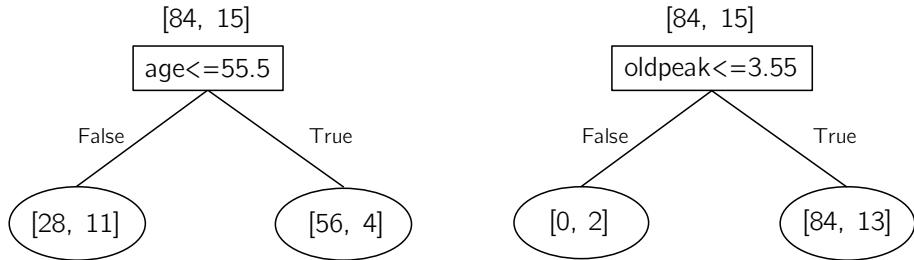


**Practice Midterm 2**

1. *Central Limit Theorem.* Going back to our class year example, say we expect the following probabilities of each class year: [0.125, 0.125, 0.25, 0.5] for [first-year, sophomore, junior, senior]. Let  $Y$  denote this random variable for year.
  - (a) If the class years are represented as the values [0,1,2,3] (respectively), what is the mean (expected value)  $E[Y]$  of this distribution?
  - (b) Compute the variance  $\text{Var}(Y)$  of this distribution.
  - (c) In reality we observe a class with  $n = 40$  students and sample mean  $\bar{Y}_n = 1.9$ . We wish to test the hypothesis that there are more first-years and sophomores in the class than we expected. First, use the CLT to compute the associated Z-score.
  - (d) The associated p-value is 0.08833 (double check after class). Sketch out the position of the test statistic on a standard normal distribution. Shade the area(s) representing the p-value. What do you conclude about your observed data?

2. *Entropy*. Consider the two feature choices below (for the heart disease dataset), and their associated splits. Counts of label -1 vs. 1 are shown in brackets.



- (a) After splitting the data based on each feature, what is the *classification error* for each tree (assuming that we are classifying based on the majority class)?
- (b) Before considering the feature, what is  $H(Y)$ , the entropy of the initial partition?
- (c) Which tree do you think produces more information gain?

3. *Logistic regression.* Say I train a logistic regression model and obtain the following weights  $\vec{w} = [1, 4]^T$ .
- Compute the decision boundary. Your answer should be an inequality describing when  $\hat{y} = 1$  (i.e. predict 1).
  - Sketch the model, labeling the decision boundary and graph axes.
  - How would you classify a new point  $x_{\text{test}} = -0.3$ ?
  - If the weight vector had instead been  $\vec{w} = [2.5, 10]^T$ , would the decision boundary change? Would the prediction change?
4. We are performing SGD to train a logistic regression model. We start with  $\vec{w} = [w_0, w_1]^T = [0, 0]^T$  and  $\alpha = 0.01$ . What are the new weights after analyzing data point  $(x, y) = (-3, 1)$ ?

5. *Bayesian probability.* For a specific disease, the incidence in the general population is  $\frac{1}{500}$ . Say I have a clinical test for this disease that comes back either positive or negative. Given a positive test result, there is an 80% chance the person has the disease. What is the *accuracy* of the test? In other words, compute the probability of a positive test result, given that the person has the disease. You may assume this value equals the probability of a negative test result, given the person is healthy.

