

WEEK 2

CLASS 3: MACHINE LEARNING, K-NEAREST NAIGHBORS (KNN)

- Iris dataset [[What does an iris look like?](#)]
- [[Data](#)] hosted by the UCI Machine Learning Repository

Machine learning and KNN

- [Reddit AMA with Yann LeCun](#)
- [Characteristics of your zip code](#)
- Introduction to scikit-learn (Classroom Notebook)
- Documentation: [[user guide](#)] [[module reference](#)] [[class documentation](#)]

Homework:

- Complete your visualization homework assigned in class 2
- Reading assignment on the Pre-work Bias-Variance tradeoff

Optional:

- For a more in-depth introduction to machine learning, read section 2.1 (14 pages) of Hastie and Tibshirani's excellent book, [[An Introduction to Statistical Learning](#)]. (It's a free PDF download!)
- For a really nice comparison of supervised versus unsupervised learning, plus an introduction to reinforcement learning, watch this [[video](#)] (13 minutes) from Caltech's [[Learning From Data](#)] course.

Resources:

- Quora has a [[data science topic FAQ](#)] with lots of interesting Q&A.
- [Five most popular similarity measures](#) – Euclidean, Manhattan, Minkowski, Cosine Similarity, and Jaccard Similarity
- [Intuitive Explanation of Bias Variance](#) – discussion on Quora

CLASS 4: BIAS-VARIANCE TRADEOFF AND MODEL EVALUATION

- Exploring the bias-variance tradeoff (Classroom Notebook)
- Discussion of the [assigned reading] on the bias-variance tradeoff
- Model evaluation procedures (Classroom Notebook)

Homework:

- Complete the glass_id problem.

Resources:

- If you would like to learn the IPython Notebook, the official [[Notebook tutorials](#)] are useful.
- To get started with Seaborn for visualization, the official website has a series of [[tutorials](#)] and an [[example gallery](#)].
- Hastie and Tibshirani have an excellent [[video](#)] (12 minutes, starting at 2:34) that covers training error versus testing error, the bias-variance tradeoff, and train/test split (which they call the "validation set approach").
- Caltech's Learning from Data course includes a fantastic [[video](#)] (15 minutes) that may help you to visualize bias and variance.