

# Machine Learning in Practice

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What do you want/expect to learn from this class?

Why does this class exist?

# What we want you to learn from this class

- How to responsibly and effectively solve real-world problems using ML
  - Understand the \*entire\* Machine Learning process (and get hands-on experience doing most of it)
  - Build (and use) reusable ML pipelines
  - Learn how to formulate ML problems, use, understand, evaluate, and communicate ML methods (that you have covered in earlier classes) in the context of a real problem

# How is this course different than typical ML classes you've taken before?

We'll assume everyone knows

- Methods/algorithms/models
- Assumptions behind them
- How to implement them

And focus on everything else that comes before the matrix and after the models are built (99% of the work done in a real-world project)

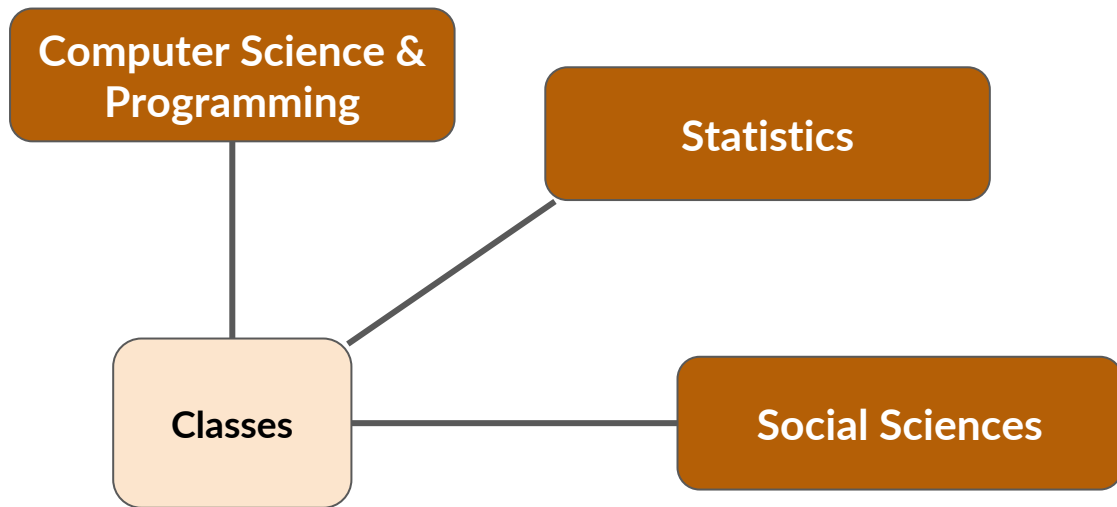
# Grading

- Pass/Fail
- We want the focus to be on learning and not on the grade.
- Levels of learning
  - Exposed to information covered in this course
  - Applying the information (correctly) covered to the class project
  - Generalizing to the next ML problem you tackle

