

# AN EMPIRICAL BAYES APPROACH TO STATISTICS

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Let  $X$  be a random variable which for simplicity we shall assume to have discrete values  $x$  and which has a probability distribution depending in a known way on an unknown real parameter  $\Lambda$ ,

$$(1) \quad p(x|\lambda) = Pr[X = x | \Lambda = \lambda],$$

$\Lambda$  itself being a random variable with a *a priori* distribution function

$$(2) \quad G(\lambda) = Pr[\Lambda \leq \lambda].$$

The unconditional probability distribution of  $X$  is then given by

$$(3) \quad p_G(x) = Pr[X = x] = \int p(x|\lambda) dG(\lambda),$$

and the expected squared deviation of any estimator of  $\Lambda$  of the form  $\varphi(X)$  is

$$(4) \quad \begin{aligned} E[\varphi(X) - \Lambda]^2 &= E\{E[(\varphi(X) - \Lambda)^2 | \Lambda = \lambda]\} \\ &= \int \sum_x p(x|\lambda) [\varphi(x) - \lambda]^2 dG(\lambda) \\ &= \sum_x \int p(x|\lambda) [\varphi(x) - \lambda]^2 dG(\lambda), \end{aligned}$$

which is a minimum when  $\varphi(x)$  is defined for each  $x$  as that value  $y = y(x)$  for which

$$(5) \quad I(x) = \int p(x|\lambda) (y - \lambda)^2 dG(\lambda) = \text{minimum}.$$

But for any fixed  $x$  the quantity

$$(6) \quad \begin{aligned} I(x) &= y^2 \int p dG - 2y \int p\lambda dG + \int p\lambda^2 dG \\ &= \int p dG \left( y - \frac{\int p\lambda dG}{\int p dG} \right)^2 + \left[ \int p\lambda^2 dG - \frac{(\int p\lambda dG)^2}{\int p dG} \right] \end{aligned}$$

is a minimum with respect to  $y$  when

$$(7) \quad y = \frac{\int p\lambda dG}{\int p dG},$$

the minimum value of  $I(x)$  being

$$(8) \quad I_G(x) = \int p(x|\lambda) \lambda^2 dG(\lambda) - \frac{[\int p(x|\lambda) \lambda dG(\lambda)]^2}{\int p(x|\lambda) dG(\lambda)}.$$

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