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VC Dimension

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must lie in $[a, b]$ where a & b are real numbers.

examples

$VC = 2$ ✓

$a=1$ $b=7 \Rightarrow [1, 7]$

Input Output

0 (0)

3 (1)

$VC = 2$ ✓

$a=1$ $b=7$

Input Output

0, 8 \rightarrow (0, 0)

0, 6 \rightarrow (0, 1)

2, 9 \rightarrow (1, 0)

2, 4 \rightarrow (1, 1)

$VC = 3$ X

$a=1$ $b=7$

Input Output

0, 1, 12 (0, 0, 0)

1, 0, 2 (0, 0, 1)

6, 9, 10 (0, 1, 0)

5, 6, 10 (1, 0, 0)

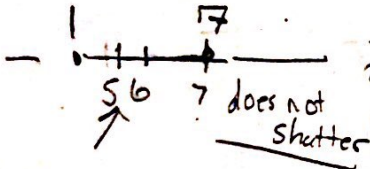
5, 6, 10 (0, 1, 1)

5, 6, 7 (1, 0, 1) +

5, 6, 8 (1, 1, 0)

2, 3, 4 (1, 1, 1)

VC dimension = 2



Number Line

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a) $f(x|\theta) = \frac{1}{\theta-1} e^{\frac{-x}{\theta-1}} \quad x > 0, \theta > 1$

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Step 1: Convert to likelihood func..

$$L(\theta) = \prod_{i=1}^n \left(\frac{1}{\theta-1} \right) e^{\frac{-x_i}{\theta-1}} \Rightarrow \sum_{i=1}^n \ln \left(\frac{1}{\theta-1} \right) e^{\frac{-x_i}{\theta-1}}$$

$$\Rightarrow \ln \left(\frac{1}{\theta-1} \right) + \sum_{i=1}^n \frac{-x_i}{\theta-1} \ln(e) \Rightarrow \ln \left(\frac{1}{\theta-1} \right) + \sum_{i=1}^n \frac{-x_i}{\theta-1}$$

Step 2: derivative

$$\frac{d}{d\theta} \left[\ln \left(\frac{1}{\theta-1} \right) + \sum_{i=1}^n \frac{-x_i}{\theta-1} \right]$$

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

Step 3 set to 0

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

$$\Rightarrow \frac{(\theta-1)^2}{(\theta-1)^3} = \sum_{i=1}^n x_i \Rightarrow \frac{1}{(\theta-1)} = \sum_{i=1}^n x_i$$

$$\theta-1 = \frac{1}{\sum_{i=1}^n x_i}$$

$$\frac{1}{\theta-1} \cdot (\theta-1)^{-1} = -1(\theta-1)^{-2}$$

$$\theta = \frac{1}{\sum_{i=1}^n x_i} + 1$$

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$$b) f(x|\theta) = (\theta - 1)x^{\theta-2}, \quad 0 \leq x \leq 1, \quad \theta > 1$$

Step 1: likelihood function

$$L(\theta) = \prod_{i=1}^n f(x_i|\theta) = \prod_{i=1}^n (\theta - 1)x_i^{\theta-2} = \sum_{i=1}^n \ln[(\theta - 1)x_i^{\theta-2}]$$

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

Step 2: differentiate wrt θ Product Rule

$$\frac{\partial L(\theta)}{\partial \theta} = \frac{\sum_{i=1}^n \partial \ln(\theta - 1) + (\theta - 2)\ln(x)}{\partial \theta} \Rightarrow \sum_{i=1}^n \frac{1}{\theta - 1} + (1)(\ln(x)) + (\theta - 2)(1)$$

