

Stat 543 1-28-05

Correction to last example:

$$R(x, 1, 1) = \exp\left(\sum x_i - \frac{n}{2}\right)$$

$$R(x, 0, 2) = \left(\frac{1}{\sqrt{2}}\right)^n \exp\left(\frac{1}{4} \sum y_i^2\right)$$

A type of model in which it is easy to identify a minimal sufficient statistic (and which has lots of other convenient mathematical properties) is an "exponential family" model

Def If T_1, T_2, \dots, T_k are linearly independent real-valued functions on \mathbb{R}^1 or on some discrete space, $h(x) \geq 0$ and for parameters $\eta \in \mathbb{R}^k$ pdfs or pmfs

$$f(x|\eta) = A(\eta) h(x) \exp \sum_{j=1}^k \eta_j T_j(x)$$

we'll call the family of dsns an exponential family
(with "natural parameter" η)

Fact: $\Sigma = \left\{ \eta \in \mathbb{R}^k \mid \int h(x) \exp \sum_{j=1}^k \eta_j T_j(x) dx < \infty \right\}$

is a convex subset of \mathbb{R}^k - for a given observation space and set of functions T_1, \dots, T_k this is the largest possible parameter space for the exponential family

$$A(\eta) = \left(\int \right)^{-1}$$

- Σ is called the "natural parameter space"

Note: If X_1, X_2, \dots, X_n are iid $f(x|\eta)$

the joint pdf or pmf is of the form

$$f(x|\eta) = A(\eta)^n \prod_{i=1}^n h(x_i) \exp \sum_{j=1}^k \eta_j \left(\sum_{i=1}^n T_j(x_i) \right)$$

and the factorization Thm says that

$$\rightarrow T(X) = \left(\sum T_1(x_i), \sum T_2(x_i), \dots, \sum T_k(x_i) \right)$$

is sufficient for η in any $\mathcal{E}^* \subset \mathcal{E}$ -

This is called the "natural sufficient statistic"
for the parameter η (or for the family

$$\mathcal{P} = \{ P_\eta \}_{\eta \in \mathcal{E}^*}$$

not only is the natural sufficient statistic sufficient but provided \mathcal{E}^* is "big enough" it is also minimal sufficient - that is

Thm If \mathcal{E}^* ($\subset \mathcal{E}$) contains an open rectangle in \mathbb{R}^k then $T(X)$ above is minimal sufficient

