
- Core idea: Reason about data counterfactuals. And social implications!

#### **Data Valuation Key Reading**

Hammoudeh & Lowd (2024) Training data influence analysis and estimation: A survey

# **Data Scaling Key Readings**

Hestness et al. (2017) Deep learning scaling is predictable Villalobos et al. (2024) Will we run out of data?

## Why Focus on Data?

- Some HCAI work doesn't focus on data
- Legislation, licensing, behavior change
- We'll focus more on data levers (everyone is a data creator!)

# **Module 2 Learning Goals**

- Explain influence calculation & applications
- Describe scaling patterns
- Interpret scaling plots
- New forms of collective action

#### Module 3: Platforms & Content

- Peer production (Wikipedia, Linux)
- "User-generated content"
- Quantitative reasoning about "content ecosystems"

#### Module 4: Frontiers in Data Governance

- Overlap: Governing, markets, ecosystems
- Protocols for dataset maintenance
- Opt-in/out policies
- Platform governance
- Public interest Al

# **Voting & Markets**

- Vote with/for data
- Market conditions change counterfactuals

#### If time:

• In-person and/or Google doc "interest formation" network