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

Step 3: set to 0

$$\sum_{i=1}^n \frac{1}{\theta - 1} + \ln(x) + (\theta - 2) = 0$$

$$-\ln(x) = \frac{1}{(\theta - 1)} + (\theta - 2) \Rightarrow -\ln(x) = \frac{1}{(\theta - 1)} + \frac{(\theta - 2)(\theta - 1)}{(\theta - 1)}$$

$$-\ln(x) = \frac{(\theta - 2)(\theta - 1)}{(\theta - 1)}$$

$$= -\ln(x) = \theta - 2$$

$$\boxed{-\ln(x) + 2 = \sum_{i=1}^n \theta_i}$$

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c) $f(x|\theta) = \frac{1}{\theta}$ for $0 \leq x \leq \theta$

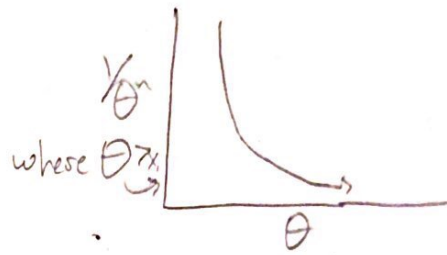
$L(\theta) = \prod_{i=1}^n f(x_i|\theta) \Rightarrow \prod_{i=1}^n \frac{1}{\theta} \Rightarrow \frac{1}{\theta^n}$ when $\max(x_i) < \theta$

We know that the $\max(x_i) < \theta$

thus,

likelihood = 0 $(0, \max(x_i))$

likelihood ≥ 0 $(\max(x_i), \infty)$



To find θ , find minimum value of $\max(x_i) \Rightarrow \theta(x_n) = \max(x_i)$

and, $\theta(x) = x_n < \theta$

so, $E_{\theta} \max(x_i) < \theta$

Expected Value of θ

$X(n) = \max_{1 \leq i \leq n} x_i \Rightarrow P(X_n < x)$

each have same distribution so,

$F_{X_n}(x) = \frac{x^n}{\theta^n}$ for $0 \leq x \leq \theta$

$f_{X_n}(x) = \frac{x^n}{\theta^n}$

$\frac{d}{dx} \frac{x^n}{\theta^n} = \frac{n x^{n-1}}{\theta^n}$

$\Rightarrow E_{\theta} = \int_0^{\theta} x \cdot \frac{n x^{n-1}}{\theta^n} dx = \frac{n}{\theta^n} \int_0^{\theta} x^n dx$

$\int_0^{\theta} x^n dx = \frac{x^{n+1}}{n+1}$

$E_{\theta} \max(x_i) = \frac{n}{\theta^n} \cdot \frac{x^{n+1}}{n+1} \Big|_0^{\theta} \Rightarrow \frac{n}{\theta^n} \cdot \frac{\theta^{n+1}}{n+1} = \frac{n}{n+1} \theta$

$X(n) = \frac{n}{n+1} \theta \Rightarrow \frac{\max(x_i)(n+1)}{n} = \theta$

Please Note: I used math.arizona.edu/~jwatkins/0-mle.pdf to help me through this problem... I was lost :)

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a) Given $P(C_1) \text{ \& } P(C_2) = \text{Prior}$
 $P_1 = P(x=0|C_1)$
 $P_2 = P(x=0|C_2)$ } likelihood

Note:

$$P(C_1) + P(C_2) = 1$$

Prior & Likelihood
 \times evidence

$$\begin{aligned} \text{evidence} = P(x=0) &= P(x=0|C_1)P(C_1) + P(x=0|C_2)P(C_2) \\ &\xrightarrow{\text{Substitute}} \\ &= P_1 P(C_1) + P_2 P(C_2) = \text{evidence} \end{aligned}$$

