# SIGNLL Meeting 2

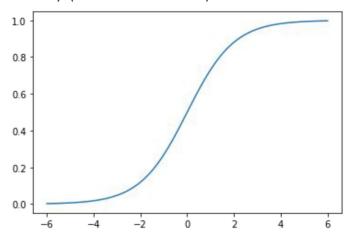
# Agenda (9/18)

- Logistic Regression
- Discuss/finalize project ideas

# What is Logistic Regression?

- GPT/Naive Bayes are generative
- Logistic Regression is discriminative
- For an input x, apply a weight and a bias -> wx + b
- Apply the sigmoid function to this to get a probability (between 0 and 1)
- Sigmoid function:

$$\sigma(z) = \frac{1}{1 + \exp(-z)}$$



### What about multiple inputs?

Suppose you have a vector input **x**. Then instead of multiplying it by a scalar weight w, you multiply it by a vector of weights **W** (dot product). Add the bias to the dot product as normal.

$$z = \mathbf{w} \cdot \mathbf{x} + b$$

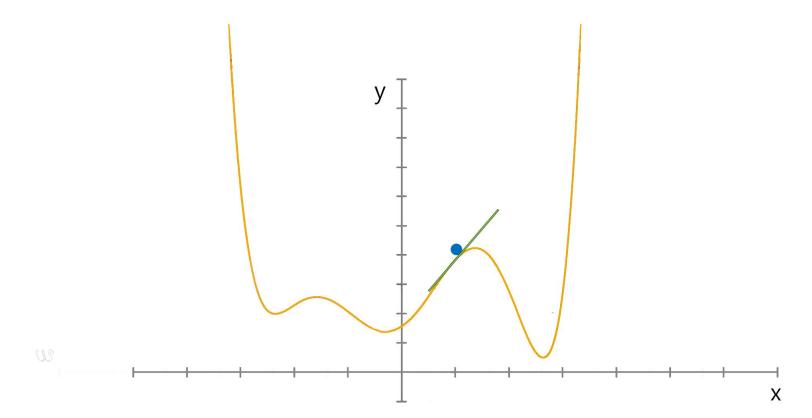
- Apply the sigmoid function to this value
- Make a decision boundary
  - E.g. If the value is > 0.5, it predicts one class. If not, it predicts another.

#### **How It Is Used**

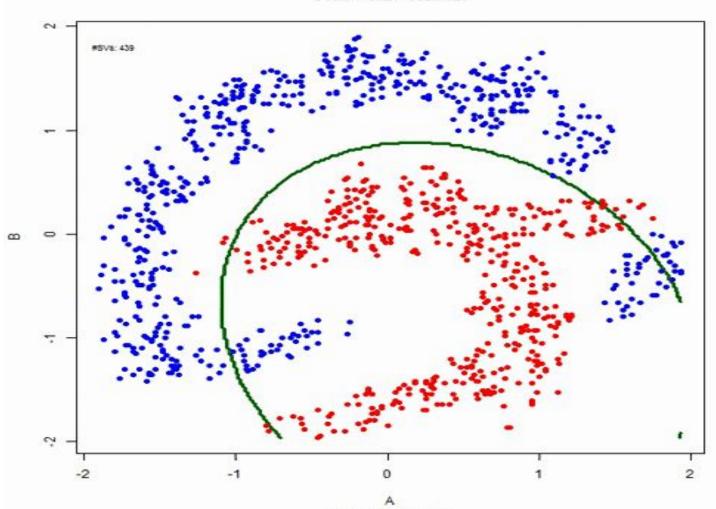
- Fast and Uncomplicated Classifier
  - Mainly used for simple classification problems
- Similar to Linear Regression except CATEGORICAL
  - I.e. yes/no, true/false, etc.

