

# Warmup: Maximum likelihood estimation

## Maximum likelihood estimation

Let  $Y_1, \dots, Y_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$ , with both  $\mu$  and  $\sigma^2$  unknown. Let  $\theta = \begin{bmatrix} \mu \\ \sigma^2 \end{bmatrix}$ . We would like to estimate  $\theta$  using the method of maximum likelihood. Recall that

$$f(Y_i|\theta) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{1}{2\sigma^2} (Y_i - \mu)^2 \right\}$$

1. Let  $\mathbf{y} = (Y_1, \dots, Y_n)$ . Find expressions for the likelihood  $L(\theta|\mathbf{y})$  and log likelihood  $\ell(\theta|\mathbf{y})$ .

2. By taking the partial derivatives  $\frac{\partial}{\partial \mu} \ell(\theta|\mathbf{y})$  and  $\frac{\partial}{\partial \sigma^2} \ell(\theta|\mathbf{y})$ , show that the MLEs for  $\mu$  and  $\sigma^2$  are

$$\hat{\mu} = \frac{1}{n} \sum_{i=1}^n Y_i$$

and

$$\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{\mu})^2 = \frac{1}{n} \sum_{i=1}^n (Y_i - \bar{Y})^2$$

(Don't worry about the second derivative for now, just set the partial derivatives equal to 0 and solve).