

## Problem Set - MLE

November 13, 2021

### 1 Maximum Likelihood Estimator

#### 1.1 Gaussian Distribution

Given a Gaussian distribution  $f(y, \mu, \sigma = 1) = \frac{1}{\sqrt{2\pi}} \exp(-\frac{(y-\mu)^2}{2})$  and a set of observations  $y_1, y_2, \dots, y_n$

1. Write down the likelihood function  $L(y_i, \mu)$  for the  $i$ th observation  $y_i$ . What is the total likelihood function for all  $n$  observations  $L(y_1, y_2, \dots, y_n, \mu)$ ?
2. What is the total log likelihood function  $\ell(y_1, y_2, \dots, y_n, \mu)$ ?
3. Prove that  $\ell$  is maximized when  $\mu = 1$ , given one observation  $y_1 = 1$ .
4. Prove that  $\ell$  is maximized when  $\mu = \sum_{i=1}^n \frac{y_i}{n}$ , given  $n$  observations  $y_1, y_2, \dots, y_n$ .

#### 1.2 Bernoulli Distribution

Given a Bernoulli distribution  $f(y, p) = p^y(1-p)^{1-y}$  and a set of observations  $y_1, y_2, \dots, y_n$ . Assume that  $k$  out of the  $n$  observations are 1, the rest are 0. We then have the total likelihood function  $L(y_1, y_2, \dots, y_n, p) = \prod_{i=1}^n p^{y_i}(1-p)^{1-y_i}$ .

1. What is the total log likelihood function  $\ell(y_1, y_2, \dots, y_n, p)$ ?
2. Prove that  $\ell$  is maximized when  $p = \frac{k}{n}$ .