#### Introducing Gradient Descent

- Cost: How far is the actual value from the predicted value?
- Cost Function: A function that calculates the cost of our model
- Gradient Descent: A way to minimize the cost function on training data

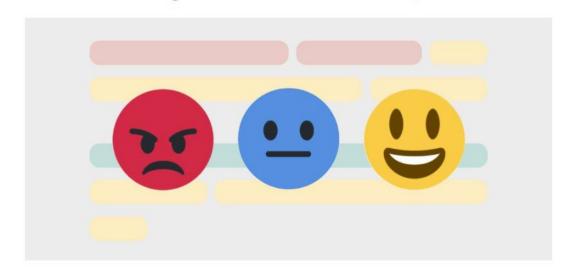


#### SVM + RBF Kernel



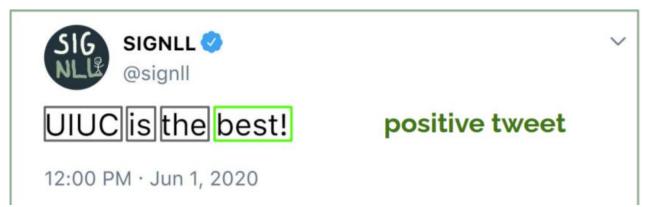
# **Sentiment Analysis**

Sentiment analysis is the interpretation and classification of emotions (positive, negative and neutral) within text data using text analysis techniques



# **Classifying Words**



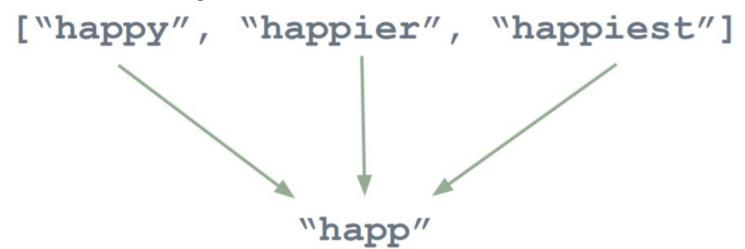


#### **Sentiment Analysis Steps**

- 1) Process and clean up our tweets
- 2) Organize them into a dictionary
- 3) Create and train a model using logistic regression
- 4) Verify our results with test data

#### **Word Tokenization**

- How do we deal with multiple word definitions?
- Fortunately, we can run a built-in algorithm to make sure that these words all mean the same thing



#### **Stopwords**

There are also some words that add no meaning to the sentence, such as:

["the", "a", "an", "I", ...]

Since these words don't help us that much, we can remove them from our tweets

### **Dictionary Conversion**

Instead of having separate columns for positive tweet count and negative tweet count, we can represent them as a single list of length 2:

$$word: \{n_{neg} \ n_{pos}\}$$

### Representing our tweets as numbers

To use logistic regression, we need to represent our tweets as numbers

We can do this by using three parameters from our dictionary:

$$\{1 \sum count_{neg} \sum count_{pos}\}$$



#### Tokens: ["wish", "friend", "like"]

```
"wish": [63, 29]

"friend": [30, 40]

"like": [182, 187]
```

tweet val = [1, 275, 256]

# Logistic Gradient Descent

$$y = \begin{cases} 1 & neg_0 & pos_0 \\ 1 & neg_1 & pos_1 \\ \dots & \dots & \dots \\ 1 & neg_{m-1} & pos_{m-1} \end{cases} \cdot \{\Theta_0 \Theta_1 \Theta_2\}$$
$$y = sigmoid(y)$$

#### GitHub!

- Step 1:
  - Install Python (<a href="https://www.python.org/downloads/">https://www.python.org/downloads/</a>) and pip
  - Run the following command: pip install notebook
- Step 2:
  - Clone this semester's GitHub repository to your machine
    - https://github.com/SIGNLL-UIUC/SIGNLL-Fall-2022
  - Navigate to the "Workshops" folder and run "jupyter notebook"
    - Select the .ipynb file to edit
  - Happy logistical regressioning!

#### **Brainstorm!**

- Text summarizer
- Calendar appointment generator
- Topic autocomplete feature