

Lecture 5

Tutorial on plotting
data

ML estimation

Examples

Poisson

Tied parameters

Lecture Notes II – Maximum Likelihood Estimation for Discrete Distributions

Marina Meilă
`mmp@stat.washington.edu`

Department of Statistics
University of Washington

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Max Likelihood Principle



ML estimation for arbitrary discrete distributions



Other ML estimation examples



ML estimate as a random variable

Reading: Ch. 4.1, 4.2

$$m = 2$$

UW Student speaks Chinese = $X \in \{0, 1\}$

$$n = 10$$

$$n_1 = 3$$

$$n_0 = 7$$

$$\theta_1^{ML} = \frac{3}{10} = \underline{0.3}$$

$$\theta_0^{ML} = 0.7$$

estimate
My "guess" Model
Pr[a random
UW student
speaks Chinese]

Ex 2 $m = 4$

S = Species of trees on UW campus

= { Hisakura Cherry, Douglas Fir, Maple,
W. Hemlock }

x = species of tree

true $m^* \gg 4$

Extra credit!

ML estimation for arbitrary discrete distributions

$$n = 20$$

C, M, H, D

$$n_C = 10$$

$$\Rightarrow \theta_C^{ML} = 0.5$$

$$n_H = 1$$

$$\theta_H^{ML} = 0.05$$

$$n_D = 2$$

$$\theta_D^{ML} = 0.1$$

$$n_M = 7$$

$$\theta_M^{ML} = 0.35$$

S = Species of trees on UW campus
 = { Hisakura Cherry, Douglas Fir, Maple, W. Hemlock }
 x = species of tree

$$P = (\theta_C, \theta_M, \theta_H, \theta_D)$$

Model class

Estimates

Model

3. Starting car gas ignition OTHER

S = { 0, 1, 2, 3, 4, 5 }

car starts

battery

alternator

nominal

error "codes"

$$\theta_0^{ML} = 0.5$$

$$\theta_1^{ML} = 0.3 \dots$$

$$n = 20$$

$$n_0 = 10$$

$$n_1 = 6$$

$$n_2 = 1 = n_3 = \dots$$

ML estimation for arbitrary discrete distributions

$$S = \{0, 1, \dots, m-1\} \quad \text{but some } n_j = 0$$

$$S = \{0, 1, 2, \dots\}$$

$$|S| = \infty, \quad n \text{ finite} \Rightarrow \text{most } n_j = 0$$

Idea 1 Smoothing

$$n_j = 0 \Rightarrow \theta_j^{\text{ML}} = 0 \rightarrow \text{choose } \tilde{\theta}_j > 0 \quad \text{NOT ML}$$

UNACCEPTABLE!

⇒ Idea 2 Get more data if possible

→ Idea 3 Choose model class with ~~finite~~ few parameters

{ Poisson, Geometric
Tie parameters

Other ML estimation examples

Poisson

$$S = \{0, 1, 2, \dots\}$$

$$x \in S$$

Queuing systems

$x = \# \text{ queries / time unit}$

$$P(x) = \frac{\lambda^x}{x!} e^{-\lambda} \quad \lambda > 0$$

$$\sum_{x=0}^{\infty} \frac{\lambda^x}{x!} = e^{\lambda} = Z \quad \text{normalization constant}$$

$$n = 10$$

$$x^{1:10} = 3, 5, 3, 4, 0, 1, 1, 12, 2, 2 = \mathcal{D} \quad s = 35$$

ML Estimation for λ

1. Likelihood

$$P(\mathcal{D} | \lambda) \equiv L(\lambda) = \prod_{i=1}^n P(x^i | \lambda) =$$

$$= \prod_{i=1}^n \left(\frac{\lambda^{x^i}}{x^i!} \cdot e^{-\lambda} \right) = \frac{\lambda^{\sum_{i=1}^n x^i}}{\prod_{i=1}^n (x^i!)} \cdot e^{-n\lambda}$$

$$1. \log \cdot l. \quad \ln L(\lambda) = \underbrace{\ln \lambda \cdot \sum_{i=1}^n x^i}_{\text{data}} - \underbrace{n\lambda}_{\text{STAT}} - \sum_{i=1}^n \ln(x^i!)$$

STAT
↑
CMLC
↓

Other ML estimation examples Poisson (cont)