Note:

$$\begin{aligned} P(x=1|C_1) &= 1 - P(x=0|C_1) \\ &= 1 - P_1 \end{aligned}$$

$$P(x=1|C_2) = 1 - P(x=0|C_2)$$

$$\text{Use in equation} = 1 - P_2$$

$$P(C_1|x=0) = \frac{P(C_1) P_1}{P_1 P(C_1) + P_2 P(C_2)}$$

$$P(C_2|x=0) = \frac{P(C_2) P_2}{P_1 P(C_1) + P_2 P(C_2)}$$

$$P(C_1|x=1) = \frac{P(C_1) (1 - P_1)}{(1 - P_1) P(C_1) + (1 - P_2) P(C_2)}$$

$$P(C_2|x=1) = \frac{P(C_2) (1 - P_2)}{(1 - P_1) P(C_1) + (1 - P_2) P(C_2)}$$

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b) $P_{ij} = (x_j = 0 | C_i) \quad i = 1, 2$
 $j = 1, 2, \dots, 10$

$$P(C_i | x_j = 0) = \frac{P(C_i) \cdot P_{ij}}{\sum_{k=1}^2 P(C_k) \cdot P_{kj}}$$

$$P(C_i | x_j = 1) = \frac{P(C_i) \cdot (1 - P_{ij})}{\sum_{k=1}^2 P(C_k) \cdot (1 - P_{kj})}$$

c) $P_{11} = 0.6$ 8 calculations for each prior, $P(C_i) = 0.2, 0.6, 0.8$ (4 calculations)
 $P_{12} = 0.1$
 $P_{21} = 0.6$
 $P_{22} = 0.9$

$P(C_1) = 0.2 \quad P(C_2) = 0.8$

$j = 1$

$$P(C_1 | x_1 = 0) = \frac{P(C_1) \cdot P_{11}}{\sum_{k=1}^2 P(C_k) \cdot P_{k1}} = \frac{0.2 \cdot 0.6}{(0.2)(0.6) + (0.8)(0.6)} = \frac{0.12}{0.6} = 0.2$$

$$P(C_1 | x_1 = 1) = \frac{P(C_1) \cdot (1 - P_{11})}{\sum_{k=1}^2 P(C_k) \cdot (1 - P_{k1})} = \frac{0.2 \cdot (1 - 0.6)}{(0.2)(1 - 0.6) + (0.8)(1 - 0.6)} = \frac{0.08}{0.4} = 0.2$$

$$P(C_2 | x_1 = 0) = \frac{P(C_2) \cdot P_{21}}{\sum_{k=1}^2 P(C_k) \cdot P_{k1}} = \frac{0.8 \cdot 0.6}{(0.2)(0.6) + (0.8)(0.6)} = \frac{0.48}{0.6} = 0.8$$

$$P(C_2 | x_1 = 1) = \frac{P(C_2) \cdot (1 - P_{21})}{\sum_{k=1}^2 P(C_k) \cdot (1 - P_{k1})} = \frac{0.8 \cdot (1 - 0.6)}{(0.2)(1 - 0.6) + (0.8)(1 - 0.6)} = \frac{0.32}{0.4} = 0.8$$

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3c) Continued.

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$$j=2 \quad P(C_1)=0.2 \quad P(C_2)=0.8$$

$$P(C_1|x_2=0) = \frac{P(C_1) \cdot P_{12}}{\sum_{k=1}^2 P(C_k) \cdot P_{k2}} = \frac{(0.2) \cdot (0.1)}{(0.2)(0.1) + (0.8)(0.9)} = \frac{.02}{.74} = \underline{0.027}$$

$$P(C_1|x_2=1) = \frac{P(C_1) \cdot (1-P_{12})}{\sum_{k=1}^2 P(C_k) \cdot (1-P_{k2})} = \frac{(0.2)(1-0.1)}{(0.2)(1-0.1) + (0.8)(1-0.9)} = \frac{.18}{.26} \approx \underline{0.6923}$$

$$P(C_2|x_2=0) = \frac{P(C_2) \cdot P_{22}}{\sum_{k=1}^2 P(C_k) \cdot P_{k2}} = \frac{(0.8)(0.9)}{(0.2)(0.1) + (0.8)(0.9)} = \frac{.72}{.74} = \underline{0.9729}$$

$$P(C_2|x_2=1) = \frac{P(C_2) \cdot (1-P_{22})}{\sum_{k=1}^2 P(C_k) \cdot (1-P_{k2})} = \frac{(0.8)(1-0.9)}{(0.2)(1-0.1) + (0.8)(1-0.9)} = \frac{.08}{.26} \approx \underline{0.30769}$$

$$j=1 \quad P(C_1)=0.6 \quad P(C_2)=0.4$$

Note, I will stop writing equation before, as they are all listed previously...

