

Convergence Revisited

We've considered several modes of convergence of random variables:

$$X_n \rightarrow X \text{ a.s.}$$

$$X_n \rightarrow X \text{ in } L^p$$

$$X_n \rightarrow_p X$$

All of these require information about the **joint** distribution of $\{X, X_n\}$.

We're now going to turn to some convergence notions that only care about the **individual** distributions.

Total Variation

Let $\{\mu_n\}_{n=1}^{\infty}$ be a sequence of probability measures on (S, \mathcal{B}) .

They're just \mathbb{R} -valued functions on \mathcal{B} , so we can use any function convergence notion we like.

Def: Let μ, ν be probability measures on (S, \mathcal{B}) . The **total variation distance** between them is

$$d_{TV}(\mu, \nu) = \sup_{B \in \mathcal{B}} |\mu(B) - \nu(B)|$$

If X, Y are (S, \mathcal{B}) -valued random variables, we set

$$\begin{aligned} d_{TV}(X, Y) &= d_{TV}(\mu_X, \mu_Y) \\ &= \sup_{B \in \mathcal{B}} |P(X \in B) - P(Y \in B)| \end{aligned}$$

Lemma: (Scheffé) If α is a finite measure on (S, \mathcal{B}) such that $\mu, \nu \ll \alpha$ with $d\mu = u d\alpha$, $d\nu = v d\alpha$, then

$$d_{TV}(\mu, \nu) = \frac{1}{2} \|u - v\|_{L^1(\alpha)}.$$

Pf. For $B \in \mathcal{B}$,

$$|\mu(B) - \nu(B)|$$

$$|\mu(B^c) - \nu(B^c)|$$

Note: it is always possible to find such an α .

In fact: if $\{\mu_n\}_{n=1}^{\infty}$ is any countable collection of finite measures,

take
$$\alpha = \sum_{n=1}^{\infty} 2^{-n} \mu_n.$$

Cor: d_{TV} is a complete metric on $\text{Prob}(S, \mathcal{B})$.

Pf. (This can be shown directly, but the present approach is slicker.)

Let $\{\mu_n\}_{n=1}^{\infty} \subset \text{Prob}(S, \mathcal{B})$. Fix α as above; then $d\mu_n = u_n d\alpha$.

- $0 = d_{TV}(\mu_1, \mu_2)$

- $d_{TV}(\mu_1, \mu_3)$

- If $\{\mu_n\}_{n=1}^{\infty}$ is d_{TV} -Cauchy,
$$d_{TV}(\mu_n, \mu_m) \rightarrow 0$$

Cor: If h is a bounded r.v. on (S, \mathcal{B}) , then $\forall \mu, \nu \in \text{Prob}(S, \mathcal{B})$

$$\left| \int_S h d\mu - \int_S h d\nu \right| \leq 2 d_{TV}(\mu, \nu) \cdot \sup_S |h|.$$

Moreover, $d_{TV}(\mu, \nu) = \frac{1}{2} \sup \left\{ \left| \int_S h d\mu - \int_S h d\nu \right| : \sup_S |h| \leq 1 \right\}$

Pf.

Total variation works well when S is countable.

Lemma: If S is countable, and $\mu, \nu \in \text{Prob}(S, \mathcal{B})$, then

$$d_{TV}(\mu, \nu) = \frac{1}{2} \sum_{k \in S} |\mu(\{k\}) - \nu(\{k\})|$$

$\therefore \mu_n \rightarrow_{TV} \mu$ iff $\mu_n(\{k\}) \rightarrow \mu(\{k\}) \quad \forall k \in S$.

Pf.

Eg. $\mu_p \stackrel{d}{=} \text{Bernoulli}(p)$.

$$d_{TV}(\mu_p, \mu_q)$$

Eg. $\nu_\lambda \stackrel{d}{=} \text{Poisson}(\lambda)$.

$$d_{TV}(\nu_\lambda, \nu_\eta)$$

Eg. $d_{TV}(\mu_p, \nu_p)$