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Course: CS 6364, 002

Homework 3

①: MLE - questions #1

$$\text{Likelihood } L(\theta) = \prod_{i=1}^N \theta x_i^{\theta-1}$$

Taking natural logarithm

$$\ln L(\theta) = \ln \left(\sum_{i=1}^N \theta x_i^{\theta-1} \right)$$

$$= \sum_{i=1}^N \ln(\theta) + (\theta-1) \sum_{i=1}^N \ln(x_i)$$

$$\frac{dL}{d\theta} = \frac{N}{\theta} + \sum_{i=1}^N \ln(x_i) = C$$

$$\Rightarrow \frac{\partial}{N} = - \left(\sum_{i=1}^N \ln(x_i) \right)^{-1}$$

$$= - [\ln(0.4) + \ln(0.7) + \ln(0.9) + \ln(0.6) \\ + \ln(0.5) + \ln(0.7)]^{-1} \times 6$$

$$= 0.34025 \times 6 = 2.0415$$

② MLE-questions #2

Given uniform distribution

$$0 \leq x_i \leq \theta$$

$$\Rightarrow f(x_i) = \frac{1}{\theta-0} = \frac{1}{\theta}$$

Log likelihood

$$L(\theta) = \prod_{i=1}^N \frac{1}{\theta} = \theta^{-N}$$

Taking natural logarithm

$$\ln L(\theta) = -n \ln(\theta)$$

$$\frac{dL}{d\theta} = -n \cdot \frac{1}{\theta} < 0 \text{ given } \theta < 0$$

Hence $L(\theta)$ is decreasing and it's maximized at

$$\theta = x_N$$

that x_N is the largest value on the interval $(0, \theta)$

$$\boxed{\theta = x_N}$$

③ MLE - questions #3

$$f(x) = \frac{1}{\beta} e^{-\frac{x}{\beta}} \quad \text{with } 0 < x < \infty$$

Log likelihood

$$L(\beta) = \prod_{i=1}^N \frac{1}{\beta} e^{-\frac{x_i}{\beta}}$$

Taking natural logarithm

$$\ln L(\beta) = \sum_{i=1}^N \ln \left(\frac{1}{\beta} \cdot e^{-\frac{x_i}{\beta}} \right)$$

$$= \sum_{i=1}^N \ln \left(\frac{1}{\beta} \right) + \sum_{i=1}^N \ln e^{-x_i/\beta}$$

$$= N \ln \frac{1}{\beta} - \frac{1}{\beta} \sum_{i=1}^N x_i$$

$$\frac{dL}{d\beta} = N \cdot \beta + \frac{1}{\beta^2} \sum_{i=1}^N x_i = 0$$

$$\Rightarrow \frac{1}{\beta^2} \left(N \cdot \beta^3 + \sum_{i=1}^N x_i \right) = 0$$

$$\text{Since } \frac{1}{\beta^2} > 0$$

$$\Rightarrow N \cdot \beta^3 + \sum_{i=1}^N x_i = 0$$

$$\Rightarrow \beta^3 = -\frac{1}{N} \sum_{i=1}^N x_i$$

$$\boxed{\Rightarrow \beta = \sqrt[3]{-\frac{1}{N} \sum_{i=1}^N x_i}}$$

④ MLE - questions # 4

Since $0 < x < \infty$

$$\Rightarrow r(d) = (d-1)!$$

Then

$$f(x) = \frac{1}{(\lambda-1)!} x^\lambda x^{\lambda-1} e^{-\lambda x}$$

Log likelihood

$$L(\lambda, x_i) = \prod_{i=1}^N \frac{1}{(d-1)!} \lambda^d x_i^{d-1} e^{-\lambda x_i}$$

Taking natural logarithm

$$\ln L(\lambda, x_i) =$$

$$\sum_{i=1}^N \ln \frac{1}{(d-1)!} + \sum_{i=1}^N \ln \lambda^d + \sum_{i=1}^N \ln(x_i^{d-1}) + \sum_{i=1}^N \ln(e^{-\lambda x_i})$$

$$= -N \ln(d-1)! + N \ln \lambda^d + (d-1) \sum_{i=1}^N \ln x_i - \lambda \sum_{i=1}^N x_i$$

