

Convergence Revisited

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

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

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

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

$$0 = \lim_{n \rightarrow \infty} \|X_n - X\|_{L^p}^p = E[|X_n - X|^p] = \int_{\mathbb{R}^2} |y - x|^p \mu_{X_n, X}(dy dx)$$

$$X_n \rightarrow_p X$$

$$0 = \lim_{n \rightarrow \infty} P(|X_n - X| > \varepsilon) = \int_{\mathbb{R}^2} \mathbb{1}_{|y-x| > \varepsilon} \mu_{X_n, X}(dy dx)$$

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 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 $B \in \mathcal{B}$,

$$\begin{cases} |\mu(B) - \nu(B)| = \left| \int_B u d\alpha - \int_B v d\alpha \right| \leq \int_B |u - v| d\alpha \\ |\mu(B^c) - \nu(B^c)| \leq \int_{B^c} |u - v| d\alpha \end{cases}$$

$$|(1 - \mu(B)) - (1 - \nu(B))| = |\mu(B) - \nu(B)|$$

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

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

$$\begin{aligned} 0 = \int u d\alpha - \int v d\alpha &= \int_S (u - v) d\alpha = \int_A (u - v) d\alpha + \int_{A^c} (u - v) d\alpha \\ &= \int_A |u - v| d\alpha - \int_{A^c} |u - v| d\alpha \end{aligned}$$

$$|\mu(A) - \nu(A)| = \int_A |u - v| d\alpha = \int_{A^c} |u - v| d\alpha \quad \therefore 2|\mu(A) - \nu(A)| = \int_S |u - v| d\alpha \quad //$$

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

\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 α 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,

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

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

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

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

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

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

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

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

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

Total variation works well when S is countable.

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

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

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

Pf. Fix α s.t. $d\mu = u d\alpha$, $d\nu = v d\alpha$

$$\alpha_k = \alpha(\{k\})$$

$$\alpha = \sum_{k \in S} \alpha_k \delta_k.$$

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

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

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

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

$\mu(\{k\}) \quad \nu(\{k\})$

$$\Rightarrow \sum_k |\mu_n(\{k\}) - \mu(\{k\})| \rightarrow 0$$

$$\Rightarrow \mu_n(\{k\}) \rightarrow \mu(\{k\}) \quad \checkmark$$

\Leftarrow HW.

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

$$d_{TV}(\nu_\lambda, \nu_\eta) = \frac{1}{2} \sum_{k=0}^{\infty} \left| e^{-\lambda} \frac{\lambda^k}{k!} - e^{-\eta} \frac{\eta^k}{k!} \right| \leq |\lambda - \eta| \quad (\text{HW}).$$

$$\begin{aligned} \text{Eg. } d_{TV}(\mu_p, \nu_p) &= \frac{1}{2} \sum_{k=0}^{\infty} |\mu_p(\{k\}) - \nu_p(\{k\})| \\ &= \frac{1}{2} |\mu_p(0) - \nu_p(0)| + \frac{1}{2} |\mu_p(1) - \nu_p(1)| + \underbrace{\frac{1}{2} \sum_{k=2}^{\infty} \nu_p(k)}_{\frac{1}{2} (1 - \nu_p(0) - \nu_p(1))} \\ &= \frac{1}{2} |1 - p - e^{-p}| + \frac{1}{2} |p - e^{-p} p| + \frac{1}{2} (1 - \nu_p(0) - \nu_p(1)) \\ &= \frac{1}{2} (e^{-p} + p - 1) + \frac{1}{2} p - \frac{1}{2} e^{-p} p + \frac{1}{2} - \frac{1}{2} e^{-p} - \frac{1}{2} e^{-p} p \\ &= p(1 - e^{-p}). \end{aligned}$$