

**Theorem 1** (Factorization Theorem). *Let  $(\mathcal{X}, \mathcal{A})$  be a measurable space, and let  $\mathcal{P} = \{P_\theta : \theta \in \Theta\}$  be a family of probability measures on  $(\mathcal{X}, \mathcal{A})$ . Suppose there exists a  $\sigma$ -finite measure  $\mu$  on  $(\mathcal{X}, \mathcal{A})$  such that each  $P_\theta$  is absolutely continuous with respect to  $\mu$ , and let  $p_\theta(x) = \frac{dP_\theta}{d\mu}(x)$  denote the Radon-Nikodym derivative.*

*A statistic  $T : \mathcal{X} \rightarrow \mathcal{Y}$  is sufficient for  $\theta$  if and only if there exist measurable functions  $u : \mathcal{Y} \times \Theta \rightarrow [0, \infty)$  and  $v : \mathcal{X} \rightarrow [0, \infty)$  such that for all  $x \in \mathcal{X}$  and all  $\theta \in \Theta$ ,*

$$p_\theta(x) = u(T(x), \theta) v(x).$$

*Proof.* We will prove both directions of the equivalence.

**( $\Rightarrow$ ) Sufficiency implies factorization:**

Assume that  $T$  is sufficient for  $\theta$ . By the definition of sufficiency, the conditional distribution of  $X$  given  $T(X) = t$  does not depend on  $\theta$ . Formally, for any measurable set  $B \in \mathcal{A}$  and for all  $\theta \in \Theta$ ,

$$P_\theta(X \in B \mid T(X) = t) = P(X \in B \mid T(X) = t),$$

where the right-hand side is independent of  $\theta$ .

Consider the conditional density of  $X$  given  $T(X) = t$ , which we denote by  $k(x \mid t)$ . Since this density does not depend on  $\theta$ , we can write

$$p_\theta(x) = P_\theta(X = x) = P_\theta(T(X) = t, X = x) = P_\theta(X = x \mid T(X) = t) P_\theta(T(X) = t).$$

However, in continuous settings, we need to be careful with densities. Instead, we can use the following approach.

Let  $f_{T,\theta}(t)$  be the marginal density of  $T(X)$  under  $P_\theta$ , and let  $k(x \mid t)$  be the conditional density of  $X$  given  $T(X) = t$ , which does not depend on  $\theta$  due to sufficiency. Then,

$$p_\theta(x) = f_{T,\theta}(t) k(x \mid t).$$

Since  $k(x \mid t)$  does not depend on  $\theta$ , we can let

$$v(x) = k(x \mid t),$$

and since  $t = T(x)$ ,  $v(x)$  is a function of  $x$  alone.

Now, define

$$u(t, \theta) = f_{T,\theta}(t).$$

Therefore, we have

$$p_\theta(x) = u(T(x), \theta) v(x).$$

**( $\Leftarrow$ ) Factorization implies sufficiency:**

Assume that there exist measurable functions  $u$  and  $v$  such that

$$p_\theta(x) = u(T(x), \theta) v(x).$$

We need to show that  $T$  is sufficient for  $\theta$ .

Consider any measurable set  $B \in \mathcal{A}$ . We aim to show that the conditional distribution of  $X$  given  $T(X) = t$  does not depend on  $\theta$ .

First, compute the conditional probability density of  $X$  given  $T(X) = t$  under  $P_\theta$ :

$$P_\theta(X \in B \mid T(X) = t) = \frac{\int_{B \cap T^{-1}(t)} p_\theta(x) d\mu(x)}{\int_{T^{-1}(t)} p_\theta(x) d\mu(x)}.$$

Substitute  $p_\theta(x) = u(T(x), \theta) v(x)$ :

$$P_\theta(X \in B \mid T(X) = t) = \frac{\int_{B \cap T^{-1}(t)} u(t, \theta) v(x) d\mu(x)}{\int_{T^{-1}(t)} u(t, \theta) v(x) d\mu(x)}.$$

Since  $u(t, \theta)$  is constant with respect to  $x$  over  $T^{-1}(t)$ , it can be factored out:

$$P_{\theta}(X \in B \mid T(X) = t) = \frac{u(t, \theta) \int_{B \cap T^{-1}(t)} v(x) d\mu(x)}{u(t, \theta) \int_{T^{-1}(t)} v(x) d\mu(x)}.$$

The  $u(t, \theta)$  terms cancel out:

$$P_{\theta}(X \in B \mid T(X) = t) = \frac{\int_{B \cap T^{-1}(t)} v(x) d\mu(x)}{\int_{T^{-1}(t)} v(x) d\mu(x)}.$$

This expression is independent of  $\theta$ , which means that the conditional distribution of  $X$  given  $T(X) = t$  does not depend on  $\theta$ . Therefore,  $T$  is sufficient for  $\theta$ .  $\square$