Max:

$$\ln L(\lambda) = \ln \lambda \cdot \underbrace{\sum_{i=1}^n x_i}_{\text{data}} - \underbrace{n\lambda}_{\text{indep of } \lambda} + \text{constant} = \ell(\lambda) \quad \lambda \in (0, \infty)$$

$$\ell'(\lambda) = 0 = \frac{1}{\lambda} \cdot \underbrace{\sum_{i=1}^n x_i}_S - n + 0 \Rightarrow \frac{S}{\lambda} = n \Rightarrow \lambda = \frac{S}{n}$$

$$\lambda^{\text{ML}} = \frac{1}{n} \sum_{i=1}^n x_i$$

ML estimator

average $\bar{x}$

$$S = 35 \Rightarrow \lambda^{\text{ML}} = 3.5$$

$$n = 10$$

$$S = \sum_{i=1}^n x_i$$

SUFFICIENT STATISTIC

Other ML estimation examples Model with tied parameters

$S = \{0, 1, \dots, 6\}$ car starting?

$$n = 20$$

$$n_0 = 10$$

$$n_1 = 6$$

$$n_2 = 0$$

$$n_3 = 2$$

$$n_4 = n_5 = 1$$

$$n_6 = 0$$

Model $\mathcal{P} = (\theta_0, \theta_1, \theta_*)$ 3 params.?
class

same Probability for $x = 2, 3, 4, 5, 6$
 $\theta_2 = \theta_3 = \dots = \theta_6 = \theta_*$
Tied params.

1. Likelihood

$$L(\theta_0, \theta_1, \theta_*) = \theta_0^{10} \theta_1^6 \theta_*^{\underbrace{n_2 + n_3 + n_4 + n_5 + n_6}_{\text{suff stat}} = 4}$$

$$n_0 = 10$$

$$n_1 = 6$$

$$n_* = 4$$

1'. Log-L

$$l(\theta_0, \theta_1, \theta_*) = n_0 \ln \theta_0 + n_1 \ln \theta_1 + n_* \ln \theta_*$$

STAT
↑
calc

Other ML estimation examples

$$\text{Max}_{\theta_0, \theta_1, \theta_*} \ell(\theta_{0,1,*}) = \underline{n_0} \ln \underline{\theta_0} + \underline{n_1} \ln \underline{\theta_1} + \underline{n_*} \ln \underline{\theta_*}$$

$$\theta_{0,1,*} \in [0, 1]$$

$$\text{s.t. } \theta_0 + \theta_1 + \theta_* \cdot \underline{5} = 1$$

Lagrange multiplier

$$f = n_0 \ln \theta_0 + n_1 \ln \theta_1 + n_* \ln \theta_* - \lambda (\theta_0 + \theta_1 + 5\theta_* - 1)$$

$$\frac{\partial f}{\partial \theta_0} = \frac{n_0}{\theta_0} - \lambda = 0 \Rightarrow \theta_0 = \frac{n_0}{\lambda}$$

$$\theta_1 = \frac{n_1}{\lambda}$$

$$\frac{\partial f}{\partial \theta_1} = \frac{n_1}{\theta_1} - \lambda = 0$$

$$\theta_* = \frac{n_*}{5\lambda}$$

$$\frac{\partial f}{\partial \theta_*} = \frac{n_*}{\theta_*} - \lambda \cdot 5 = 0$$

$$\begin{aligned} \theta_0^{ML} &= \frac{n_0}{n} & \theta_1^{ML} &= \frac{n_1}{n} \\ \theta_*^{ML} &= \frac{n_*/5}{n} \end{aligned}$$

$$\frac{\partial f}{\partial \lambda} = * = \frac{n_0}{\lambda} + \frac{n_1}{\lambda} + 5 \frac{n_*}{5\lambda} = 1 \Rightarrow$$

$$\lambda = n_0 + n_1 + n_* = n$$

Other ML estimation examples

How much more data?

$$S = \{0, 1\} \quad \theta_0 \gg \theta_1$$

$$\text{w.p. } 1 - \delta \quad \delta = 10^{-3}$$

Q $n = ?$ so that $n_1 \geq 1$ w.h.p. with high probability

Q' n known

θ_1 known

$$P[n_1 = 0] = ?$$

$$P[n_0 = n] = \theta_0^n = (1 - \theta_1)^n = \delta$$

solve for n

$$n \ln(1 - \theta_1) = \ln \delta \Rightarrow n^* = \frac{\ln \delta}{\ln(1 - \theta_1)} = \frac{-\ln \frac{1}{\delta}}{\theta_1}$$

$$\ln(1 + z) \approx z$$

$$\ln(1 - \theta_1) \approx -\theta_1$$

$$0. \quad n > n^* \Rightarrow P[n_0 = n] \leq \delta$$

$$1. \quad n^* \propto \frac{1}{\theta_1}$$

$$2. \quad n^* \propto \ln \frac{1}{\delta}$$

intuitive

↑ when larger

$1 - \delta$ confidence wanted

~~Other ML estimation examples~~ What next?

Idea 1 $\eta_j = 0$ what to do? $\leftarrow$ $\approx$ next week

ML estimator as r.variable $\leftarrow$ next lecture