

2 Basic Approach

M-estimators are solutions of the vector equation

$$\sum_{i=1}^n \psi(\mathbf{Y}_i, \boldsymbol{\theta}) = \mathbf{0}.$$

but what are they estimating?

Example (Sample Mean, cont'd): Recall we said $\theta = \bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$ is an M-estimator for $\psi(Y_i, \theta) = Y_i - \theta$. What is the true parameter?

To arrive at the sandwich estimator, assume $\mathbf{Y}_1, \dots, \mathbf{Y}_n \stackrel{iid}{\sim} F$ and define

$$\mathbf{G}_n(\boldsymbol{\theta}) = \frac{1}{n} \sum_{i=1}^n \boldsymbol{\psi}(\mathbf{Y}_i; \boldsymbol{\theta}).$$

Taylor expansion of $\mathbf{G}_n(\boldsymbol{\theta})$ around $\boldsymbol{\theta}_0$ evaluated at $\hat{\boldsymbol{\theta}}$ yields

Define $\mathbf{A}(\boldsymbol{\theta}_0) = \mathbb{E}_F[-\boldsymbol{\psi}'(\mathbf{Y}_1, \boldsymbol{\theta}_0)]$.

2.1 Estimators for A, B

If the data truly come from the assumed parametric family $f(y; \theta)$,

One of the key contributions of M-estimation theory is to point out what happens when the assumed parametric family is not correct.

We can use empirical estimators of A and B :

Example (Coefficient of Variation): Let Y_1, \dots, Y_n be iid from some distribution with finite fourth moment. The coefficient of variation is defined at $\hat{\theta}_3 = s_n/\bar{Y}$.

Define a three dimensional $\boldsymbol{\psi}$ so that $\hat{\theta}_3$ is defined by summing the third component. What is the vector valued function $\boldsymbol{\psi}$ which yields an M-estimator for the coefficient of variation?

What parameter vector is being estimated by the M-estimator?

What are the matrices \mathbf{A} and \mathbf{B} ?

Write out the asymptotic variance, \mathbf{V} .

Assume Y_i are iid from a normal distribution with mean 10 and standard deviation 1. Calculate $V_{3,3}$. Assume you have a sample of size 25 and you get an estimated coefficient of variation of 0.11. Give the asymptotic 95% confidence interval.