

# Recap

LING572

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# Outline

- Summary for the 1<sup>st</sup> three weeks
- Reading materials
- Math formulas
- Workload for ling572

# So far

- Unit #1 (2 weeks): simple classification algorithms
  - kNN
  - Decision tree
  - Naïve Bayes
- Other topics (1 week):
  - Information theory
  - 570 Recap
    - Introduction to classification task
    - Mallet
    - Probability
    - Condor submission
  - Hw1-Hw3, Reading 1

# Main steps for solving a classification task

- Reformulate the task into a learning problem
- Define features
- Prepare training and test data
- Select ML learners
- Implement the learner
- Run the learner
  - Tune parameters on the dev data
  - Error analysis
  - Conclusion

# Comparison of three learners

	kNN	Decision Tree	Naïve Bayes
Modeling	Vote by your neighbors	Vote by your groups	Choose the $c$ that $\max P(c x)$
Training	No	Build a tree	Learn $P(c)$ and $P(f c)$
Decoding	Find neighbors	Traverse the tree	Calculate $P(c)P(x c)$
Tuned parameters	$K$ Similarity func	Max depth Split function Thresholds	Delta for smoothing

# Implementation issue

- Take the log:
- Ignore some constants:

$$P(d_i|c) = P(|d_i|) |d_i|! \prod_{k=1}^{|V|} \frac{P(w_k|c)^{N_{ik}}}{N_{ik}!}$$

- Increase small numbers before dividing

$$\log P(x, c_1) \text{ is } -200, \log P(x, c_2) \text{ is } -201.$$

# Implementation issue (cont)

- Reformulate the formulas: e.g., entropy calc

$$\begin{aligned} P(d_i, c) &= P(c) \prod_{w_k \in d_i} P(w_k|c) \prod_{w_k \notin d_i} (1 - P(w_k|c)) \\ &= P(c) \prod_{w_k \in d_i} \frac{P(w_k|c)}{1 - P(w_k|c)} \prod_{w_k} (1 - P(w_k|c)) \end{aligned}$$

- Store the useful intermediate results

$$\prod_{w_k} (1 - P(w_k|c))$$

# Lessons learned

- Don't follow the formulas blindly.
  - Ex1: Multinomial NB

$$P(c) \prod_{k=1}^{|V|} P(w_k|c)^{N_{ik}}$$

- Ex2: cosine function for kNN

$$\cos(d_i, d_j) = \frac{\sum_k a_{i,k} a_{j,k}}{\sqrt{\sum_k a_{i,k}^2} \sqrt{\sum_k a_{j,k}^2}}$$



# Next

- Unit #2 (3 weeks): two more advanced classification algorithms
  - MaxEnt
  - CRF
- 1-1.5 weeks per topic
  - Main intuition, final formulas used for training and testing
  - Mathematical foundation
  - Implementation issues

# Reading material

# The purpose of having reading material

- Something to reply on besides the slides
- Reading before class could be beneficial
- Papers (not textbooks) could be the main source of information in the future

# Problems with the reading material

- The authors assume that you have known the algorithm already:
    - Little background info
    - Page limit:
    - Style:
  - The notation problem
- ➔ It could take a long time to understand everything

# Some tips

- Look at several papers and slides at the same time
    - Skim through the papers first to get the main idea
    - Go to class and understand the slides
    - Then go back to the papers (if you have time)
  - Do not try to understand all the detail in the paper. Focus on the main ideas.
- ➔ Try to keep the reading time to be  $\leq 5$  hrs/week.

# Math formulas

# The goal of ling572

- Understand ML algorithms
  - The core of the algorithms
  - Implementation: e.g., efficiency issues
- Learn how to use the algorithms:
  - Reformulate a task into a learning problem
  - Select features
  - Write pre- and post-processing modules

# Understanding ML methods

- 1: have never heard about it
- 2: know very little
- 3: know the basics
- 4: understand the algorithm on paper
- 5: have implemented the algorithm
- 6: know how to modify/extend the algorithm

➔ Our goal: kNN, DT, NB: 5

MaxEnt, SVM, TBL: 3-4

Math is important for 4-6, especially for 6.



# Why are math formulas hard?

- Notation, notation, notation.
  - Same meaning, different notations:  $f_k, w_k, t_k$
- Calculus, probability, statistics, optimization theory, linear programming, ...
- People often have typos in their formulas.
- A lot of formulas to digest in a short period of time.

# Some tips

- No need to memorize the formulas
- Determine how crucial the formulas are
  - e.g., “\*\*” is for reference only

$$P(d_i|c_j) = P(|d_i|)|d_i|! \prod_{t=1}^{|V|} \frac{P(w_t|c_j)^{N_{it}}}{N_{it}!}$$

$$classify(d_i) = \operatorname{argmax}_c P(c) \prod_{k=1}^{|V|} P(w_k|c)^{N_{ik}}$$

- It is normal if you do not understand it the 1<sup>st</sup>/2<sup>nd</sup> time around.

# Understanding a formula

$$P(w_t|c_j) = \frac{1 + \sum_{i=1}^{|D|} N_{it} P(c_j|d_i)}{|V| + \sum_{s=1}^{|V|} \sum_{i=1}^{|D|} N_{is} P(c_j|d_i)}$$

$$\begin{aligned} P(w_t|c_j) &= \frac{\sum_{i=1}^{|D|} N_{it} P(c_j|d_i)}{\sum_{s=1}^{|V|} \sum_{i=1}^{|D|} N_{is} P(c_j|d_i)} \\ &= \frac{\sum_{i=1}^{|D|} N_{it} P(c_j|d_i)}{Z(c_j)} \\ &= \frac{\sum_{d_i \in D(c_j)} N_{it}}{Z(c_j)} \end{aligned}$$