## Problem 46

### Overview

Chromatin folding is process in biology. We assume that chromatin polymers follow a random walk model. In this model, the two-dimensional distance between two chromatin fibers, R, follows a Raleigh distribution, with density

$$f(r|\theta) = \frac{r}{\theta^2} e^{\frac{-r^2}{2\theta^2}}$$

where  $r \geq 0$  and  $\theta > 0$ .

### **Estimating Theta**

#### Maximum Likelihood

#### **Derivation of the Estimator**

The likelihood function is

$$lik(\theta) = \frac{1}{\theta^{2n}} e^{-\frac{1}{2\theta^2} \sum_{i=1}^n r_i^2} \prod_{i=1}^n r_i$$

The log likelihood is thus

$$l(\theta) = -2n \ln(\theta) + \sum_{i=1}^{n} \ln(r_i) - \frac{1}{2\theta^2} \sum_{i=1}^{n} r_i^2$$

The first-order condition for maximizing the likelihood satisfies

$$0 = \frac{-2n}{\hat{\theta}} + \frac{1}{\hat{\theta}^3} (\sum_{i=1}^n r_i^2)$$

Which, after rearranging, gives the maximum likelihood estimator for  $\theta$ ,  $\hat{\theta}_{MLE}$ 

$$\hat{\theta}_{MLE} = \frac{1}{\sqrt{2}} \sqrt{\frac{1}{n} \sum_{i=1}^{n} r_i^2}$$

#### **Asymptotic Variance**

As usual, the asymptotic variance of a maximum likelihood estimator is roughly

$$Var(\hat{\theta}_{MLE}) \approx \frac{1}{nI(\theta)}$$

where under sufficient smoothness conditions that at the level of this textbook we will assume hold,

$$I(\theta) = E \left[ \frac{\partial}{\partial \theta} \log(f(x|\theta)) \right]^2 = -E \left[ \frac{\partial^2}{\partial \theta^2} \log(f(x|\theta)) \right]$$

$$f(r|\theta) = \frac{r}{\theta^2} e^{\frac{-r^2}{2\theta^2}}$$
, so

$$\begin{split} \frac{\partial}{\partial \theta} \log f(r|\theta) &= -\frac{2}{\theta} + \frac{r^2}{\theta^3}, \\ \frac{\partial^2}{\partial \theta^2} \log f(r|\theta) &= \frac{2}{\theta^2} - 3\frac{r^2}{\theta^4}, \\ I(\theta) &= -\frac{2}{\theta^2} + 3\frac{E(R^2)}{\theta^4} \end{split}$$

The expectation of  $\mathbb{R}^2$  is

$$E(R^2) = \frac{1}{\theta^2} \int_0^\infty r^3 e^{\frac{-r^2}{2\theta^2}} dr = 4\theta^2 \int_0^\infty x^3 e^{-x^2} dx = 2\theta^2 \int_0^\infty u e^{-u} du$$

by making the substitutions  $x = \frac{r}{\sqrt{2\theta}}$  and  $u = x^2$ . Integrating by parts,

$$E(R^{2}) = 2\theta^{2} \left[ -ue^{-u}|_{0}^{\infty} + \int_{0}^{\infty} e^{-u} du \right] = -2\theta^{2} (e^{-u})|_{0}^{\infty} = 2\theta^{2}$$

Thus

$$I(\theta) = \frac{4}{\theta^2},$$
$$Var(\hat{\theta}_{MLE}) \to \frac{\theta^2}{4n}$$

#### Method of Moments

#### **Derivation of the Estimator**

The expectation of R is

$$E(R) = \frac{1}{\theta^2} \int_0^\infty r^2 e^{\frac{-r^2}{2\theta^2}} dr = (2\sqrt{2})\theta \int_0^\infty x^2 e^{-x^2} dx$$

by making the substitution  $x = \frac{r}{\sqrt{2\theta}}$ . Integrating by parts,

$$E(R) = (2\sqrt{2})\theta \left[ \frac{1}{2}xe^{-x^2}|_0^\infty + \frac{1}{2}\int_0^\infty e^{-x^2}dx \right] = \sqrt{2}\theta \int_0^\infty e^{-x^2}dx = \theta \sqrt{\frac{\pi}{2}}$$

since  $\int_0^\infty e^{-x^2} dx = \sqrt{\pi}$ , and the integrand is an even function. Thus

$$\hat{\theta}_{MoM} = \bar{X}\sqrt{\frac{2}{\pi}}$$

#### Asymptotic Variance

By the Central Limit Theorem,  $\bar{X}$  converges to a normal random variable with mean E(R) and variance  $\frac{Var(R)}{n}$  as n approaches infinity. From previous results.

