

# Announcements

- Last day of classes, you've almost made it!
- Final exam is **6-8pm on December 22**
- Today is the last day of office hours
- We will still respond on Ed Discussion

# Lecture 27: Review Part 2

CIS 4190/5190

Fall 2022

# Final Exam Tentative Format

- Similar length/format to the practice exam
  - $\approx$  15 questions in increasing difficulty
  - Fewer questions but more parts per question
  - Will require **less memorization** than the practice exam
  - But no cheat sheet
- Make sure you know the written homework well!
  - Also questions from practice final exam that we cover today

# Final Exam Format

- We will provide any **complicated** equations necessary
  - **You do not need to know:** Entropy, sigmoid function, logistic NLL, neural network model families
  - **You should know:** Linear regression model family, decision tree model family
  - **You should know:** How to compute a derivative, probability identities, etc.
  - **You should know:** K-means clustering algorithm structure, gradient descent algorithm structure, AdaBoost structure (but not the detailed formulas)
- You should also know how different design choices/hyperparameters affect performance of each algorithm
  - E.g.,  $k$  in kNN,  $\lambda$  in linear/logistic regression, feature dimension  $d$ , number of examples  $n$ , AdaBoost iterations  $T$ , random forest base models  $k$ , etc.

# Incomplete List of Potential Topics

- **Models/algorithms**

- What is the model family? How does its decision boundary look?
- What is the loss function? How does it compare to the “true” loss (e.g., NLL vs. accuracy for logistic regression)?
- What is the optimizer? Is it guaranteed to find the global optimizer?

# Incomplete List of Potential Topics

- **Models/algorithms**

- Linear/logistic regression
- KNNs
- Decision trees
- Random forests, gradient boosted decision trees
- Feedforward neural networks, convolutional neural networks
- K-means clustering
- PCA
- Bayesian networks
- Q iteration, Q learning, epsilon-greedy exploration
- Collaborative filtering

# Incomplete List of Potential Topics

- **Concepts**

- Supervised vs. unsupervised vs. reinforcement learning
- Loss minimization framework
- Maximum likelihood framework
- Bias-variance tradeoff
- Regularization
- Exploration in reinforcement learning
- ML ethics

Good Luck!!!