ECE 1786:
Creative Applications of Natural Language Processing

Fall 2022

Instructor: Jonathan Rose

Department of Electrical & Computer Engineering

Land Acknowledgement

Toronto is a city of immigrants. My parents came here in 1952, and it is very likely that you or your parents are very new to this country. This is not a new feature of Toronto. For over 15,000 years Toronto has been a gathering site for humans including the Huron-Wendat and Petun First Nations, the Seneca, and most recently, the Mississaugas of the Credit River.

Today, Toronto is still a meeting place for Indigenous people from across Turtle Island (North America), and immigrants, both new and old, from across the world. I am grateful to have the opportunity to work in this community, and on this territory.

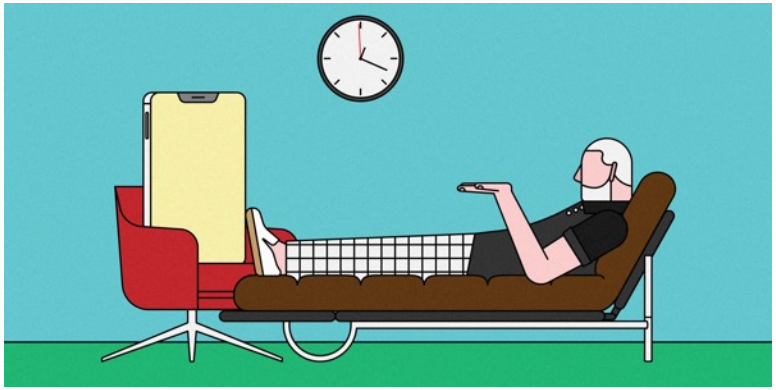
Welcome to ECE 1786!

- This course is about methods for automatic understanding and generation of human language
- Language is a pathway in and out of the brain!
- Until recently, it was difficult to connect humans and computers with language
- Course covers inventions that have made sophisticated communication possible and has led to other remarkable capabilities

Those Capabilities Include …

- In your lifetime, you will be able to talk to a computer-based doctor to get decent basic medical care

- You'll be able to get mental-health *talk therapy* any time, anywhere from a computer.



My Current Research Focus

- Automatic Talk Therapy for Smoking Addiction
  - As one example of talk therapy

Scripted

BOT: What will it look like when you have made this change in your smoking habit?

USER: I will feel good about my accomplishment, feel health improvements and will be able to save more money

BOT: You think that by making this change you will be healthier and be able to save more money.

Generated!

Scripted

BOT: What will it look like when you have made this change in your smoking habit?

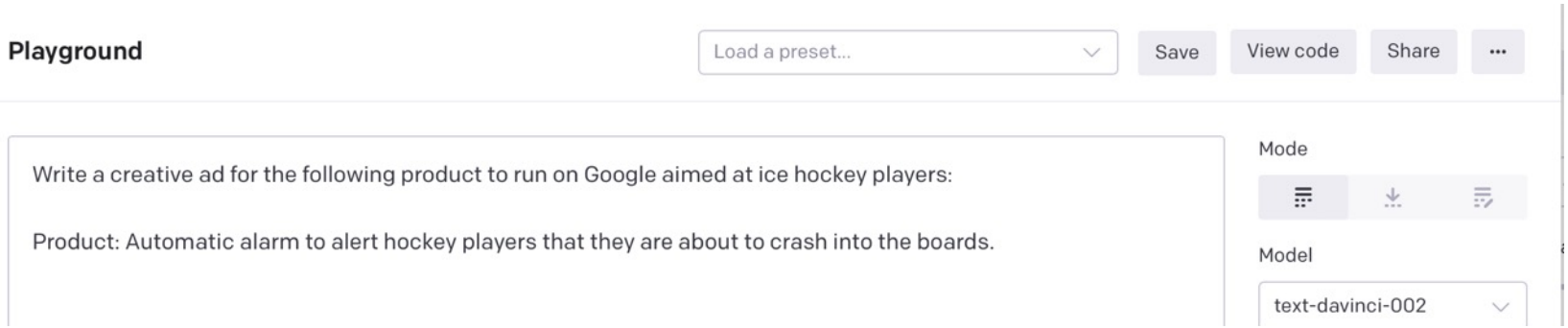
USER: I wonder if I could do it, because I'm an addict

BOT: You are concerned that you would be unable to stop if you tried.

