

## Convergence Revisited

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

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

$$P\left(\lim_{n \rightarrow \infty} X_n = X\right) = 1 \quad \leftarrow \text{need } T = \bigcap_{n \geq 1} \sigma(X_n, X_{n+1}, \dots)$$

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

$$\Omega = \lim_{n \rightarrow \infty} \|X_n - X\|_{L^p}^p = E[|X_n - X|^p] = \int_{\mathbb{R}^2} |y-x|^p \mu_{X_n, X}(dy dx)$$

$$X_n \rightarrow_p X$$

$$\Omega = \lim_{n \rightarrow \infty} P(|X_n - X| > \varepsilon) = \int_{\mathbb{R}^2} \mathbb{1}_{|y-x| > \varepsilon} \mu_{X_n, X}(dy dx)$$

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  $\mu, \nu$  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

$$d_{TV}(X, Y) = d_{TV}(\mu_X, \mu_Y)$$

$$= \sup_{B \in \mathcal{B}} |P(X \in B) - P(Y \in B)|$$

Lemma: (Scheffé) If  $\alpha$  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  $B \in \mathcal{B}$ ,

$$\begin{aligned} |\mu(B) - \nu(B)| &= \left| \int_B u d\alpha - \int_B v d\alpha \right| \leq \int_B |u - v| d\alpha \\ |\mu(B^c) - \nu(B^c)| &\leq \int_{B^c} |u - v| d\alpha \\ |(1 - \mu(B)) - (1 - \nu(B))| &= |\mu(B) - \nu(B)| \\ \Rightarrow 2|\mu(B) - \nu(B)| &\leq \int_B |u - v| d\alpha + \int_{B^c} |u - v| d\alpha = \int_S |u - v| d\alpha = \|u - v\|_{L^1(\alpha)}. \end{aligned}$$

$$A = \{u > v\}$$

$$\begin{aligned} Q = \int_S u d\alpha - \int_S v d\alpha &= \int_A (u - v) d\alpha + \int_{A^c} (u - v) d\alpha \\ &= \int_A |u - v| d\alpha - \int_{A^c} |u - v| d\alpha \\ |\mu(A) - \nu(A)| &= \int_A |u - v| d\alpha \approx \int_{A^c} |u - v| d\alpha \quad \therefore 2|\mu(A) - \nu(A)| = \int_S |u - v| d\alpha \quad // \end{aligned}$$

Note: it is always possible to find such an  $\alpha$ .

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

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

$\therefore$  by R-N thm,  $d\mu_k = u_k d\alpha$  ( $u_k \geq 0$ )  $\int u_k d\alpha = 1$ .

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  $\alpha$  as above; then  $d\mu_n = u_n d\alpha$ .

- $0 = d_{TV}(\mu_1, \mu_2) = \frac{1}{2} \int |u_1 - u_2| d\alpha \Rightarrow u_1 = u_2 \text{ a.s. } [\alpha] \Rightarrow \text{a.s. } [\mu_n] \Rightarrow \mu_1 = \mu_2$
- $d_{TV}(\mu_1, \mu_3) = \frac{1}{2} \|u_1 - u_3\|_{L^1} \leq \frac{1}{2} (\|u_1 - u_2\|_{L^1} + \|u_2 - u_3\|_{L^1}) = d_{TV}(\mu_1, \mu_2) + d_{TV}(\mu_2, \mu_3)$

- If  $\{\mu_n\}_{n=1}^{\infty}$  is  $d_{TV}$ -Cauchy,

$$\frac{1}{2} \|u_n - u_m\|_{L^1(\alpha)} = d_{TV}(\mu_n, \mu_m) \rightarrow 0 \quad \therefore \{\mu_n\}_{n=1}^{\infty} \text{ is Cauchy in } L^1(\alpha)$$

$\therefore u_n \rightarrow u \in L^1(\alpha)$  Define  $d\mu := u d\alpha$ .

$$d_{TV}(\mu_n, \mu) = \frac{1}{2} \|u_n - u\|_{L^1(\alpha)} \rightarrow 0$$

$$\left| \int u_n d\alpha - \int u d\alpha \right| \leq \int |u_n - u| d\alpha \rightarrow 0$$

