

# Deep Learning Projects

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# Deep Learning Projects

## Overview

- 25% of your grade
- Groups of 3 students
- Applying deep learning techniques to real-world data
- Very open-ended...

# Course Projects

## **Important dates**

- March 23: Proposal Due
- May 1: Presentations
- May 3: Presentations
- May 14: Final Reports Due

# Project Proposal

## **Project Proposal** (due Mar 23)

- 1-2 pages
  - Who is in your group?
  - What dataset(s) do you plan to use?
  - What deep learning tasks?
  - What techniques do you plan to employ?
- 10% your project grade

# Presentation

- 12-13 minute oral overview of the project
  - Problem addressed
  - Approaches used
  - Evaluation measures
  - Resulting performances
  - Lessons learned?
- 2-3 minute question and answer
- 30% project grade

# Final Report

- Final Report
  - Use the Latex or WORD template from <https://nips.cc/Conferences/2015/PaperInformation/StyleFiles>
  - Max number of pages: eight (8), including figures and tables but excluding references.
  - One additional ninth page containing only the cited references is allowed.
  - At least 4 pages (excluding reference) recommended
  - Summary of problem, approach, evaluation, results
  - What did you learn from this project?
  - 60% project grade **(NOT the longer report the better!)**

# Data Data Data

- Success depends on finding lots of data
  - Can't learn something if you have no relevant data
- 30 Amazing (And Free) Public Data Sources For 2018

<https://www.forbes.com/sites/bernardmarr/2018/02/26/big-data-and-ai-30-amazing-and-free-public-data-sources-for-2018/#728a29985f8a>

- Other sources
  - UIC projects (if you have data)
  - Kaggle.com
  - mldata.org
  - CORGIS Datasets (<https://think.cs.vt.edu/corgis/>)
  - US Census bureau (time use data)
  - Microsoft Research GPS dataset
  - Divvy!

# Evaluation

- Partially based on results
  - Does approach work well for problem?
- Partially based on effort
  - Did you put effort into evaluating different techniques, exploring the features used, etc.?
- Can I just implement an algorithm from a paper?
  - Yes, as long as the algorithm is not too trivial and you add your own insights.
- Can I use the code of other authors? Yes, but more in other aspects
- Can I try something that's not covered in the course? Sure
- Can I writeup my own code? Sure
- Can I do a mathematical project – Yes!

**Implementation of algorithm + experiment on (big) dataset + insights from it + mathematical analysis**



# Example Projects

- Entropy-SGD optimization with applications to Fourier domain convolution
- Improved Unsupervised Representation Learning with Deep Convolutional Generative Adversarial Networks
- Question-Answer Matching using Deep Learning Models
- Natural Language Question Answering with Memory Networks: A Comparative Analysis
- Deep residual learning for large amount classes classification problem
- Keyphrase Extraction Using Deep Recurrent Neural Networks on Twitter
- Semantic Image Inpainting with Context Encoders
- Recurrent Attention Model on Sentiment Analysis
- Representation and regularization study of deep and wide neural network for image recognition

# Math project

- Novelties on the mathematical side are surely more related with theory rather than application. So a mathematical project on deep learning is likely to explore some unsolved mathematical problems.
- Examples include deriving new invariance convolution networks (e.g. [link](#)), or prove recovery guarantees (e.g. [link](#)), or improved optimization guarantees such as escaping saddle points (e.g. [link](#)). Obviously these papers themselves are way more than a course project. However, based on them some extensions could make good topics. For example
- 1. Verify whether assumptions made in those theoretical papers actually hold on real datasets. For example, the local strongly convex assumption in [link](#), the independence assumptions made in [link](#).
- 2. Make improvements in some aspects on the existing algorithms, and give solid mathematical justifications. For example: [Entropy-SGD](#) as improvements to SGD/ADAM, and computing pooling in frequency domain as in [link](#), improving upon the standard spatial domain pooling (after all, Fourier transform has been used extensively in convolutional neural networks [[link](#)]).

# Questions?