

STAT 417. Lecture Note 5.

① why need statistics? what is Indiana annual salary? \bar{x}
on the road, randomly pick up ten persons, their annual salaries are

121.2 K, 92.1 K, 56.9 K, 72.4 K, 31.7 K,
64.5 K, 21.8 K, 46.9 K, 82.3 K, 51.1 K

Question: How much does a typical person in Indiana earn?

of course, we don't know the true Indiana annual salary.
but we want to use the above ten samples to infer this \bar{x} .
Denote the simplest way is to use the sample mean of the
ten samples to estimate \bar{x} . In the above example, $\bar{x} = 64.1 \text{ K}$

② How do we handle statistics? probability model is used to
describe variation in the data.

E.g.1. Assuming Indiana annual salary \bar{x} follows some probability model.

say, $\bar{x} \sim \text{Exponential}(10)$, then the above ten samples will
be considered as random numbers drawn
~~from~~ from this exponential distribution.

③ main components of statistics:
 ◀ Estimation
 ◀ testing hypothesis
 ◀ confidence interval.

Statistical Models:

E.g. 3. X is the lifetime of ~~house~~^{BMW}. $X \sim \text{Exponential}(\theta)$. θ unknown.

E.g. 2 suppose 8 patients each has ~~per~~ chance θ to die. $0 \leq \theta \leq 1$

let $X = \#$ of patients dying. Then $X \sim \text{Binomial}(8, \theta)$.

In general, we suppose $\{P_\theta : \theta \in \Theta\}$, is a collection of probability models.

In e.g. 2. $\Theta = \{0, 1\}$. $P_\theta \sim \text{Binomial}(8, \theta)$. $\mathcal{S} = \{0, 1, \dots, 8\}$.

e.g. 3. $\Theta = (0, \infty)$, $P_\theta \sim \text{Exponential}(\theta)$. $\mathcal{S} = (0, \infty)$.

E.g. 4. $\Theta = \{1, 2\}$. $P_1 \sim \text{unif}\{1, 2, \dots, 1000\}$, $P_2 \sim \text{unif}\{1, 2, \dots, 10000\}$.

Sample space $\mathcal{S} = \{1, 2, \dots, 3\}$ is for a single draw.

~~E.g.~~ sample space $\mathcal{S} = \{(1, 1), (1, 2), \dots, (2, 1), (2, 2), \dots\}$ is for two draws.

~~(normal model):~~ ~~continuous~~ ~~1-dim~~.

E.g. 5. $\Theta = \{(\mu, \sigma^2) : \mu \in \mathbb{R}, \sigma^2 > 0\}$. $P_\theta \sim N(\mu, \sigma^2)$

$P_\theta \sim \mu(\mu, \sigma^2)$, where $\theta = (\mu, \sigma^2)$. For a single draw $X \sim P_\theta$.

The sample space $\mathcal{S} = \mathbb{R}$.

E.g. 6. (poisson model): $P_\theta \sim \text{poisson}(\theta)$, $\theta \in \Theta = (0, \infty)$.

The sample space of a single draw $X \sim P_\theta$ is $\mathcal{S} = \{0, 1, 2, \dots\}$.

E.g. 7. (Bernoulli model). $P_\theta \sim \text{Bernoulli}(\theta)$, $\theta \in \Theta = \{0, 1\}$.

The sample space of a single draw $X \sim P_\theta$ is $\mathcal{S} = \{0, 1\}$.

Ch6 Likelihood Inference.

A warm-up example: 8 patients, each has probability θ to die, $0 \leq \theta \leq 1$.

Let $X = \#$ of patients dying, $X \sim \underbrace{\text{Binomial}(8, \theta)}_{P_\theta}$, $\theta \in \Theta = [0, 1]$.

Suppose we know that there are 4 people dying. Can you infer θ ?

ambitiously, $\theta = \frac{4}{8} = \frac{1}{2}$.

§6.1. Likelihood function.

Suppose $\{P_\theta : \theta \in \Theta\}$. we ~~have~~ have data $s \sim P_\theta$

~~In discrete model, let $f_\theta(s)$ be the~~

In discrete model, f_θ is the ~~prob~~ probability distribution under model P_θ .

In continuous model, f_θ is ^{the} pdf. under model P_θ .

The likelihood function, denoted $L(\theta | s)$; is defined to be

$$L(\theta | s) = f_\theta(s).$$

Eg1. Suppose sample space $\{s = \{1, 2, \dots\}\}$, the statistical model is

$\{P_\theta : \theta \in \{1, 2\}\}$, where ~~P₁~~ $P_1 \sim \text{unif}\{1, 2, \dots, 1000\}$.

$P_2 \sim \text{unif}\{1, 2, \dots, 10,000\}$. Furthermore, $s = 10$,

then θ under P_1 , $f_1(s) = \frac{1}{1000}$.

under P_2 , $f_2(s) = \frac{1}{10,000}$.

so the likelihood function. $L(\theta | s) = \begin{cases} \frac{1}{1000}, & \theta=1, \\ \frac{1}{10,000}, & \theta=2. \end{cases}$