6. *Naive Bayes.*

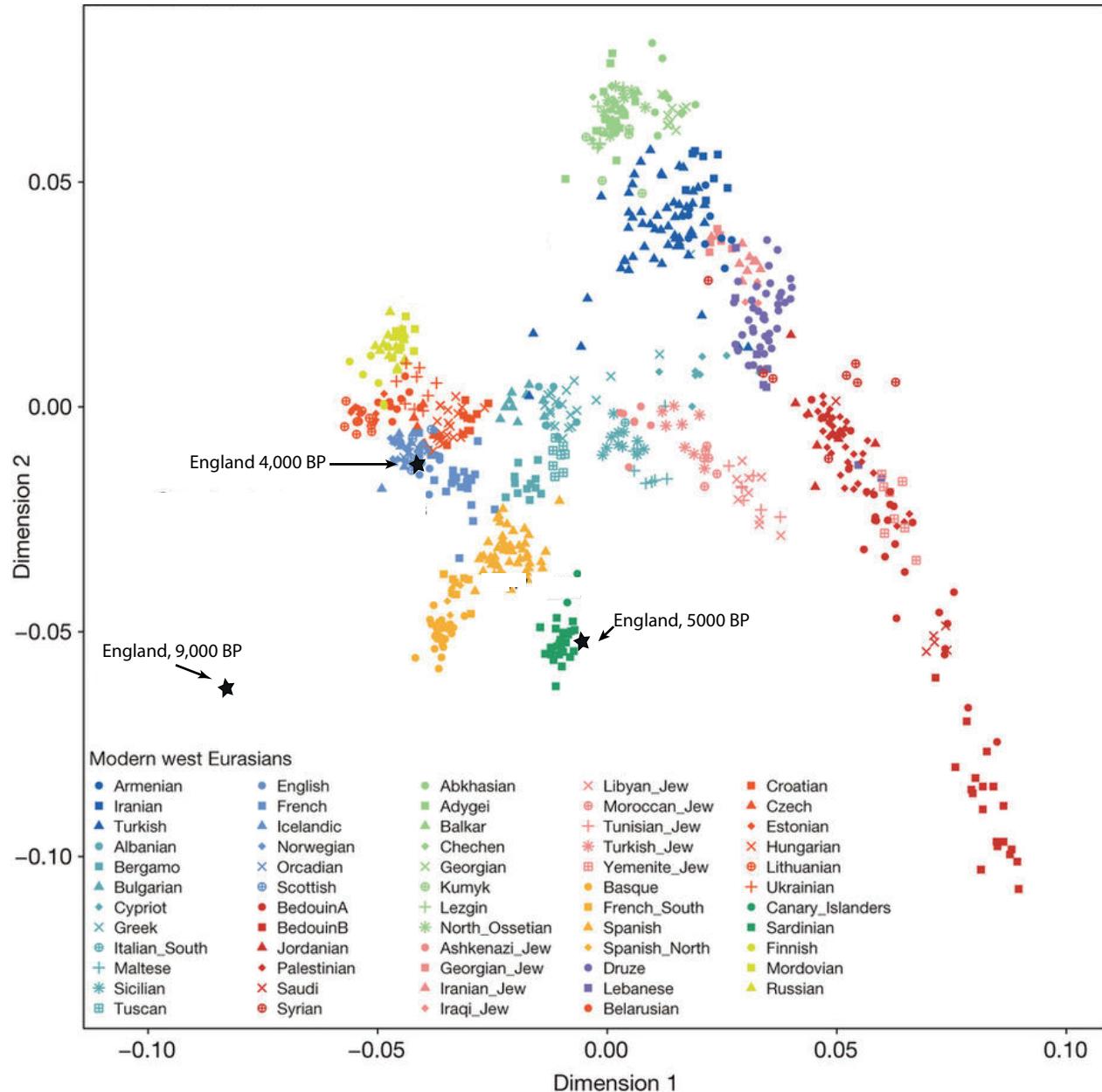
- (a) Write out a generic Bayesian model for  $p(y|\vec{x})$ , labeling the likelihood, prior, evidence, and posterior.
- (b) Now say that  $\vec{x} = [x_1, x_2, x_3]^T$  (i.e. 3 features). Rewrite the likelihood, applying the Naive Bayes assumption. Challenge question: redo the steps of the derivation in this small case.
- (c) Explain in words how we compute  $p(x_2 = v|y = k)$  (i.e. given  $y$  is some class label  $k$ , what is the probability feature  $x_2$  takes on the value  $v$ ).

7. Say we have the following training data with  $p = 2$  features. Feature  $f_1$  can take on three values  $\{1, 2, 3\}$  and  $f_2$  can take on five values  $\{A, B, C, D, E\}$ . Using Naive Bayes, which class  $y \in \{0, 1\}$  would you predict for the test example  $\vec{x} = [1, D]$ ? Show all work.

$\mathbf{x}$	$f_1$	$f_2$	$y$
$\mathbf{x}_1$	3	A	0
$\mathbf{x}_2$	2	B	1
$\mathbf{x}_3$	1	C	0
$\mathbf{x}_4$	2	E	0
$\mathbf{x}_5$	1	A	1

8. *Disparate impact.* Hypothetically, of the applicants for loans at a bank, 27.5% of the Black applicants got a loan compared to 35% for white applicants. Is there disparate impact in the bank's decisions? Explain your reasoning.

9. *PCA*. The figure below shows the first two PCs of genome-wide data from 777 present-day people from West Eurasia, along with three ancient British people who lived 9000, 5000 and 4000 years ago (labeled stars, “BP” means “[years] Before Present”).



What can you infer about the relationship between each of the ancient people and present-day Europeans? What does this figure suggest about the history of Britain, and the people living there, over the past ten thousand years?

*Acknowledgements: Iain Mathieson*