— Argument for this is on a handout... it involves showing

- i) $T(X)$ has a property called "completeness"
- ii) Bahadur's Thm to be true

This is not really the form in which one likes to apply this result

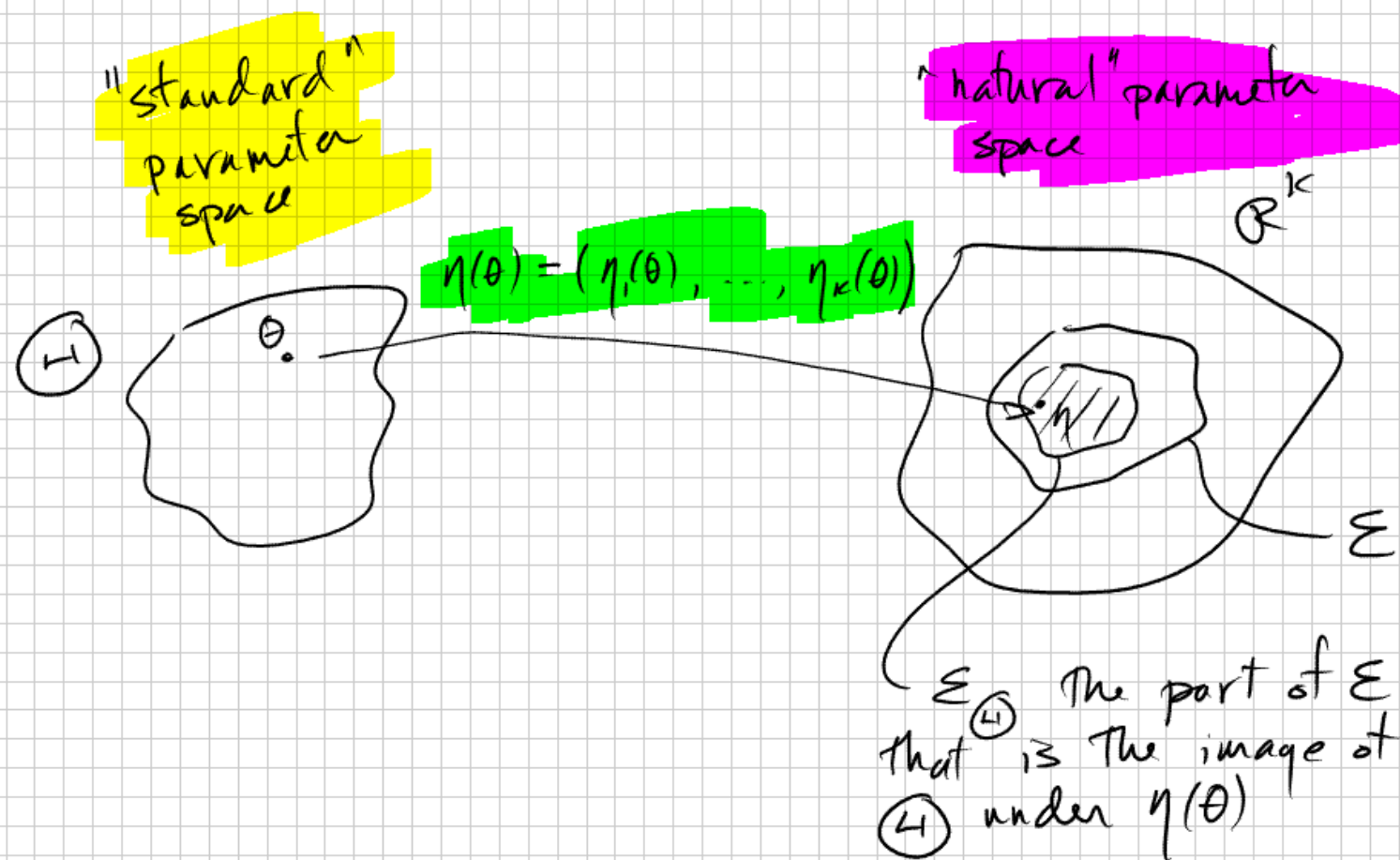
Example Poisson (λ) — marginal pmf here on $\{0, 1, 2, \dots\}$ is

$$f(x|\lambda) = \frac{e^{-\lambda} \lambda^x}{x!}$$

$$= e^{-\lambda} \left(\frac{1}{x!} \right) e^{x \log \lambda}$$

The diagram shows the decomposition of the Poisson pmf into its natural parameterization. The term $\frac{1}{x!}$ is circled and labeled $h(x)$. The term $e^{x \log \lambda}$ is circled and labeled $T_1(x)$. The parameter λ is labeled η_1 .

That is, the "standard" parameterization (involving λ) is not the "natural" parameterization (involving η)



$T(X)$ sufficient for $\eta \in \mathcal{E}$
 $\Rightarrow T(X)$ sufficient of $\eta \in \mathcal{E}_{\Theta}$

and if \mathcal{E}_{Θ} contains an open rectangle we
 apply the theorem to conclude that $T(X)$ is
 minimal sufficient for $\eta \in \mathcal{E}_{\Theta}$
 for $\theta \in \Theta$

Back to Poisson Example
 $f(x|\lambda) = e^{-\lambda} \frac{1}{x!} e^{x \log \lambda}$
 $\eta(\lambda) = \log \lambda \quad \lambda \in (0, \infty) = \Theta$

E_{Θ} = the image of Θ under the log function
 $\Theta = \mathbb{R}^1$ contains an open interval

So in an iid problem, since we're working in an exponential family

$$T(X) = \sum_{i=1}^n T_i(X_i) = \sum_{i=1}^n X_i$$

is not only sufficient but is minimal sufficient

Example Normal (μ, σ^2) marginal

$$f(x | \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{x^2}{2\sigma^2} + \frac{x\mu}{\sigma} - \frac{\mu^2}{2\sigma^2}\right)$$

use $T_1(x) = x^2$ $T_2(x) = x$

$$\eta_1 = -\frac{1}{2\sigma^2} \quad \eta_2 = \frac{\mu}{\sigma^2}$$