Generated!

GPT-3 Playground

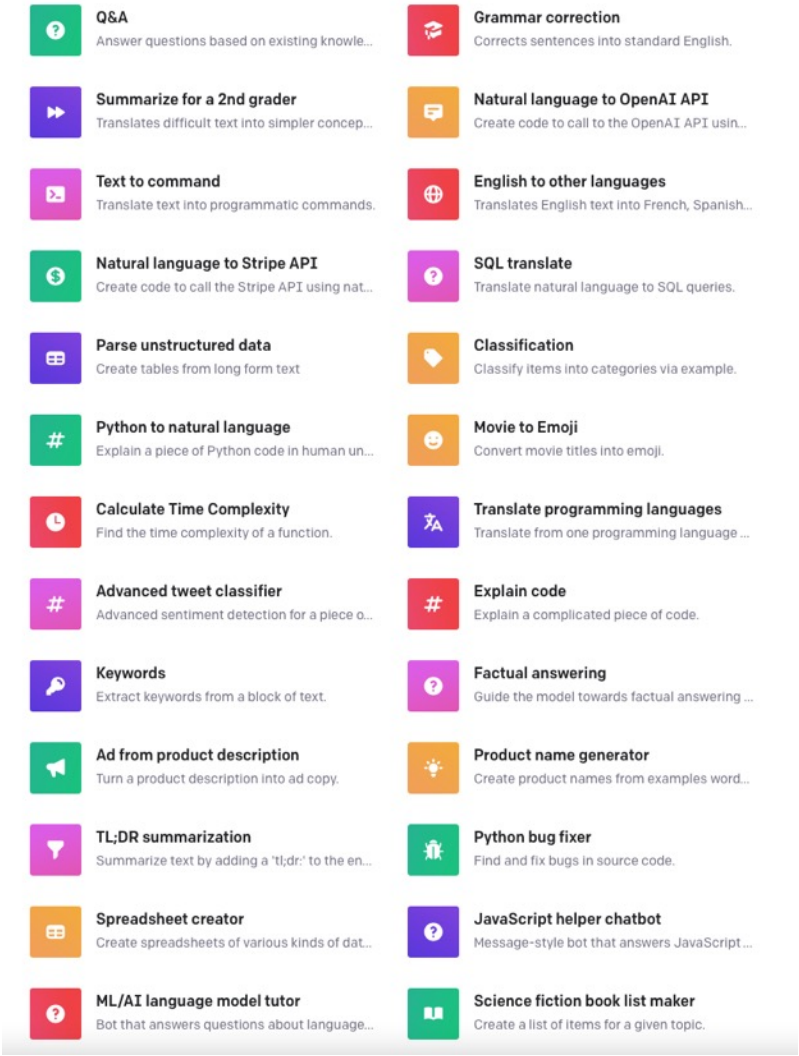
- State-of-the-art language capability at your fingertips!
  - https://beta.openai.com/playground



Looking to avoid those costly and painful collisions with the boards? Look no further than the Board-B-Gone Automatic Alarm! This handy little device attaches to your hockey stick and uses sensors to detect when you're about to crash into the boards. It then emits a loud alarm to warn