# Pre-requisites

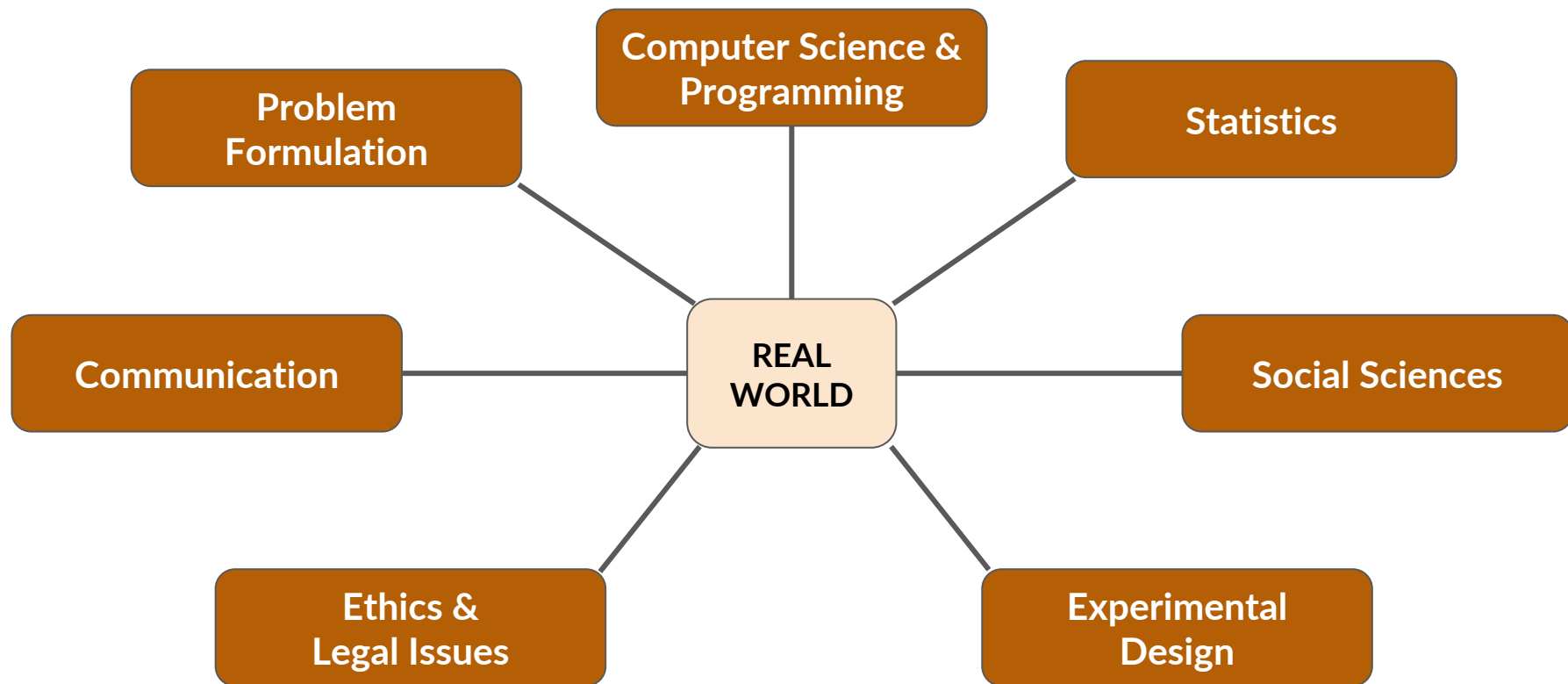
- Machine Learning (methods and overall process)
- Python (pandas, sklearn, tensorflow, matplotlib)
- Ideally: experience with SQL, command line (bash), git(hub), working on remote servers

# Skills needed to solve real-world problems (with ML)





# Skills needed to solve real-world problems (with ML)



## Scope

- Goals, Actions, Data, Analysis, Ethics



## Data

- Get Data
- Store Data
- Link Data



## Exploration

- Entities
- temporal
- Spatial
- ...



## Modeling

- Rows
- Labels
- Features
- Models



## Model Selection

- Train-Test Splits
- Performance Metrics



## Model Interpretation



## Dealing with Bias and Fairness



## Field Trial Design



## Deployment



## Monitoring

# Structure of the class

- Module 1: End-to-end ML Pipeline
  - Formulation, Modeling Setup, Features, Models, Model Selection
- Module 2: Model Interpretability
- Module 3: Fairness