Taking 1^{st} derivative

$$\frac{dL}{d\lambda} = 0 + N \cdot \frac{1}{\lambda^d} d(\lambda^{d-1}) + 0 - \lambda \sum_{i=1}^N x_i = 0$$

$$= \frac{Nd}{\lambda} - \lambda = 0$$

$$\Rightarrow \frac{Nd - \lambda^2}{\lambda} = 0$$

$$\Rightarrow \lambda = \frac{N\bar{x}}{\lambda} \quad \text{①} \quad \text{that } \lambda \neq 0$$

Taking 2nd derivative

$$\frac{dL}{d\lambda} = \frac{N\bar{x} - \lambda}{\lambda} \Rightarrow \frac{d^2L}{d\lambda^2} = -\frac{N\bar{x}}{\lambda^2} - 1$$

$$\text{From ① } \Rightarrow \lambda^2 = N\bar{x} \Rightarrow 1 = \frac{\lambda^2}{N} \text{ that } N > 0$$

$$\Rightarrow \lambda > 0$$

$$\text{hence } \frac{d^2L}{d\lambda^2} = -\frac{N\bar{x}}{\lambda^2} - 1 < 0 \text{ giving} \\ \lambda = \pm \sqrt{N\bar{x}}$$

\Rightarrow MLE

$$\boxed{\lambda = \pm \sqrt{N\bar{x}} \text{ that } \lambda > 0}$$

(5) MLE-questions #5

$$f(x) = \frac{1}{2} e^{-|x-\theta|}$$

Log likelihood

$$L(\theta) = \prod_{i=1}^N \frac{1}{2} e^{-|x_i - \theta|}$$

$$\begin{aligned}\ln L(\theta) &= \sum_{i=1}^N \ln \frac{1}{2} + \sum_{i=1}^N \ln e^{-|x_i - \theta|} \\ &= -N \ln 2 - \sum_{i=1}^N |x_i - \theta|\end{aligned}$$

Taking derivative

$$\begin{aligned}\frac{dL}{d\theta} &= 0 - \sum_{i=1}^N (|x_i - \theta|)' = - \sum_{i=1}^N \left(\sqrt{(x_i - \theta)^2} \right)' \\ &= - \sum_{i=1}^N \frac{1}{2} \frac{2 \cdot (x_i - \theta)}{\sqrt{(x_i - \theta)^2}} (-1) \\ &= \sum_{i=1}^N \frac{(x_i - \theta)}{|x_i - \theta|} = 0\end{aligned}$$