More Examples from GPT-3

- See https://beta.openai.com/examples

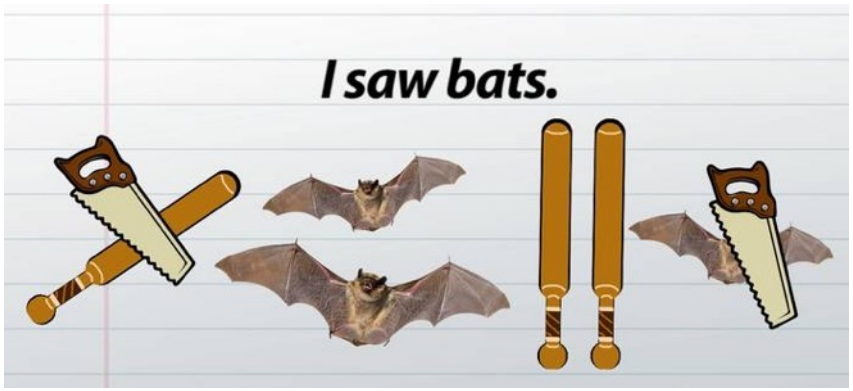


NLP Advances are Part of DALL-E-2

- One of several amazing models that turn can turn words into remarkable pictures:
  - “A closeup wide angle photo of a cat wearing sunglasses”
  - See https://www.reddit.com/r/dalle2/ for more examples

What Makes Language Difficult

- The ambiguity of language:
  1. Multiple meanings of words
     - tank, bank, duck ….
  2. Context needed to figure out:
     - Milk drinkers are turning to powder
     - Juvenile Court Tries Shooting Defendant
     - Grandmother of Eight makes Hole in One

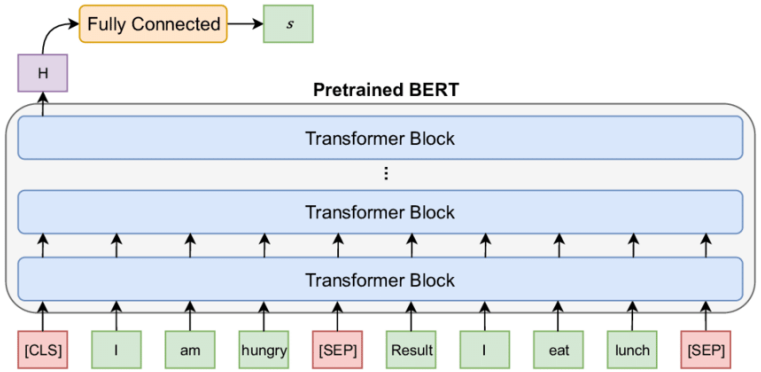


Is Now Much Easier to Deal with

- Using neural-net-based approaches to NLP
- Since the Deep Learning revolution began in 2012, many fields have been changed
  - A key step in word embeddings occurred then
- Key next steps for NLP occurred in 2018
  - 'Transformer Architecture' Vaswani et. al
  - BERT - https://huggingface.co/bert-base-uncased
  - GPT-2 - https://huggingface.co/gpt2
  - GPT-3 - https://openai.com/blog/gpt-3-apps/
  - Bloom - https://huggingface.co/bigscience/bloom
  - LaMDA - https://blog.google/technology/ai/lamda/

Success in Key Applications

1. Translation
2. Classification
3. Dialogue



Related Courses & Focus of this Course

Course Focus vs. Related Courses

- Course focus is neural-network approaches to NLP
- NLP is a 60-year-old field which previously used a procedural methodology:
  - Based on human understanding of language
  - Referred to as 'Computational Linguistics'
    - Grammar – e.g. parts of speech tagging – noun, verb
    - Parsing of language to interpret
  - CSC 485/2501 - Title: **Computational Linguistics**
  - https://www.cs.toronto.edu/~gpenn/csc485/
  - Principally taught by Professor Gerald Penn
- We will make some, limited use of computational linguistic approach