# Class Schedule

| Week | Dates      | Topic  | Required Readings                           | Assignments   |
|------|------------|--|---|---|
| 1    | Tu: Aug 31 | <a href="#">Class Intro and Overview</a>           |   |   |
| 1    | Th: Sep 2  | <a href="#">ML Project Scoping</a>                 | <a href="#">ML Project Scoping Guide</a>    | Project Team Selection  |
| 2    | Tu: Sep 7  | <a href="#">Getting, Storing, and Linking Data</a> | <a href="#">Optional readings on github</a> |   |
| 2    | Th: Sep 9  | <a href="#">Analytical Formulation / Baselines</a> | <a href="#">List on github</a>              |   |
| 3    | Tu: Sep 14 | Model Selection Methodology                        |   | Project Assignment 1: Formulation and Baseline (due Monday)   |
| 3    | Th: Sep 16 | Performance Metrics                                |   |   |
| 4    | Tu: Sep 21 | Feature Engineering and Imputation                 |   | Project Assignment 2:<br>Validation set up<br>Initial pipeline with train and validation set(s) and baseline implemented (due Monday) |
| 4    | Th: Sep 23 | Hands-on Session for ML Pipeline review            |   |   |
| 5    | Tu: Sep 28 | Models/hyperparameters in practice                 |   | Project Assignment 3:<br>list of features and some subset implemented (due Monday)  |
| 5    | Th: Sep 30 | Temporal Model Selection                           |   |   |
| 6    | Tu: Oct 5  | Module 1 Review: Applied ML - End to End Pipelines |   | Project Assignment 4:<br>modeling results (due Monday)  |
| 6    | Th: Oct 7  | Mid-term week - no class                           |   | Mid-term exam due Friday  |
| 7    | Tu: Oct 12 | Interpretability: Intro and Overview, taxonomy     |   |   |
| 7    | Th: Oct 14 | No Class - Mid-semester break                      |   |   |

# Class Schedule

| Week | Dates            | Topic   | Required Readings | Assignments                            |
|------|------------------|---|-------------------|--|
| 8    | Tu: Oct 19       | Understanding the Models  |                   |  |
| 8    | Th: Oct 21       | Interpretability Methods: Inherently Interpretable (GA2Ms, RiskSLIM, etc.)    |                   |  |
| 9    | Tu: Oct 26       | Interpretability Methods:: Post-Hoc Local/Feature-based (LIME, SHAP, MAPLE)   |                   |  |
| 9    | Th: Oct 28       | Interpretability Methods: Other methods (counterfactual, example-based, etc.) |                   |  |
| 10   | Tu: Nov 2        | Module 2 Review: ML Interpretability  |                   |  |
| 10   | Th: Nov 4        | ML Ethics Issues Overview   |                   | Interpretability Writeup Due on Friday |
| 11   | Tu: Nov 9        | Fairness in ML Overview   |                   |  |
| 11   | Th: Nov 11       | Fairness Methods: Pre-processing (removing sensitive attribute, sampling)     |                   |  |
| 12   | Tu: Nov 16       | Fairness Methods: In-processing (Zafar, Celis, fairlearn, etc.)               |                   |  |
| 12   | Th: Nov 18       | Post-Processing: Hardt, LA, etc   |                   |  |
| 13   | Tu: Nov 23       | Module 3 Review: ML Fairness  |                   |  |
| 13   | Th: Thanksgiving | Thanksgiving holiday  |                   |  |
| 14   | Tu: Nov 30       | Field Trials and Causality  |                   | Bias Writeup Due                       |
| 14   | Th: Dec 2        | Wrap-Up   |                   |  |
| 15   | Tu: Dec 7        | No Class - Finals Week  |                   |  |
| 15   | Th: Dec 9        | No Class - Finals Week  |                   | Final Research Writeup Due             |

# Logistics

- Attendance (is not optional)
- Platforms:
  - Latest content will be on **github**
  - **Canvas** (for assignment submissions)
  - **Slack (and email)** for communications and project and teamwork
- Wednesday sessions
- Office hours
- TAs: to help manage and help access AWS infrastructure

# Project Teams

- Make sure to fill out the survey: We need your github username and public SSH key
- You should create 5-person teams by the end of this week (and fill out the spreadsheet to let us know)

# Project



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# Tech Setup Options

- Get set up on github for your project
- Compute Infrastructure
  - Use your own laptop/machine for running things locally
  - Or use AWS resources we have set up to run larger jobs
- Data Infrastructure
  - Use CSVs on your own laptop/machine
  - **Or use Postgres database we have set up with data loaded**
- If you use AWS, make sure you have the following things set up:
  - ssh (to connect to the server) **server.mlinpractice.dssg.io**
  - dbeaver and psql (to connect to the database) **server.mlinpractice.dssg.io**
- Nice to have: get familiar with
  - Postgresql (to analyze and query data)
  - \*nix command line
  - Remote server workflow

# Prep for next class

- Reading
  - Project Scoping guide
- Assignments
  - Survey (if you haven't already): link from email
  - Project Team Selections: link in canvas
- No Wednesday Session This Week