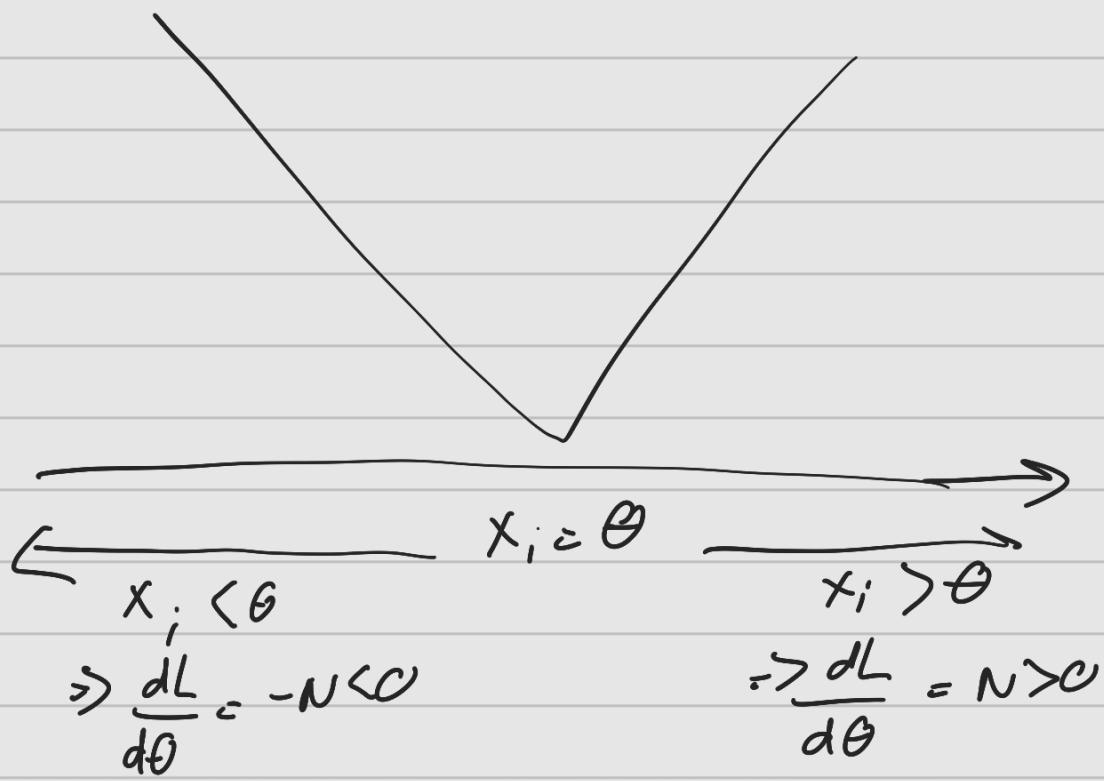
If $x_i < \theta$

$$\Rightarrow \frac{dL}{d\theta} = \sum_{i=1}^N \frac{(x_i - \theta)}{-|x_i - \theta|} = - \sum_{i=1}^N 1 = -N < 0$$

If $x_i > \theta$

$$\Rightarrow \frac{dL}{d\theta} = \sum_{i=1}^N \frac{(x_i - \theta)}{|x_i - \theta|} = \sum_{i=1}^N 1 = N > 0$$

We can plot & take $x_i = \theta$ as non-existing critical point



Taking 2nd derivative

$$\frac{dL}{d\theta} = \sum_{i=1}^N \frac{x_i - \theta}{|x_i - \theta|} \Rightarrow \frac{d^2L}{d\theta^2} = \sum_{i=1}^N \frac{-1|x_i - \theta| - \frac{(x_i - \theta) \cdot (x_i - \theta)}{|x_i - \theta|}}{(x_i - \theta)^2}$$

$$= \sum_{i=1}^N \frac{-1|x_i - \theta| - \frac{(x_i - \theta)^2}{|x_i - \theta|}}{(x_i - \theta)^2 |x_i - \theta|} < 0 \text{ given}$$

$x_i \neq \theta$

hence $x_i \neq \theta$

\Rightarrow MLE of θ
 when $\theta \neq x$

⑥ PRML # 2.2 page 127

Distribution is normalized if sum of all results = 1

When

$$x = -1$$

$$\Rightarrow \sum_{\mu} p(-1|u) = \left(\frac{1-\mu}{2}\right)^{(1+1)/2} \cdot \left(\frac{1+\mu}{2}\right)^{(1-1)/2}$$

$$= \frac{1-\mu}{2}$$

When $x = 1$

$$\cdot \left(\frac{1-\mu}{2}\right)^{(1-1)/2} \cdot \left(\frac{1+\mu}{2}\right)^{(1+1)/2}$$

$$\Rightarrow \sum_m p(l, m) = \left(\frac{1-m}{2}\right)^l \cdot \left(\frac{1+m}{2}\right)$$

$$= \frac{1+m}{2}$$

$$\Rightarrow \sum_{x \in M} \sum p(x|m) = \left(\frac{1+m}{2}\right) + \left(\frac{1-m}{2}\right)$$

$$= \frac{1+m+m-m}{2} = 1$$

\Rightarrow $p(x|m)$ is normalized

Hence,

~~$$\text{Avg } E(x) = \sum_{x \in \{-1, 1\}} x \cdot p(x|m)$$~~

$$= (-1)p(x=-1|m) + 1 \cdot p(x=1|m)$$

$$= -\left(\frac{1-m}{2}\right) + \frac{1+m}{2}$$

$$= -\frac{1}{2} + \frac{m}{2} + \frac{1}{2} + \frac{m}{2} = m$$

$$\Rightarrow \text{Avg } E(x) = m$$

~~Var~~ Variancia

$$\text{Var}(p) = \sum_{x \in \{-1, 1\}} (x - E(p))^2 p(x|n)$$

$$= \sum_{x \in \{-1\}} (x - n)^2 p(-1|n)$$

$$= (-1 - n)^2 p(-1|n)$$

$$+ (1 - n)^2 p(1|n)$$

$$= (1 + n)^2 \left(\frac{-1-n}{2}\right) + (1-n)^2 \left(\frac{1+n}{2}\right)$$

$$= (1 + 2n + n^2) \left(\frac{-1-n}{2}\right)$$

$$+ (1 - 2n + n^2) \left(\frac{1+n}{2}\right)$$

$$= (1 + n^2) \left(\frac{-1-n}{2} + \frac{1+n}{2}\right) + 2n \left(\frac{-1-n - 1+n}{2}\right)$$