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

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

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. Fix finite meas.  $\alpha$  s.t.  $d\mu = u d\alpha$ ,  $d\nu = v d\alpha$ .

$$\begin{aligned} \left| \int_S h d\mu - \int_S h d\nu \right| &= \left| \int_S (hu - hv) d\alpha \right| \leq \int_S |h| |u - v| d\alpha \\ &\leq \underbrace{\sup_S |h|}_{\text{sup } |h|} \cdot \underbrace{\int_S |u - v| d\alpha}_{2d_{TV}(\mu, \nu)}. \end{aligned}$$

$\therefore$  If  $\sup_S |h| \leq 1$ ,  $\left| \int_S h d\mu - \int_S h d\nu \right| \leq 2d_{TV}(\mu, \nu)$

Define  $h = \mathbb{1}_{u>v} - \mathbb{1}_{u \leq v} \therefore |h| \leq 1$ .

$$\begin{aligned} \int_S h d\mu - \int_S h d\nu &= \int_S (\mathbb{1}_{u>v} - \mathbb{1}_{u \leq v})(u - v) d\alpha \\ &= \int_{u>v} |u - v| d\alpha + \int_{u \leq v} |u - v| d\alpha \\ &= \int_S |u - v| d\alpha = 2d_{TV}(\mu, \nu). \end{aligned}$$

Total variation works well when  $S$  is countable.

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

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

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

Pf. Fix  $\alpha$  s.t.  $d\mu = u d\alpha$ ,  $d\nu = v d\alpha$   $\alpha_k = \alpha(\{k\})$

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

$$= \frac{1}{2} \int |u - v| d\alpha$$

$$= \frac{1}{2} \sum_{k \in S} \alpha_k |u(k) - v(k)|$$

$$= \frac{1}{2} \sum_{k \in S} |u(k) \alpha_k - v(k) \alpha_k| \\ \mu(\{k\}) \quad \nu(\{k\}).$$

$$\begin{aligned} (\Rightarrow) \quad & \sum_k |\mu_n(\{k\}) - \mu(\{k\})| \rightarrow 0 \\ & \Rightarrow \mu_n(\{k\}) \rightarrow \mu(\{k\}), \quad \checkmark \end{aligned}$$

$\Leftarrow$  L.H.S.

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

$$d_{TV}(\mu_p, \mu_q) = \frac{1}{2} \sum_{k=0}^1 |\mu_p(\{k\}) - \mu_q(\{k\})| = \frac{1}{2} (|(1-p) - (1-q)| + |p-q|) = |p-q|$$

Eg.  $\nu_\lambda \stackrel{d}{=} \text{Poisson}(\lambda)$ .  $\nu_\lambda(\{k\}) = e^{-\lambda} \frac{\lambda^k}{k!}$

$$d_{TV}(\nu_\lambda, \nu_\eta) = \sum_{k=0}^{\infty} \left| e^{-\lambda} \frac{\lambda^k}{k!} - e^{-\eta} \frac{\eta^k}{k!} \right| \leq |\lambda - \eta| \quad (\text{HW}).$$

$$\text{Eg. } d_{TV}(\mu_p, \nu_p) = \frac{1}{2} \sum_{k=0}^{\infty} |\mu_p(\{k\}) - \nu_p(\{k\})|$$

$$= \frac{1}{2} |\mu_p(0) - \nu_p(0)| + \frac{1}{2} |\mu_p(1) - \nu_p(1)| + \underbrace{\frac{1}{2} \sum_{k \geq 2} \nu_p(k)}_{\frac{1}{2} (1 - \nu_p(0) - \nu_p(1))}$$

$$= \frac{1}{2} |1-p - e^{-p}| + \frac{1}{2} |p - e^{-p}p| \rightarrow \frac{1}{2} (1 - e^{-p} - p + e^{-p}p)$$

$$= \frac{1}{2} (e^{-p} + p - 1) + \frac{1}{2} p - \frac{1}{2} e^{-p}p + \frac{1}{2} - \frac{1}{2} e^{-p} - \frac{1}{2} e^{-p}p$$

$$= p(1 - e^{-p})$$