Midterm exam, 機器學習, Fall 2020. Open book but no calculators/cell phones allowed. Answers may include  $e^2$ ,  $\sqrt{2}$ , etc. but simplify when possible.

Exam written by Paul Horton ©2020.

Your Name: \_\_\_\_\_

# Problem 1.

Suppose

1. we have some observed data X (a set of real numbers  $\{x_1, ..., x_n\}$ 2. We assume the data are random samples generated by a normal distribution of unknown mean  $\mu$ .

Question 1a What is the maximum likelihood estimator for the  $\mu$ ?

Question 1b Given  $\mu = 0$ , what is the maximum likelihood estimator for  $\sigma^2$ ?

Give the mathematical derivation for your answers:

Your Name: \_\_\_\_\_

#### Problem 2.

Let v be a random variable defined by these values and probabilities.

| v | probability |
|---|-------------|
| 2 | 0.4         |
| 3 | 0.3         |
| 4 | 0.2         |
| 5 | 0.1         |

Let  $V(n)=v_1+v_2+\dots+v_n$  be the sum of n independent samples of v.

## Question 2

Derive the mean and standard deviation of V(n).

Your Name: \_\_\_\_\_

#### Problem 3.

The Poisson distribution has parameter  $\lambda\geqq 0$  , defining a probability distribution over the non-negative integers  $(0,1,\ldots)$  as follows:

$$\operatorname{Pois}(k;\lambda) \quad \stackrel{\text{\tiny def}}{=} \quad P[k] = \frac{\lambda^k}{k! \exp(\lambda)}, \quad k \in \mathbb{N}_0$$

This problem involves inference from data generated by one of two Poisson distributions:  $Pois(\lambda_1)$  or  $Pois(\lambda_2)$ . The following experiment is done.

1.  $\lambda$  is set to  $\{\lambda_1, \lambda_2\}$  with probability  $m_1$  and  $m_2 = 1 - m_1$ .

2. A random sample y is drawn from  $\operatorname{Pois}(\lambda)$ 

## Question 3a

What is the posterior probability  $\mathbf{P}[\lambda=\lambda_1|y=k]?$ 

**Question** 3b: What kind of prior is this? Is it conjugate? Why or why not? Your Name: \_

### Problem 4.

A standard poker deck has 52 cards. 13 each of:  $\bullet \bullet \bullet \bullet$ . The entropy of a single card drawn at random is  $\approx 5.7$  bits of information. You cannot see the card, but I can.

## Question 4a

If I told you the card is black (i.e. ' $\bullet$  ' or ' $\bullet$  '); how much entropy would remain? (give numerical answer and reason)

#### Question 4b

If I then told you the card was a spade '• ', how much entropy would remain then? (give numerical answer and reason)

## Question 4c

Two cards are drawn from a fresh deck of cards. Let  $S_1, S_2$  denote the first and second cards respectively. What is the mutual information  $I(S_1, S_2)$  (answer can include lg symbol). Your Name: \_

**Problem 5.** Consider a classification problem with two features  $F1 \in \{0, 1, 2\}$ ,  $F2 \in \{0, 1, 2, 3\}$ , Assume we know the two classes occur with equal probability: P[C = A] = P[C = B] = 0.5 (so you do not need to estimate P[C = A], just take it as given to be 0.5).

| Training Data |    | Te    | est Dat | $a \mid P[C=A]$ | P[C=A F1,F2]:P[C=B F1,F2] Using Prior: |          |         |  |
|---------------|----|-------|---------|-----------------|--|----------|---------|--|
| F1            | F2 | Class | FI      | F2              | MLE                                    | Jeffreys | Laplace |  |
| 2             | 3  | А     | 2       | 3               |  |          |         |  |
| 0             | 0  | А     | 1       | 1               |  |          |         |  |
| 2             | 2  | А     | 0       | 3               |  |          |         |  |
| 2             | 0  | А     | 1       | 0               |  |          |         |  |
| 2             | 0  | В     | 0       | 0               |  |          |         |  |
| 2             | 0  | В     | 1       | 3               |  |          |         |  |
| 1             | 0  | В     | 0       | 2               |  |          |         |  |
| 2             | 0  | В     | 0       | 1               | -+                                     |          |         |  |

The above table gives the P[F|C] probabilities for each feature and class.

## Question 5

Compute the probability a Naïve Bayes classifier would assign to P[C = A], using maximum estimation, Jeffrey's priors or Laplace priors respectively when estimating probabilities involving feature values. You may report the answer in terms of odds, so for example, if the  $P[C = A] = \frac{1}{3}$ , you can report that as 1:2 (hint: it is easier to work with odds).

You may find the following worksheet helpful.

|                          | Feature F1 |   |   | Feature F2 |   |   |   | Class |
|--------------------------|------------|---|---|------------|---|---|---|-------|
| Value                    | 0          | 1 | 2 | 0          | 1 | 2 | 3 |       |
|                          |            |   |   |            |   |   |   | A     |
| counts                   |            |   |   |            |   |   |   | B     |
| "counts"                 |            |   |   |            |   |   |   | A     |
| Jeffreys                 |            |   |   |            |   |   |   | B     |
| $2 \times \text{counts}$ |            |   |   |            |   |   |   | A     |
| Jeffreys                 |            |   |   |            |   |   |   | B     |
|                          |            |   |   |            |   |   |   | A     |
| Laplace                  |            |   |   |            |   |   |   | B     |





## Question 6

What parameter values in the table on the previous path match which plot in the table above? Fill in the "Label" column and use the "Comment" column and/or space at bottom to explain your answers.

| ID | Distribution   | Label (A–I) | Comment |
|----|--|-------------|---------|
| 1  | $0.8{\rm Beta}(9,9) + 0.2{\rm Beta}(9,0.1)$                        |             |         |
| 2  | Beta(0.5, 0.5)   |             |         |
| 3  | $0.5\operatorname{Beta}(3,6) + 0.5\operatorname{Beta}(6,3)$        |             |         |
| 4  | Beta(2,3)  |             |         |
| 5  | $0.5{\rm Beta}(20,20) + 0.5{\rm Beta}(20,1)$                       |             |         |
| 6  | $0.3{\rm Beta}(3,2) + 0.7{\rm Beta}(0.2,1)$                        |             |         |
| 7  | Beta(0.2,9)  |             |         |
| 8  | $0.5{\rm Beta}(0.5,5)+0.5{\rm Beta}(10,2)$                         |             |         |
| 9  | $0.8 \operatorname{Beta}(9,9) + 0.2 \operatorname{Beta}(0.2, 0.2)$ |             |         |