

ECON 4130

HG 2005 H

Sketch answers to the ordinary exam 2004 H

(Note that the answers given are just sketchy. The exam paper would normally require some more commenting or details.)

Problem 1

1a.

$$\int_0^1 x^\theta dx = \frac{1}{\theta+1}. \quad \text{Graph}$$

1b.

$$F(x) = x^5,$$

$$(i) P(0,5 \leq X \leq 0,8) = 0,8^5 - 0,5^5 = 0,328 - 0,031 = 0,297,$$

$$(ii) P(X > 0,8) = 1 - 0,8^5 = 1 - 0,328 = 0,672,$$

$$(iii) \text{Median: } m^5 = 0,5 \Rightarrow m = 0,5^{1/5} = 0,871$$

1c.

$$E(X^r) = \frac{\theta+1}{\theta+1+r} \Rightarrow E(X) = \frac{\theta+1}{\theta+2} \quad \text{and} \quad E(X^2) = \frac{\theta+1}{\theta+3} \quad \text{from which}$$
$$\text{var}(X) = \frac{\theta+1}{\theta+3} - \left(\frac{\theta+1}{\theta+2}\right)^2 = \frac{(\theta+1)[(\theta+2)^2 - (\theta+1)(\theta+3)]}{(\theta+3)(\theta+2)^2} = \frac{\theta+1}{(\theta+3)(\theta+2)^2}$$

1d.

$0 < X \leq 1 \Rightarrow 0 \leq Y < \infty$. For $y > 0$ we have

$P(Y \leq y) = P(-\ln(X) \leq y) = P(\ln(X) \geq -y) = P(X \geq e^{-y}) = e^{-y(\theta+1)}$, which is the cdf of $\exp(\theta+1)$.

Problem 2

2a.

$$(1 - 0,8^{\theta+1} \geq 0,75) \Leftrightarrow (0,8^{\theta+1} \leq 0,25) \Leftrightarrow ((\theta+1)\ln(0,8) \leq \ln(0,25))$$

$$\Leftrightarrow \left(\theta \geq \frac{\ln(0,25)}{\ln(0,8)} - 1 = 5,21 \right)$$

2b.

MME: $\frac{\tilde{\theta}+1}{\tilde{\theta}+2} = \bar{X}$ gives $\tilde{\theta} = \frac{2\bar{X}-1}{1-\bar{X}} = 6,4204$ where $\bar{X} = 0,88124$

MLE: Log likelihood: $l(\theta) = n\ln(\theta+1) + \theta \sum_i \ln(x_i)$ with

derivative $l'(\theta) = \frac{n}{\theta+1} + \sum_i \ln(x_i)$ giving $\hat{\theta} = -\frac{n}{\sum_i \ln(X_i)} - 1 = 6,3180$

2c.

The (weak) law of large numbers. $Y_i = \ln(X_i)$ are *iid* with constant expectation, $-1/(\theta+1)$ and variance in **1c**. The probability limit follows then by Tshebysjev's inequality. Since $\hat{\theta} = g(\bar{Y})$ is continuous, we have

$$\text{plim}(\hat{\theta}) = g(\text{plim } \bar{Y}) = g(-1/(\theta+1)) = -(-(\theta+1)) - 1 = \theta \quad ,$$

2d.

The mgf of $Y_i = -\ln(X_i)$ is according to **1d** the mgf of a $\Gamma(1, \theta+1)$ distribution, i.e. $M_Y(t) = \frac{\theta+1}{\theta+1-t}$. Because of the independence, the mgf of $V = \sum Y_i$ is

$$M_V(t) = \left(\frac{\theta+1}{\theta+1-t} \right)^n \text{ which is the mgf of } \Gamma(n, \theta+1).$$

2e.

$$\mathbb{E}\left(\frac{1}{V}\right) = \frac{\lambda^\alpha}{\Gamma(\alpha)} \int_0^\infty v^{\alpha-2} e^{-\lambda v} dv = \lambda \frac{\Gamma(\alpha-1)}{\Gamma(\alpha)} \frac{\lambda^{\alpha-1}}{\Gamma(\alpha-1)} \int_0^\infty v^{\alpha-2} e^{-\lambda v} dv = \frac{\lambda}{\alpha-1}$$

We find $\mathbb{E}(\hat{\theta}) = \mathbb{E}\left(\frac{n}{V} - 1\right) = n \frac{\theta+1}{n-1} - 1 = \frac{n}{n-1} \theta + \frac{1}{n-1}$ which $\rightarrow \theta$ as $n \rightarrow \infty$.