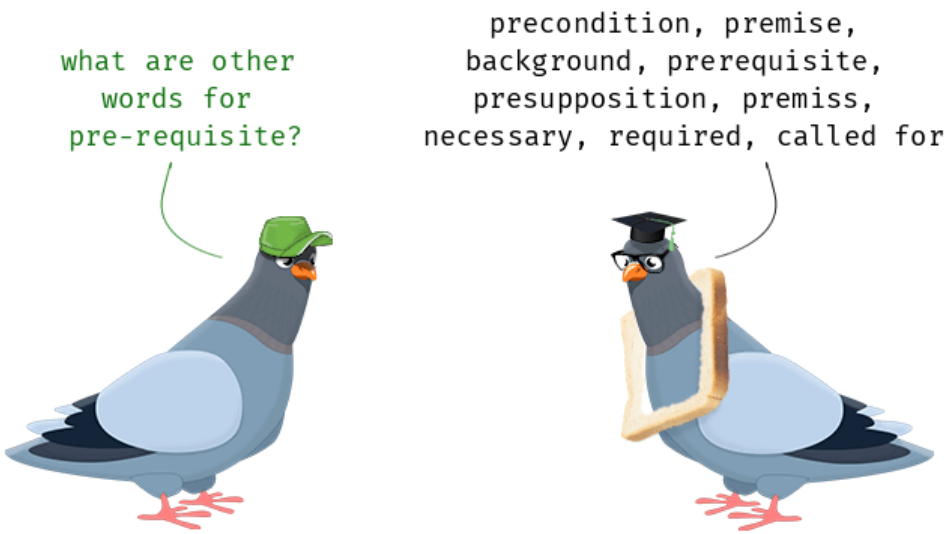
Course Focus vs. Related Courses

- A more closely related course in Computer Science
- CSC 401/2511 - **Natural Language Computing**
  - https://www.cs.toronto.edu/~frank/csc401/
  - Principally taught by Professor Frank Rudzicz
- CSC 2511 has a broader coverage of topics surrounding language
  - a super-set of this course,
  - including things such as Markov Models, Entropy, Automatic Speech Recognition, retrieval and dialogue
- ECE 1786 focuses more narrowly on word embeddings, deep learning, statistical language models,Transformers
  - more depth less breadth
  - more of an engineering, software focus

Learning Outcomes: Understanding

- Word embeddings
- Use of word embeddings in classification tasks
- Transformers
  - Global structure
  - Training
  - Attention, Encoder-Decoder, Transformer Stack
  - Classification
  - Probabilistic & Auto-regressive Generation
- PyTorch, Huggingface
- Limitations and Biases of Language models
- Navigation of open-ended problems in a project

Course Pre-Requisites



We Need To Talk about Pre-requisites

ECE1786H Creative Applications of Natural Language Processing

Prerequisites: APS360H, CSC311H, ECE324H, ECE1513H, or equivalent

This course is about learning and applying deep learning approaches in Natural Language Processing. We assume that a prior course in neural networks training and software has been taken. This course begins with the basics of word-level embeddings – their properties and training. These form the basis of neural-network-based classifiers employed to do classification of sentiment, named entity recognition and many other language tasks. We will include a description of the Transformer deep learning network – its structure, training and use. This will include the use of the transformer as a classifier, but also as in generative mode, in which language is produced in response to input language. Much of the learning will be applied in 3 or 4 hands-on programming assignments. Students will also do a major project of their own choosing to make use of these capabilities. There will also be an emphasis on developing good software engineering practices in the creation of the project.

- Graduate office does not check pre-requisites
- I will do that now, in two ways …

This Course does not introduce ML

- You **must** have background from a course in machine learning that has depth in neural networks
  - Otherwise, you won't understand what to do in the assignments!

- Acceptable UofT undergraduate courses (any one)
  - ECE 324 Machine Intelligence, Software and Neural Networks
  - ECE 421 Introduction to Machine Learning
  - APS 360 Applied Fundamentals of Deep Learning
  - CSC 311 – Introduction to Machine Learning
  - CSC 413 – Neural Networks and Deep Learning

- ECE 1513 Introduction to Machine Learning

Pre-requisite, cont'd

- If undergraduate degree is from elsewhere (most of you):
  - You must have taken a course that is equivalent to one of these University of Toronto courses
  - Next few slides describe the necessary background in detail

- Everyone must fill out the Quercus survey, posted on the course main website,
  - You'll say which course you've taken, and provide a link to any non-UofT course
  - You'll try to answer questions based on the following….