$$Var(R) = E(R^2) - E(R)^2 = 2\theta^2 - \frac{\pi}{2}\theta^2 = \frac{4-\pi}{2}\theta^2$$

and the asymptotic variance of the method of moments estimator for  $\theta$  is

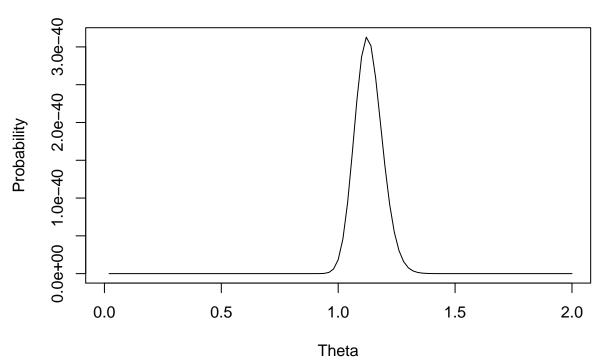
$$Var(\hat{\theta}_{MoM}) \to \left(\frac{4}{\pi} - 1\right) \frac{\theta^2}{n}$$

in which  $\frac{4}{\pi}-1\approx .273295...$ 

### Data Analysis

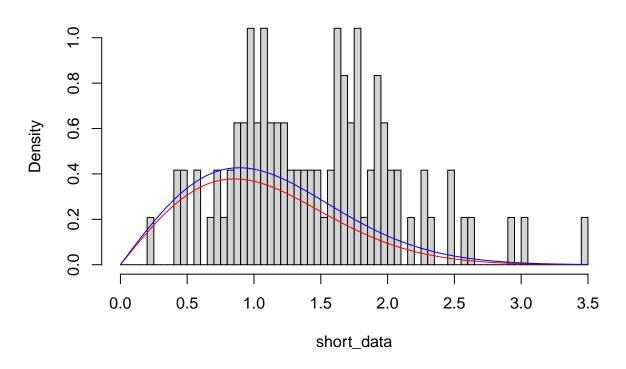
#### **Short Experiment**

## **Likelihood Function for the Short Experimental Data**



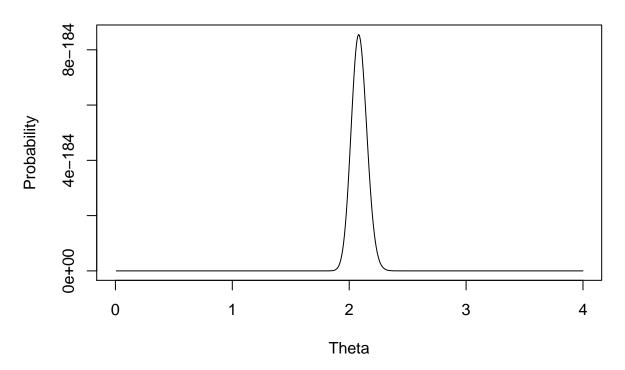
The MLE estimate is  $\hat{\theta}_{MLE}=1.1237776$ , with an estimated variance of 0.0029265. The MoM estimate is  $\hat{\theta}_{MoM}=1.1705631$ , with an estimated variance of 0.0039.

# Histogram of short\_data



## Medium Experiment

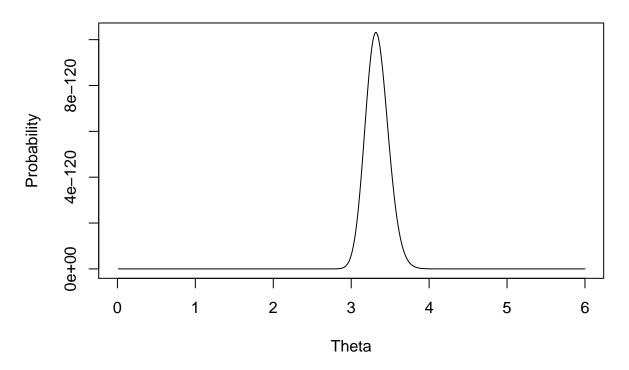
# **Likelihood Function for the Medium Experimental Data**



The MLE estimate is  $\hat{\theta}_{MLE}=2.0812133$ , with an estimated variance of 0.0020896. The MoM estimate is  $\hat{\theta}_{MoM}=2.0748203$ , with an estimated variance of 0.0047239.

## Long Experiment

# **Likelihood Function for the Long Experimental Data**



The MLE estimate is  $\hat{\theta}_{MLE}=3.3168463$ , with an estimated variance of 0.0062819. The MoM estimate is  $\hat{\theta}_{MoM}=3.4044888$ , with an estimated variance of 0.0239924.