Bayesian Statistics

Fabio Sigrist

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Today's topics

► Empirical Bayes methods

Examples

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Empirical Bayes methods

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Empirical Bayes method

Recall that the marginal posterior can be computed as

$$\pi(\theta \mid \mathbf{x}) \propto \int \pi(\theta \mid \mathbf{x}, \xi) f(\mathbf{x} \mid \xi) \pi(\xi) d\xi$$

 Instead of approximating this integral, the empirical Bayes method uses

$$\pi(\theta \mid x) \approx \pi(\theta \mid x, \widehat{\xi}(x)) = \frac{f(x|\theta)\pi(\theta|\widehat{\xi}(x))}{f(x|\widehat{\xi}(x))}$$

where

$$\widehat{\xi}(x) = \arg\max_{\xi} f(x \mid \xi)$$

and

$$f(x \mid \xi) = \int f(x|\theta)\pi(\theta|\xi)d\theta$$

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Empirical Bayes method

- $\widehat{\xi}(x)$ is the marginal maximum likelihood estimator of the hyperparameter
- Instead of taking a weighted average, one takes the value with maximal weight (assuming that $\pi(\xi)$ is flat around $\widehat{\xi}(x)$)

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Hierarchical Bayes models Empirical Bayes methods

Comments on the empirical Bayes method

Advantage

This method avoids not only the computation of the integral, but also the choice of a hyperprior $\pi(\xi)$

Drawbacks

- ► The data x is used twice:
 - 1. First, to select the prior $\pi(\theta \mid \widehat{\xi}(x))$
 - 2. Then, to compute the posterior according to Bayes formula
 - This is somewhat undesirable from a conceptual point of view
- In general, the uncertainty is underestimated

Conclusion

- Due to this, Bayesians often avoid empirical Bayes methods
- From a pragmatic point of view, empirical Bayes methods can be useful and have good frequentist properties

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Examples of the application of empirical Bayes methods

- Normal means
- Hierarchical Poisson model
- One-way ANOVA model

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Normal means example

Assume the following model:

- Likelihood: $X_1, \ldots X_n$ i.i.d. $\sim \mathcal{N}(\theta, 1)$
- Hierarchical prior:
 - \bullet $\theta \sim \mathcal{N}(\mu, \tau^2)$, μ known
 - $ightharpoonup au^{-2} \sim \operatorname{Gamma}(\gamma, \lambda)$

See blackboard for more details

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Hierarchical Poisson model example

Assume the following model:

- ▶ Likelihood: X_j independent \sim Poisson (θ_j) , j = 1, 2, ..., J
- Hierarchical prior:
 - ▶ θ_j i.i.d. ~ Gamma(γ , λ)
 - Some prior $\pi(\gamma, \lambda)$

See blackboard for more details

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One-way ANOVA model example

Assume the following model:

- Likelihood: $y_{ij} = \theta_i + \varepsilon_{ij}$, ε_{ij} i.i.d $\sim \mathcal{N}(0, \sigma_{\varepsilon}^2)$, $j = 1, \ldots, n_i, i = 1, \ldots, I$
- Hierarchical prior:
 - $ightharpoonup heta_i ext{ i.i.d } \sim \mathcal{N}(\mu, \tau^2)$
 - For simplification, σ_{ε} is assumed to be known
 - $\pi(\mu, \tau^2) = \pi(\mu)\pi(\tau^2)$ (specified later)

See blackboard for more details

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Fabio Sigrist 9/9