What to Know Already

- Machine Learning
  - Classification vs. regression (Logistic v. Linear regression)
  - Binary vs. multi-class classification
  - Supervised vs. unsupervised learning
  - Data labelling

What to Know Already

- Basic Neural nets and training
  - Linear neurons – weights and biases
  - Non-linear activation functions, e.g sigmoid, ReLU …
  - Multi-layer perceptron (MLP)
  - Loss functions; binary & multinomial cross entropy
  - Softmax function
  - Training, Validation and Test sets
  - Training & Validation ‘curves’
  - Gradient Descent
  - Stochastic Gradient Descent
  - Hyper-parameter tuning
  - Regularization: normalization, dropout, weight decay

What to Know Already, continued

- Advanced Neural Networks
  - Convolutional Neural Networks for computer vision (CNN)
    - Kernels, batch normalization
  - Recurrent Neural Networks (RNN)

- Transfer Learning:
  - Pre-trained networks
  - Fine-tuning of pre-trained networks

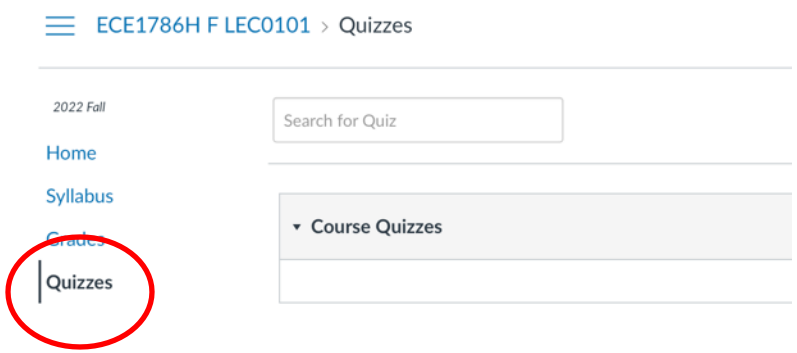
What to Know Already, continued

- **How to build all that** with Software Frameworks
  - Experience in Tensorflow or PyTorch (implies Python experience)
  - Should have written full training and test loops applications with significant data sizes
  - Tensors, shape
  - Numpy framework,
  - **How to debug a neural network**

- Data Science concepts
  - False positive, False Negative, True Positive, True Negative
  - Sensitivity, Specificity
  - Area Under the ROC Curve
  - Confusion Matrix

For you to do after finishing this video

- Go to the Quercus Website for this course
  - Assuming you're enrolled and not on the waitlist
- Click on 'Quizzes' on the left-hand side
  - Answer all of the questions to the best of your ability
  - This will not be graded
  - You must give evidence that you have the pre-requisite knowledge and training to remain in course



Teaching Philosophy

Teaching Philosophy

1. Teach less, but more in-depth
   - try to explain the intuition behind each approach
   - believe with solid grounding, you can learn more on your own

2. Your learning comes from doing.
   - From writing or working with software, experimenting with it
   - Answer questions about the results of experiments

3. Do open-ended projects
   - An engineer can navigate an open-ended project
   - Learn to conceive a project & describe it: **what** & **why**
   - Then **do it** – figure out **how**, and make it happen
   - Then **describe it**

Course Structure & Grading

Grading

Item	Fraction of Course
Assignments (4)	40%
Project Proposal Document/Presentation	10%
Project Interim Report	10%
Project In-Class Final Presentation	10%
Project Peer Reviews	5%
Project Final Report/Software	25%
Total	**100%**

Project is 60%

Assignments

#	Date Assigned*	Assignment	Due*
1	September 13	Word Embeddings – Properties, Meaning and Training	September 26th
2	September 27	Classification of Subjective/Objective Text	October 10
3	October 11	Understanding, Training and Using Transformers	October 24
4	October 25	Question Answering Using Transformers	November 14

