

Name: _____ NetID: _____

STATISTICS AND DATA SCIENCE 355 / 555

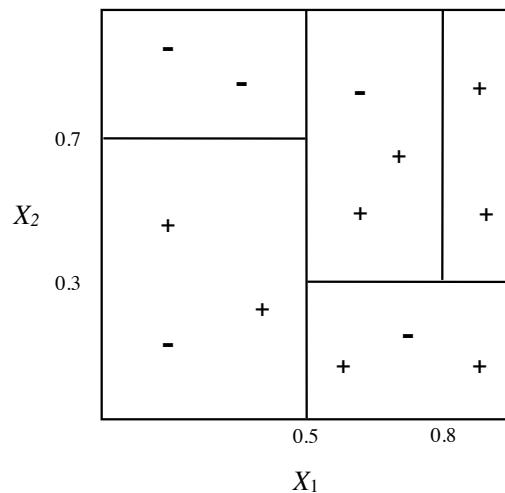
Introductory Machine Learning

Quiz 2 (practice), Thursday, October 31, 2019

1. *Decision trees* (5 points)

Consider the following figure showing 13 points in \mathbb{R}^2 and a partition of the unit square $[0, 1] \times [0, 1]$. Eight points are from class $Y = 1$ (labeled “+”) and five points are from class $Y = -1$ (labeled “-”).

draw tree below:



(a) Draw (to the right of the above figure) the decision tree that corresponds to the illustrated partition. Label the questions asked at each node.

(b) What is the training error (as a percentage) for these 13 data points?

$$100 \cdot \frac{3}{13} \%$$

(c) What is the predicted value of Y for the point $X = (X_1, X_2) = (.6, .2)$?

$$\hat{Y} = 1$$

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No notes or computers are allowed

1. *Bayesian inference* (5 points)

Suppose that X is a random variable denoting a coin flip, where $X = 1$ is “heads” with probability θ , and $X = 0$ is “tails” with probability $1 - \theta$. We want to carry out Bayesian inference on θ , using a $\text{Beta}(\alpha, \alpha)$ prior $p(\theta)$. Suppose that we flip the coin five times and observe X_1, X_2, \dots, X_5 , with three heads and two tails.

- (a) Give the formula for $p(\theta)$. You can state it up to a constant of proportionality.

$$p(\theta) = \frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2} \theta^{\alpha-1} (1-\theta)^{\alpha-1} \propto \theta^{\alpha-1} (1-\theta)^{\alpha-1}$$

- (b) Give an expression for the likelihood of the data given θ .

$$p(X | \theta) = \theta^3 (1-\theta)^2$$

- (c) What is the posterior distribution $p(\theta | X_1, \dots, X_5)$?

$$\text{Beta}(3 + \alpha, 2 + \alpha)$$

Posterior = likelihood * prior

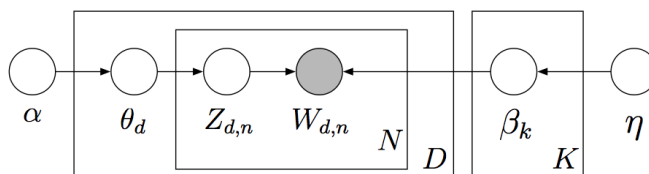
$$\theta^3 (1-\theta)^2 * \theta^{\alpha-1} (1-\theta)^{\alpha-1}$$

$$\theta^{3+\alpha} (1-\theta)^{2+\alpha}$$

$$\implies \text{Beta}(3 + \alpha, 2 + \alpha)$$

2. Topic modeling (5 points)

The latent Dirichlet allocation topic model is represented by the diagram



where $\theta_d \sim \text{Dirichlet}(\alpha)$ are the per-document topic proportions, $Z_{d,n} \sim \text{Multinomial}(\theta_d)$ are the per-word topic assignments, $W_{d,n} \sim \text{Multinomial}(\beta_{Z_{d,n}})$ are the observed words, and $\beta_k \sim \text{Dirichlet}(\eta)$ are the topics.

Circle the correct answers:

- ☒ TRUE ☐ FALSE (1) The model is generative, and can assign a probability to documents that are not in the training data.
- TRUE ☒ FALSE (2) According to the model, each document is generated by a single topic.
- TRUE ☒ FALSE (3) According to the model, the words are generated independently.
- ☒ TRUE ☐ FALSE (4) As α decreases from one toward zero, the topic proportions vector θ_d tends to have small values for a larger number of topics.
- TRUE ☒ FALSE (5) The Gibbs sampling algorithm chooses the most probable topic $Z_{d,n}$ for a selected word $W_{d,n}$ while holding all of the other Z values fixed.