

**Calculus Warm-Up Quiz — Day 7**

Maximum Likelihood Estimation and Normal Regression

*Complete before lecture. 10 minutes.***Name:** \_\_\_\_\_

Today we maximize likelihood functions. This requires the product rule, the derivative of  $\log x$ , and the chain rule applied to exponentials. We also compute second derivatives (the Hessian) to characterize curvature.

**1. Product Rule.** Compute the derivative.

$$(a) \frac{d}{dp} [p^3(1-p)^2] =$$

(b) Set your answer equal to zero and solve for  $p$ . (Hint: factor out  $p^2(1-p)$ .)

**2. Derivatives of  $\log$  and  $1/x$ .** Compute:

$$(a) \frac{d}{d\lambda} [-3 \log \lambda] =$$

$$(b) \frac{d}{d\lambda} \left[ -\frac{5}{\lambda} \right] =$$

$$(c) \frac{d}{d\lambda} \left[ -3 \log \lambda - \frac{5}{\lambda} \right] = \quad \text{Set this to zero and solve for } \lambda.$$

**3. Log-Likelihood of the Normal.**

The log-density of a  $N(\mu, \sigma^2)$  random variable  $x$  is:

$$\log f(x|\mu, \sigma^2) = -\frac{1}{2} \log(2\pi) - \frac{1}{2} \log(\sigma^2) - \frac{(x-\mu)^2}{2\sigma^2}$$

- (a) Treating  $\sigma^2$  as known, compute  $\frac{\partial}{\partial \mu} \log f(x|\mu, \sigma^2)$ .
- (b) For  $n$  independent observations, the log-likelihood is  $\ell(\mu) = \sum_{i=1}^n \log f(x_i|\mu, \sigma^2)$ . Set  $\frac{d\ell}{d\mu} = 0$  and solve for  $\hat{\mu}$ .

**4. Second Derivative (Hessian Preview).**

From Question 2(c), let  $\ell(\lambda) = -3 \log \lambda - 5/\lambda$ .

- (a) Compute  $\ell'(\lambda)$  (you already did this).
- (b) Compute  $\ell''(\lambda)$ . (Differentiate  $\ell'(\lambda)$  again.)
- (c) Evaluate  $\ell''(\lambda)$  at  $\hat{\lambda}$  from 2(c). Is it negative? (This confirms a maximum.)

**5. Chain Rule with Exponentials.** Compute:

(a)  $\frac{d}{dx} e^{-2x} =$

(b)  $\frac{d}{dx} e^{-x^2/2} =$

**Answer Key — Day 7**

- 1.** (a)  $3p^2(1-p)^2 - 2p^3(1-p) = p^2(1-p)[3(1-p) - 2p] = p^2(1-p)(3-5p)$   
 (b)  $p^2(1-p)(3-5p) = 0$ . Solutions:  $p = 0, p = 1, p = 3/5$ . The interior maximum is  $p = 3/5$ .
- 2.** (a)  $-3/\lambda$   
 (b)  $5/\lambda^2$   
 (c)  $-3/\lambda + 5/\lambda^2 = 0 \implies 5/\lambda^2 = 3/\lambda \implies \lambda = 5/3$ .
- 3.** (a)  $\frac{\partial}{\partial \mu} \log f = \frac{x - \mu}{\sigma^2}$   
 (b)  $\frac{d\ell}{d\mu} = \frac{1}{\sigma^2} \sum_{i=1}^n (x_i - \mu) = 0 \implies n\mu = \sum x_i \implies \hat{\mu} = \bar{x} = \frac{1}{n} \sum x_i$ .
- 4.** (a)  $\ell'(\lambda) = -3/\lambda + 5/\lambda^2$   
 (b)  $\ell''(\lambda) = 3/\lambda^2 - 10/\lambda^3$   
 (c) At  $\hat{\lambda} = 5/3$ :  $\ell''(5/3) = 3/(25/9) - 10/(125/27) = 27/25 - 270/125 = 27/25 - 54/25 = -27/25 < 0$ . Confirmed: maximum.
- 5.** (a)  $-2e^{-2x}$   
 (b)  $-x e^{-x^2/2}$