

# **Logistic Regression Revisited**

**CS/DSA 5970: Machine Learning Practice**

# Logistic Regression Review

- Add a sigmoid non-linearity to the end of our linear model
- Sigmoid: output range from 0 to 1
  - Can interpret this as a probability
  - For classification, this can be the probability of being in the positive class
- Prior classification conversation:
  - Used the MSE cost function (mean squared differences between ground truth label and the probability)
  - Problematic because the derivative can become very flat



# Parameter Selection for Likelihood Functions

From statistics:

- Given:
  - A set of samples drawn independently from a distribution
  - A form of distribution from which the samples are drawn (e.g., a Normal distribution)
- Find the “best” parameters that explain the set of samples
  - Typical approach: use a likelihood function



# Log-Likelihood Cost Function

- We can use a similar approach to talking about the “goodness” of a classifier
- The new twist: we now have two classes
  - The classifier should assign a high probability to the positive examples
  - And low probabilities to the negative examples



# Example: Logistic Regression

- SGDClassifier with 'log' loss:
  - Logistic regression with log likelihood loss (we already played with this class)
- LogisticRegression class:
  - Also uses log likelihood loss
  - Different solver than SGDClassifier



# Example: Logistic Regression

Both offer regularization

- L1, L2, Elastic (must pick solver appropriately)
- SGDClassifier with 'log' loss:
  - Regularization parameter: alpha
  - Increase value: more regularization
- LogisticRegression class:
  - Regularization parameter: C
  - Increase value: less regularization



# **Multiclass Case: Softmax**

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# Softmax

Want to be able to handle  $K > 2$  classes

- So far, the approach has been to create a set of binary classifiers and have them vote
- One vs all: need  $O(K)$  classifiers
- One vs one: need  $O(K^2)$  classifiers

# Softmax

Approach:

- Learned function: output a score for each of  $K$  classes
- Use the softmax function to translate the scores into probabilities
- Output:
  - Can look at the probabilities directly
  - Or can pick the class with the highest probability as the predicted class



# Example: Softmax

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# Example: Softmax

LogisticRegression class:

- Desired output can be an integer, with values encoding different classes
- Internally, the class performs one-hot encoding



