

## Recall: Bayesian Statistics

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$p \rightarrow$  prop. of goldfish in lake.

PRIOR:  $\text{Beta}(\alpha, \beta)$

LIKELIHOOD: observed  $y$

POSTERIOR:  $\text{Beta}(\alpha + y, \beta + n - y)$

Prior Guess at  $p$ :  $\eta = \frac{\alpha}{\alpha + \beta}$ .

NEW GUESS:  $\eta^* = \frac{\alpha^*}{\alpha^* + \beta^*}$ .

$$= \underbrace{\eta}_{\text{PRIOR MEAN}} \cdot \frac{\kappa}{\kappa + n} + \underbrace{\frac{y}{n}}_{\text{MLE } \hat{p}} \cdot \frac{n}{\kappa + n}$$

POSTERIOR MEAN is a  
WEIGHTED average of MLE  
and PRIOR MEAN.

Two things:

$$\text{as } \kappa \rightarrow \infty, \eta^* \rightarrow \eta$$

$$\text{as } n \rightarrow \infty, \eta^* \rightarrow \frac{y}{n} \text{ (MLE)}$$

REMARKS:

1) How do we choose a Prior?

Notice  $\rightarrow$  Prior  $\rightarrow$  Beta dist'n

Posterior  $\rightarrow$  also Beta.

Beta( $\alpha, \beta$ ) is a CONJUGATE FAMILY.

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Beta/Binomial  $\rightarrow$  "Conjugate Pair."

More Conjugate Pairs:

\* If  $\pi(\theta) = N(\mu, \tau^2)$

$$f(x|\theta) = N(\theta, \sigma^2)$$

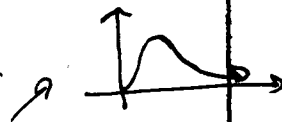
Then

$$\pi(\theta|x) = N\left(\frac{\bar{x}\tau^2 + \mu\sigma^2/n}{\tau^2 + \sigma^2/n}, \frac{\tau^2 \cdot \sigma^2/n}{\tau^2 + \sigma^2/n}\right)$$

\* Gamma/Normal

\* Gamma/Poisson

\* Gamma/Gamma



Gamma( $\alpha, \beta$ ) has PDF

$$f(x|\alpha, \beta) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta}, \quad \begin{array}{l} x > 0 \\ \alpha > 0 \\ \beta > 0 \end{array}$$

Special Cases:

If  $\alpha = 1$  then Gamma( $1, \beta$ )  
= Exponential( $\beta$ )

If  $\alpha = \nu/2$  and  $\beta = 2$

then Gamma( $\nu/2, 2$ )

$$= \chi^2(\nu)$$

"Chi-square with  $\nu$  degrees of freedom!"

Conjugate families chosen  
for priors historically because

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- EASY
- TRACTABLE
- CONVENIENT.

Unfortunately, very RESTRICTIVE.

Used to be a problem. NOT ANYMORE.

## ② Bayesian Point Estimation.

Our new guess at  $\theta$ :

Posterior MEAN  $\rightarrow \mathbb{E}[\theta|x]$

FACT:  $\mathbb{E}(\theta|x)$  is OPTIMAL in  
almost every sense under  
SQUARED ERROR LOSS.

Def<sup>n</sup>: The LOSS of an estimator  $\delta$   
which estimates  $\theta$  is:

$$L(\delta, \theta) = (\delta - \theta)^2$$

So we will need to compute  
things like this:

$$\mathbb{E}(\theta|x) = \int \theta \cdot \pi(\theta|x) d\theta$$

$$= \int \theta \cdot \frac{f(x|\theta) \cdot \pi(\theta)}{\int f(x|\theta) \cdot \pi(\theta) d\theta} d\theta$$

$$= M \cdot \int \theta f(x|\theta) \pi(\theta) d\theta$$

This is hard. Need computers.

### ③ Bayesian Interval Estimation.

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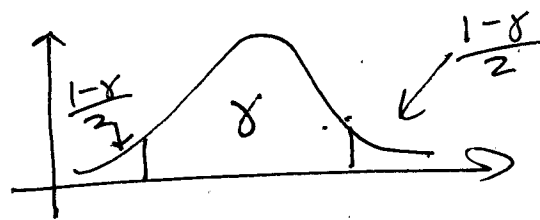
Probability interval of "content"  $\gamma$ .



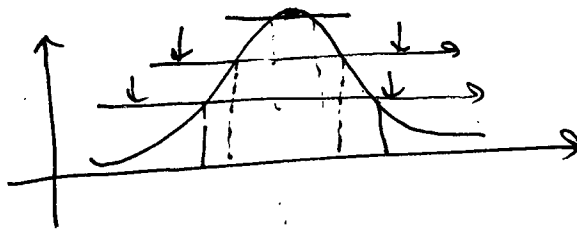
"95% CREDIBLE INTERVAL"

Two examples:

a) EQUAL TAILS interval:



b) Shortest interval of content  $\gamma$   
HDR "Highest Density Region"



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