

## Some repetition from week 42

### Sufficient statistics

A statistic  $T(\mathbf{X})$  is a sufficient statistic for  $\theta$  if the conditional distribution of the sample  $\mathbf{X}$  given the value of  $T(\mathbf{X})$  does not depend on  $\theta$ .

A sufficient statistics for a parameter (-vector)  $\theta$  is a statistic that in a certain sense, captures all the information about  $\theta$  in the sample.

### Theorem 6.2.2

If  $p(\mathbf{x}|\theta)$  is the pdf/pmf of  $\mathbf{X}$  and  $q(t|\theta)$  is the pdf/pmf of  $T(\mathbf{X})$ , then  $T(\mathbf{X})$  is a sufficient statistics for  $\theta$  if, for every  $\mathbf{x}$  in the sample

space the ratio  $\frac{p(\mathbf{x}|\theta)}{q(T(\mathbf{x})|\theta)}$  is a constant as a function of  $\theta$ .

### Theorem 6.2.6

Let  $f(\mathbf{x}|\theta)$  be the joint pdf/pmf for a sample  $\mathbf{X}$ .  $T(\mathbf{X})$  is a sufficient statistics for  $\theta$  if and only if for all  $\mathbf{x}$  and all  $\theta$ .

$$f(\mathbf{x}|\theta) = g(T(\mathbf{X}|\theta))h(\mathbf{x})$$

### **Minimal sufficient.**

Definition 6.2.11. A sufficient statistics  $T(\mathbf{X})$  is called a minimal sufficient statistics if for any other sufficient statistics  $T'(\mathbf{X})$ ,  $T(\mathbf{X})$  is a function of  $T'(\mathbf{X})$ .

### **Theorem 6.2.3**

Let  $f(\mathbf{x}|\theta)$  be the joint pdf/pmf for a sample  $\mathbf{X}$ . Suppose there exists a  $T(\mathbf{X})$  such that for every  $\mathbf{x}$  and every  $\mathbf{y}$ ,  $f(\mathbf{x}|\theta)/f(\mathbf{y}|\theta)$  is a constant as a function of  $\theta \Leftrightarrow T(\mathbf{X})=T(\mathbf{Y})$ . Then  $T(\mathbf{X})$  is a minimal sufficient statistics for  $\theta$ .

### **Definition 6.2.21**

Let  $f(t|\theta)$  be a family of pdfs/pmfs for a statistic  $T(\mathbf{X})$ . The family is complete if

$$E_{\theta}[g(T)] = 0 \Rightarrow P_{\theta}(g(T) = 0) = 1, \text{ for all } \theta.$$