

STAT 703/J703  
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-Lecture 27-

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Concepts for Bayes...

1) A player recently promoted to the major leagues has had 1 hit in his first 25 at bats. What do you estimate his batting average to be? (Batting average = % of times a hit is gotten in an at bat).



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2) Consider your answer in 1. You are then told that the batting averages of professional major league players has a mean of around 0.266 and a standard deviation of around 0.026. What do you think about your estimate in 1 now?



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## Bayesian Statistics

### Classical Setup

$\theta$  is an unknown constant

$X_1, \dots, X_n$  are observed from  $f(x|\theta)$

### Bayesian Setup

$\Theta$  is an unknown random variable, unobservable with pdf  $g(\theta)$

$X_1, \dots, X_n$  are observed from the conditional pdf  $f(x|\theta)$



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### A Bayesian Statistical Model:

1. An observed sample  $X_1, \dots, X_n$ ,  $(X_1, \dots, X_n) = \underline{X} \in \mathcal{X}$ , the data space.
2. An unobserved random vector  $\underline{\Theta}$  of parameters,  $\underline{\Theta} \in \Omega$ , the parameter space.
3. The conditional density of  $X_1, \dots, X_n$ , given  $\underline{\Theta} = \underline{\theta}$ ,  $f(\underline{x}|\underline{\theta})$ .
4. The (marginal) pdf of  $\underline{\Theta}$ ,  $g(\underline{\theta})$ , the prior distribution of  $\underline{\Theta}$ .



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Bayes Rule can be written

$$h(\theta|x) = \frac{f(x|\theta)g(\theta)}{\int f(x|\theta)g(\theta)d\theta}$$

Say we knew  $f(x|\theta)$  and  $g(\theta)$  and wanted to find the maximum likelihood estimate of  $h(\theta|x)$ . Why can we just find the value of  $\theta$  that maximizes  $f(x|\theta)g(\theta)$  and not worry about the bottom integral?



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Hence, inferences are described in terms of the posterior distribution of  $\Theta$ , the conditional pdf of  $\Theta$ , given  $\underline{X}=(x_1, \dots, x_n)$ ,  $h(\underline{\theta}|\underline{x})$ . (Interpretation of posterior).

We consider Bayesian point and interval estimators as well as Bayesian tests (decisions). The point estimate is given by a Bayes rule.




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Example - Sample  $X_1, \dots, X_n$  from a normal population whose variance is 1 and mean  $\mu$  is a value of a random variable  $M$ ,

$$X_i|M=\mu \sim N(\mu, 1),$$

$$f(x|\mu) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2}}, \mu \in R$$

Suppose  $M \sim N(\alpha, \tau^2)$  as the prior pdf.




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The “Bayesian confidence interval” is called a credibility interval:

$$(\theta_0, \theta_1) \text{ where } \int_{\theta_1}^{\theta_2} h(\theta|x) d\theta = 1 - \alpha$$




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“Bayesian hypothesis testing”  
depends on a loss function  
evaluated for competing  
hypotheses.



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