$$= (1 + n^2) + 2n \left(\frac{-2n}{2}\right) = 1 + n^2 - 2n^2 = 1 - n^2$$

⇒ $\text{Var}(x) = 1 - n^2$

$$\begin{aligned}
 H[x] &= \sum_{x \in \{-1, 1\}} p(x|u) \ln p(x|u) \\
 &= p(-1|u) \ln p(-1|u) + p(1|u) \ln p(1|u) \\
 &= \left(\frac{1-u}{2}\right) \ln \left(\frac{1-u}{2}\right) + \left(\frac{1+u}{2}\right) \ln \left(\frac{1+u}{2}\right) \\
 &= \ln \left[\left(\frac{1-u}{2}\right)^{(1-u)/2} \cdot \left(\frac{1+u}{2}\right)^{(1+u)/2} \right]
 \end{aligned}$$

③ PRML # 2.4 page 168

$$\text{Set } p(m|u) = \sum_{m=0}^N \binom{N}{m} u^m (1-u)^{N-m} = 1 \quad ①$$

$$\begin{aligned}
 \frac{dp}{du} &= \sum_{m=0}^N \binom{N}{m} \cdot \left[m \cdot u^{m-1} (1-u)^{N-m} \right. \\
 &\quad \left. + u^m \cdot (N-m) (1-u)^{N-m-1} (-1) \right] = 0
 \end{aligned}$$

$$\begin{aligned}
 &= \sum_{m=0}^N \binom{N}{m} u^{m-1} (1-u)^{N-m-1} [m(1-u) - m(N-m)]
 \end{aligned}$$

$$\begin{aligned}
 &= \sum_{m=0}^N \binom{N}{m} u^{m-1} (1-u)^{N-m-1} \cdot (m - muN) = 0 \quad ②
 \end{aligned}$$

$$\Rightarrow m = N\mu \quad (3)$$

Hence,

$$E(m) = \sum_{m=0}^N m \cdot \text{Bin}(m/N, \mu)$$

$$= m \sum_{m=0}^N \binom{N}{m} \mu^m (1-\mu)^{N-m}$$

From (1) and (3), $E(m) = m \cdot 1 = N\mu$

From (2)

$$\frac{dp}{dm} = \sum_{m=0}^N m^{m-1} (1-\mu)^{N-m-1} (m - N\mu)$$

$$\frac{d^2 p}{d\mu^2} = \sum_{m=0}^N [(m-1) \mu^{m-2} (1-\mu)^{N-m-1} (m - N\mu)]$$

$$+ \mu^{m-1} \left((N-m-1) (1-\mu)^{N-m-2} (-1) (m - N\mu) \right. \\ \left. - N (1-\mu)^{N-m-1} \right]$$

$$= \sum_{m=0}^N \left[\mu^{m-2} (1-\mu)^{N-m-1} (m-1) (m - N\mu) \right. \\ \left. - \mu^{m-1} (1-\mu)^{N-m-2} (N-m-1) (m - N\mu) \right]$$

