

Language

- Language is inherently hard to deal with
 - Speech Acts
 - Conversational Implicature
 - Shared Knowledge
 - Variation/Indexicality
- NLP involved representing language for some end
 - Dialogue
 - Translation
 - Speech Recognition
 - Text Analysis
- Certain Views of NLP
 - Information Theoretic View
 - Involves encoding and then subsequently decoding language
 - Rational Speech View
 - Communication involved recursive reasoning
 - How can X choose words that make sense to Y
 - Pragmatic View
 - Meaning is co-constructed by the people involved and the context of the speech
 - Whorfian View
 - Structure of language influences thought

Text Classification

- Sentiment Analysis
 - Is the entire text positive or negative?
 - Difficult as it is the measure of a speaker's private state, which is unobservable
- Representation for SA
 - Only positive/negative words
 - Only words in isolation
 - Conjunctions of words
 - Higher-order linguistic structure (syntax)
- Independence
 - $P(A,B) = P(A) * P(B)$
- Data Likelihood
 - Gives a strategy for choosing the single best parameter
- Conditional Probability
 - $P(X=x | Y=y)$
 - In this scenario you can use Bayes Rule

- Bayes Rule
 - $P(X=x | Y=y) = (P(X=x) P(Y=y | X=x)) / (P(X=x) P(Y=y | X=x))$
 - Training such a classifier means working with a summation of the denominator

Text Classification Methods

- Naïve Bayes
 - There is flexibility about the probability distributions used in NB
 - Includes Multinomial, Bernoulli, Normal, Poisson
- Multinomial Naïve Bayes
 - Discrete Distribution for modeling count data (words counts, proportion out of a whole)
- Bernoulli Naïve Bayes
 - Binary Event (True or False)
 - One parameter: P (Probability of an event occurring)
 - In terms of speech analysis, can put breaks within the writing to point out that the statement is negative in whole, even if a singular part is not
- Generative vs. Discriminative Models
 - Generative models specify a joint distribution over the labels and the data, can use this to generate new data
 - Care about $P(Y|X)$ but need to model more
 - Discriminative models specify the conditional distribution of the label y given the data x, focused on how to discriminate between the classes
 - Only care about the probability of $P(Y|X)$
- Logistic Regression
 - This is a method to engage in a discriminative model
 - Does not assume that features are independent like Naïve Bayes does
- Features
 - Can encode your own understanding of domain understanding of the problem
- Conditional Likelihood
 - For the training data, want probability of true label y for each data point x to be high
 - Lets us pick parameters B that maximize the probability that the training data is true
- Feature Selection
 - We can threshold features by minimum count, but this would throw away information
 - We can take a probabilistic approach and encode a prior belief that all B should be 0, unless there is evidence otherwise
 - As a ***practical*** part of this, when calculating through the gradient, only loop through nonzero valued features

Neural Networks

- Discrete, high-dimensional representation of inputs
 - Non-linear interactions of inputs
 - Multiple layers to capture hierarchical structure
- Design of Neural Networks
 - Tremendous flexibility on design choices
 - Articulate model structure and use the chain rule to derive parameter updates
- Regularization
 - Increasing the number of parameters
 - Increases the possibility for overfitting to training data
- Convolutional Networks
 - Same operation is applied to different regions of the input

Hypothesis Testing (Similar to Data8, less specific)

- Measures confidence in what we can say about a null from a sample
- Decide at which point is a sample statistic unusual enough to reject the null hypothesis
- Decide on the level of significance of p
 - Testing evaluates whether the sample statistic falls in the region that rejects the null in the rejection region defined by p
- Many such tests rely on parametric assumptions
 - Permutation test and the bootstrap do not depend on these assumptions

Language Modelling

- Provide a way to quantify the likelihood of a sequence
 - Can use multi-class logistic regression for language modelling by treating the vocabulary as the output space