*Tentative dates, subject to change

Peer Review

- Each individual student will be asked to provide feedback to other groups on their:
  - Proposal presentation/document
  - Interim report
- Asked for specific/useful feedback to group's work
- Feedback/commentary will be graded for quality

Textbooks & Hardware Acceleration

Required Text is Free: **Speech and Language Processing (3rd Edition Draft)** by Dan Jurafsky and James H. Martin:

https://web.stanford.edu/~jurafsky/slp3/ed3book_jan122022.pdf

2nd Edition has a complete first chapter (missing above):

https://github.com/rain1024/slp2-pdf

For the assignments and project, it is suggested that you purchase the for-pay Google Colab Pro:

https://colab.research.google.com/signup

The Project

The Project

- Done in Groups of 2
  - No groups of 1 or 3
- The topic is of your own choosing
  - must be approved by instructor
- Must relate to Natural Language Processing & the material covered in this course
- It should be an application of NLP
  - Need permission to do a research project on NLP itself
- Must collect and label some of your own data
  - Cannot simply work on an existing dataset
  - Why? Data labelling is at the core of all ML/AI work
  - However, this can must be careful not to do too much!

Project Stages

1. **Forming Groups**
   - Should be done by end of October
2. **Project Approval-in-Principle**
   - via email; due October 24th
3. **Project Proposal/Plan**
   - Document Due October 31
4. **Proposal & Plan Presentations**
   - November 1
   - **NOTE EXTRA LECTURE Tuesday November 1st, 6-8pm**
5. **Interim Report**
   - November 21
6. **Final Presentations**
   - December 6th; extra lecture that week as well.
7. **Final Report Due December 7**

- All dates are tentative

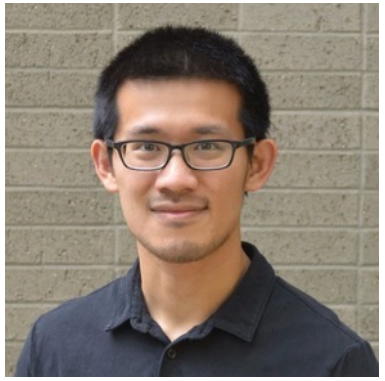
Course Instructor/TAs

Instructor Bio: Jonathan Rose

- Professor in Electrical & Computer Eng since 1989
  - Bach, Master's & PhD from UofT, Post-Doc at Stanford
- Research: Automation of Medicine/Mental Health
  - Automation of Mental Health using Machine Learning/NLP
    - Focusing on conversational systems for mental health
  - Previously: Field-Programmable Gate Arrays (FPGAs)
- Entrepreneurial/Business Experience:
  - Co-founder of Right Track CAD Corp in 1998
  - Software Engineering Director of Altera 2000-2003, now Intel
- Administration:
  - ECE Dept. Chair of ECE 2004-2009;
  - Chair Engineering Entrepreneurship **Hatchery** Advisory Board
- F.IEEE, F.ACM, F.CAE, FA NAE, FRSC, Sr Fellow Massey College

Teaching Assistants

- Zining Zhu
  - Ph.D. Candidate in Computer Science
  - Thesis: Method and Applications for Probing Deep Language Models
  - Zining.zhu@cs.toronto.edu



- Andrew Brown
  - M.A.Sc. Candidate in ECE
  - Thesis: Using NLP for a Smoking Cessation Chatbot
  - andrewm.brown@mail.utoronto.ca

Two Course Websites:

- UofT Quercus (https://q.utoronto.ca/courses/285821) for
  - Assignments release and submitted
  - Grades
  - Announcements

- **Piazza** website for a discussion board
  - See announcement on Quercus that tells you how to access
  - Email me if you don't have access to Quercus & I will add you

Questions?

- Post them to the Piazza discussion board for this course
- **or**, bring them to the first lecture on
  - Day: September 13th, 2022
  - Time: 10am-12 noon
  - Place: Galbraith Building, 35 St. George Street, Room 221:
  - See you there then!