$$= N (1-\mu)^{N-m-1} \mu^{m-1}]$$

$$= \sum_{m=0}^N m^{m-2} (1-\mu)^{N-m-2}.$$

$$\left[(1-\mu)(m-1)(m-N\mu) - \mu(N-m-1)(m-N\mu) - N(1-\mu)\mu \right] = 0$$

$$\Rightarrow (1-\mu)(m^2 - mN\mu - m + N\mu - N\mu) = 0$$

$$- \mu(N-m-1)(m-N\mu) = 0$$

$$\Rightarrow (1-\mu)(m^2 - mN\mu - m) - \mu(N-m-1)(m-N\mu) = 0$$

$$\Rightarrow (1-\mu)m(m-N\mu) - (1-\mu)m \cdot$$

$$- \mu(N-m-1)(m-N\mu) = 0$$

$$\Rightarrow (m-N\mu)(m-N\mu + \mu) - (1-\mu)m = 0$$

$$- (1-\mu)m = 0$$

$$\Rightarrow (m-N\mu)(m-N\mu + \mu) - (1-\mu)m = 0$$

$$\Rightarrow (m-N\mu)^2 + \mu m - N\mu^2 - m + \mu m = 0$$

$$\Rightarrow (m-N\mu)^2 = N\mu^2 + m - 2\mu m$$

From ③ $m = N\mu$

and $E(m) = N\mu$

$$\Rightarrow (m - E(m))^2 = N\mu^2 + N\mu - 2N\mu^2 \\ = N\mu - N\mu^2 \\ = N\mu(1 - \mu)$$

Hence,

$\text{Var}[m]$

$$= \sum_{m=0}^N (m - E[m])^2 \underbrace{\text{Bin}(m | N\mu)}_{\sim 1}$$

$$= (m - E[m])^2 = N\mu(1 - \mu)$$

⑧ PRML # 2.6 page 128

Beta dist.

$$\text{Beta}(\mu | a, b) = \frac{r(a+b)}{r(a)r(b)} \mu^{a-1} (1-\mu)^{b-1} \quad (2.13)$$

$$r(x) \text{ is gamma function } \int_0^\infty u^{x-1} e^{-u} du \quad (1.14)$$

$$\int_0^1 \text{Beta}(\mu | a, b) d\mu = 1 \quad (2.14) \text{ normalized}$$

From (2.265),

$$\int_0^1 u^{a-1} (1-u)^{b-1} du = \frac{r(a)r(b)}{r(a+b)}$$

$$E[u] = \int_0^1 u \text{Beta}(u|a,b)$$

$$= \int_0^1 u \frac{r(a+b)}{r(a)r(b)} \cdot u^{a-1} (1-u)^{b-1} du$$

$$= \frac{r(a+b)}{r(a)r(b)} \underbrace{\int_0^1 u^a \cdot (1-u)^{b-1} du}_{(1)}$$

From (1)

$$\int_0^1 u^{(a-1)+1} \cdot (1-u)^{b-1} du = \frac{r(a+1)r(b)}{r(a+b+1)}$$

$$\Rightarrow E[u] = \frac{r(a+b)}{r(a)r(b)} \cdot \frac{r(a+1)r(b)}{r(a+b+1)} = \frac{a}{a+b}$$

using identity $r(t+1) = tr(t)$

$$\text{Var}[u] = \frac{r(a+b)}{r(a)r(b)} \int_0^{\infty} u^{a+2} (1-u)^{b-1} du - \left(\frac{a}{a+b}\right)^2$$

$$= \frac{r(a+b)}{r(a)r(b)} \cdot \frac{r(a+2) r(b)}{r(a+b+2)} - \left(\frac{a}{a+b}\right)^2$$

$$= \frac{a}{a+b} \cdot \frac{(a+2)}{(a+b+1)} - \left(\frac{a}{a+b}\right)^2$$

$$\Rightarrow \text{Var}[u] = \frac{a(a+1)}{(a+b)(a+b+1)} - \frac{a^2}{(a+b)^2}$$

$$= \frac{(a^2+a)(a+b)}{(a+b)^2(a+b+1)}$$

$$(a+b)^2(a+b+1)$$

$$= \frac{a^3 + ba^2 + a^2 + ab - a^3 - a^2b - a^2}{(a+b)^2(a+b+1)}$$

$$= \boxed{\frac{ab}{(a+b)^2 (a+b+1)}} = V_{2r}[\mu]$$

$$\frac{d}{d\mu} \text{Beta}(\mu|a,b) = \frac{d}{d\mu} \cdot \mu^{a-1} \cdot (1-\mu)^{b-1}$$

$$= (a-1) \mu^{a-2} (1-\mu)^{b-1} - (b-1) (1-\mu)^{b-2} \mu^{a-1} = 0$$

$$\Rightarrow \mu^{a-2} (1-\mu)^{b-2} \left[(a-1)(1-\mu) \right.$$

$$\left. - (b-1)\mu \right] = 0$$

$$\Rightarrow (a-1) - \mu(a+b-2) = 0$$

$$\boxed{\mu = \frac{a-1}{a+b-2} = \text{mod}}$$

(3) PRML # 2.41 page 134

Def. gamma func 1.141 \rightarrow gam
2. 1ce6 normalized

(2.146)

$$\text{Gamma}(\lambda | a, b) = \frac{1}{r(a)} b^a \times \lambda^{a-1} \exp(-b\lambda)$$

