

## Semiparametric Estimation for Kernel Families

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Abstract

Suppose that the statistician has a sample of (large) size  $n$  from a population  $P_\theta$  where not only a (finite-dimensional) parameter of interest  $\theta$  is unknown, but the available information on the family  $\{P_\theta, \theta \in \Theta\}$  is in the form of a sample of (large) size  $m$  from  $P_{\theta_0}$  where  $\theta_0 \in \Theta$  is known. How to estimate  $\theta$  in this setup?

We consider a class of these problems that may, on one side, be useful for applications (at least, in the speaker's opinion) and, on the other, allows a rigorous mathematical statistical analysis.

A family  $\mathcal{P} = \{P_\theta, \theta \in \Theta\}$  of probability distributions on a measurable space  $(\mathcal{X}, \mathcal{A})$  is called a *kernel* family with kernel  $h(x; \theta) \geq 0$ , generator  $P$  (a probability distribution on  $(\mathcal{X}, \mathcal{A})$ ), and parameter  $\theta$  if

$$dP_\theta(x) = C(\theta)h(x; \theta)dP(x)$$

with a normalizer  $C(\theta)$ .

A classical example of a kernel family is a natural exponential family (NEF) of distribution functions  $F(x; \theta)$  on  $(\mathbb{R}, \mathcal{B})$ ,

$$dF(x; \theta) = \exp\{\theta x - \psi(\theta)\}dF(x)$$

depending on a scalar parameter  $\theta$ .

A general setup

$$dF(x; \theta) = C(\theta)h(x; \theta)dF(x)$$

is considered when both  $\theta$  and  $F$  are unknown and the available data are two independent samples,  $(x'_1, \dots, x'_m)$  from  $F(x)$  and  $(x_1, \dots, x_n)$  from  $F(x; \theta)$ . The setup can be interpreted as a case-control study when  $F$  is the distribution of a characteristic of interest in the control group and the parameter  $\theta$  measures the effect of a treatment.

The asymptotic behavior as  $m, n \rightarrow \infty$  of the method of moments estimator (MME) and the maximum likelihood estimator (MLE) of  $\theta$  is described in cases of (i)  $m = cn(1 + o(1))$ ,  $c > 0$ , (ii)  $n = o(m)$  and (iii)  $m = o(n)$ . An interesting phenomenon is that the asymptotic variance of the MLE is not always less than that of the MME.