$$P(C_1|x_1=0) = \frac{(0.6)(0.6)}{(0.6)(0.6) + (0.4)(0.6)} = \frac{.36}{.6} = \underline{0.6}$$

$$P(C_1|x_1=1) = \frac{(0.6)(1-0.6)}{(0.6)(1-0.6) + (0.4)(1-0.6)} = \frac{.24}{.4} = \underline{0.6}$$

$$P(C_2|x_1=0) = \frac{(0.4)(0.6)}{(0.6)(0.6) + (0.4)(0.6)} = \frac{.24}{.6} = \underline{0.4}$$

$$P(C_2|x_1=1) = \frac{(0.4)(1-0.6)}{(0.6)(1-0.6) + (0.4)(1-0.6)} = \frac{.16}{.4} = \underline{0.4}$$

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3 c) Continued

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$$j=2 \quad P(C_1) = 0.6 \quad P(C_2) = 0.4$$

$$P(C_1 | x_2 = 0) = \frac{(0.6)(0.1)}{(0.6)(0.1) + (0.4)(0.9)} = \frac{0.06}{.42} \approx \underline{0.1429}$$

$$P(C_1 | x_2 = 1) = \frac{(0.6)(1-0.1)}{(0.6)(1-0.1) + (0.4)(1-0.9)} = \frac{.54}{.58} \approx \underline{0.9310}$$

$$P(C_2 | x_2 = 0) = \frac{(0.4)(0.9)}{(0.6)(0.1) + (0.4)(0.9)} = \frac{.36}{.42} \approx \underline{0.85714}$$

$$P(C_2 | x_2 = 1) = \frac{(0.4)(1-0.9)}{(0.6)(1-0.1) + (0.4)(1-0.9)} = \frac{.04}{.58} \approx \underline{0.06897}$$

$$j=1 \quad P(C_1) = 0.8 \quad P(C_2) = 0.2$$

$$P(C_1 | x_1 = 0) = \frac{(0.8)(0.6)}{(0.2)(0.6) + (0.8)(0.6)} = \frac{.48}{.6} = \underline{0.8}$$

$$P(C_1 | x_1 = 1) = \frac{(0.8)(1-0.6)}{(0.2)(1-0.6) + 0.8(1-0.6)} = \frac{.32}{0.4} = \underline{0.8}$$

$$P(C_2 | x_1 = 0) = \frac{(0.2)(0.6)}{(0.2)(0.6) + (0.8)(0.6)} = \frac{0.12}{0.6} = \underline{0.2}$$

$$P(C_2 | x_1 = 1) = \frac{(0.2)(1-0.6)}{(0.2)(1-0.6) + (0.8)(1-0.6)} = \frac{0.08}{0.4} = \underline{0.2}$$

3c) Continued

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$$j=2 \quad P(c_1) = 0.8 \quad P(c_2) = 0.2$$

$$P(c_1|x_2=0) = \frac{(0.8)(0.1)}{(0.8)(0.1) + (0.2)(0.9)} = \frac{.08}{.26} \approx 0.3077$$

$$P(c_1|x_2=1) = \frac{(0.8)(1-0.1)}{(0.8)(1-0.1) + (0.2)(1-0.9)} = \frac{.72}{.74} \approx 0.9727$$

$$P(c_2|x_2=0) = \frac{(0.2)(0.9)}{(0.8)(0.1) + (0.2)(0.9)} = \frac{.18}{.26} \approx 0.6923$$

$$P(c_2|x_2=1) = \frac{(0.2)(1-0.9)}{(0.8)(1-0.1) + (0.2)(1-0.9)} = \frac{.02}{.74} \approx 0.027$$