Distr

$$\text{Gamma fun } r(x) = \int_0^{\infty} u^{x-1} e^{-u} du$$

Gamma Distribution is normalized

$$\int_0^{\infty} \text{Gamma}(\lambda | a, b) = 1$$

$$= \int_0^t \frac{1}{r(a)} b^a \times^{a-1} e^{-b} = L$$

$$\Rightarrow \int_0^t b^a \times^{a-1} e^{-b} = r(a)$$

$$r(a) = \int_0^\infty x^{a-1} e^{-x} dx$$

Let $x = ub$ since $x > 0 \Rightarrow 0 < u < 1$
 $\Rightarrow dx = bdu$

$$\begin{aligned} \Rightarrow r(a) &= \int_0^1 u^{a-1} \cdot b^{a-1} e^{-ub} \cdot b du \\ &= \int_0^1 b^a u^{a-1} e^{-bu} du \end{aligned}$$

\Rightarrow Gamma Distr. is normalized

$$E(\lambda) = \int_0^1 \lambda \frac{1}{r(a)} b^a \lambda^{a-1} e^{-b\lambda} d\lambda$$

$$= \frac{1}{r(a)} \int_0^1 \lambda b^a \lambda^{a-1} e^{-b\lambda} d\lambda$$

$$= \frac{1}{r(a)} \frac{1}{b} \left[\int_0^1 b^{a+1} \lambda^a e^{-b\lambda} d\lambda \right]$$

①

$$\textcircled{1} = r(a+1)$$

$$\Rightarrow E(\lambda) = \frac{r(a+1)}{r(a)} \cdot \frac{1}{b} = \frac{a}{b}$$

$$\text{Var}(\lambda) = \int_0^\infty \frac{b^a}{r(a)} \lambda^{a+1} e^{-b\lambda} - E(\lambda)^2$$

$$= \frac{b^a}{r(a)} \int_0^\infty \lambda^{a+1} e^{-b\lambda} d\lambda - \left(\frac{a}{b}\right)^2$$

$$\text{Let } u = b\lambda \Rightarrow du = b d\lambda \\ \Rightarrow b = u/\lambda \quad \lambda = u/b$$

$$\Rightarrow \text{Var}(u)$$

$$= \frac{b^a}{r(a)} \int_0^\infty \left(\frac{u}{b}\right)^{a+1} e^{-u} b du - \left(\frac{a}{b}\right)^2$$

$$= \frac{b^a}{\frac{b^{a+2}}{r(a)}} \int_0^\infty u^{a+1} e^{-u} du - \left(\frac{a}{b}\right)^2$$

$$= \frac{1}{b^2 r(a)} \int_0^\infty u^{a+1} e^{-u} du - \left(\frac{a}{b}\right)^2$$

$$= \frac{1}{b^2 r(a+2)} - \left(\frac{a}{b}\right)^2$$

$$= \frac{r(a+2) - a^2 r(a)}{b^2 r(a)}$$

$$= \frac{a(a+1)r(a) - a^2 r(a)}{b^2 r(a)}$$

$$= \frac{a^2 + a - a^2}{b^2} = \boxed{\frac{a}{b^2} = \text{Var}(E)}$$

$$\frac{d}{d\lambda} \text{Gamma}(\lambda | a, b) = \frac{d}{d\lambda} \left(b^a \lambda^{a-1} e^{-b\lambda} \right) = 0$$

$$= (a-1) \lambda^{a-2} e^{-b\lambda} - b e^{-b\lambda} \lambda^{a-1} = 0$$

$$= \lambda^{a-2} e^{-b\lambda} (a-1 - b\lambda) = 0$$

$$\Rightarrow \boxed{\lambda = \frac{a-1}{b}}$$

