### STATS 217: Introduction to Stochastic Processes I

Lecture 17

#### Total variation distance

- Let  $\mu$  and  $\nu$  be two probability distributions on  $\Omega$ .
- The **total variation distance** between them, denoted by  $\mathsf{TV}(\mu,\nu)$ , is defined by

$$\mathsf{TV}(\mu, \nu) := \max_{A \subseteq \Omega} |\mu(A) - \nu(A)|.$$

• On the homework, you will show that

$$\mathsf{TV}(\mu,\nu) = \frac{1}{2} \sum_{x \in \Omega} |\mu(x) - \nu(x)|.$$

• Note that  $\mathsf{TV}(\mu, \nu)$  is a metric on the set of probability measures on  $\Omega$ .

### Total variation distance is a metric

- $\mathsf{TV}(\mu, \nu) \geq 0$  and  $\mathsf{TV}(\mu, \nu) = \mathsf{TV}(\nu, \mu)$ .
- If  $TV(\mu, \nu) = 0$ , then  $|\mu(x) \nu(x)| = 0$  for all  $x \in \Omega$  so that  $\mu \equiv \nu$ .
- Finally, TV satisfies the triangle inequality: for probability measures  $\mu, \nu, \eta$  on  $\Omega$ ,  $\mathsf{TV}(\mu, \nu) \leq \mathsf{TV}(\mu, \eta) + \mathsf{TV}(\eta, \nu)$ .
- Indeed.

$$2 \operatorname{TV}(\mu, \nu) = \sum_{x \in \Omega} |\mu(x) - \nu(x)|$$

$$= \sum_{x \in \Omega} |\mu(x) - \eta(x) + \eta(x) - \nu(x)|$$

$$\leq \sum_{x \in \Omega} |\mu(x) - \eta(x)| + \sum_{x \in \Omega} |\eta(x) - \nu(x)|$$

$$= 2 \operatorname{TV}(\mu, \eta) + 2 \operatorname{TV}(\eta, \nu).$$

### Dual characterization of total variation distance

- Let  $\mu, \nu$  be probability measures on  $\Omega$ . Let  $\mathcal{F}$  denote the collection of all functions  $f: \Omega \to \mathbb{R}$  satisfying  $\max_{x \in \Omega} |f(x)| < 1$ .
- Then,

$$\mathsf{TV}(\mu,\nu) = \frac{1}{2} \sup_{f \in \mathcal{F}} \left\{ \sum_{x \in \Omega} f(x) \mu(x) - \sum_{x \in \Omega} f(x) \nu(x) \right\}.$$

• Why? For any  $f \in \mathcal{F}$ ,

$$\begin{aligned} |\sum_{x \in \Omega} f(x)\mu(x) - \sum_{x \in \Omega} f(x)\nu(x)| &\leq \sum_{x \in \Omega} |f(x)||\mu(x) - \nu(x)| \\ &\leq \sum_{x \in \Omega} |\mu(x) - \nu(x)|. \end{aligned}$$

• You will prove the reverse inequality on the homework.

# Coupling

- Let  $\mu$  and  $\nu$  be two probability measures on  $\Omega_1$  and  $\Omega_2$  respectively.
- A coupling of  $\mu$  and  $\nu$  is a probability measure  $\gamma$  on  $\Omega_1 \times \Omega_2$  such that

$$\gamma(A \times \Omega_2) = \mu(A) \quad \forall A \subseteq \Omega_1 \text{ and}$$
  
 $\gamma(\Omega_1 \times B) = \nu(B) \quad \forall B \subseteq \Omega_2.$ 

• Similarly, a coupling of random variables  $X:\Omega_1'\to\Omega_1$  and  $Y:\Omega_2'\to\Omega_2$  is a pair of random variables  $\widehat{X}:\Omega\to\Omega_1$  and  $\widehat{Y}:\Omega\to\Omega_2$  defined on a common probability space  $\Omega$  such that

$$\mathbb{P}[\widehat{X} = x] = \mathbb{P}[X = x] \quad \forall x \in \Omega_1 \text{ and}$$
  $\mathbb{P}[\widehat{Y} = y] = \mathbb{P}[Y = y] \quad \forall y \in \Omega_2.$ 

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### Example: independent coupling

- Let  $X \sim \text{Ber}(p)$  and  $Y \sim \text{Ber}(q)$  where  $0 \le p \le q \le 1$ .
- ullet Formally, we can think of  $X:[0,1] \to \{0,1\}$  where [0,1] is equipped with the uniform measure and

$$X(r) = \begin{cases} 0 \text{ if } r \leq 1 - p \\ 1 \text{ if } r > 1 - p. \end{cases}$$

- Y admits a similar interpretation.
- An obvious coupling of X and Y is the **independent coupling** i.e.,  $\Omega = [0,1] \times [0,1]$  equipped with the uniform measure,

$$\widehat{X}(r_1,r_2) = \begin{cases} 0 \text{ if } r_1 \leq 1-p \\ 1 \text{ if } r_1 \geq 1-p, \end{cases}$$

and similarly for  $\widehat{Y}$  (with  $r_1$  replaced by  $r_2$  and p replaced by q).

### Example: monotone coupling

A particularly useful coupling in this case is the monotone coupling.

•  $\Omega = [0,1]$  with the uniform measure.

$$\widehat{X}(r) = \begin{cases} 0 \text{ if } r \leq 1 - p \\ 1 \text{ if } r > 1 - p. \end{cases}$$

$$\widehat{Y}(r) = \begin{cases} 0 \text{ if } r \leq 1 - q \\ 1 \text{ if } r > 1 - q. \end{cases}$$

• Then,

$$\mathbb{P}[\hat{X} = 1] = \mathbb{P}[r > 1 - p] = p = \mathbb{P}[X = 1]$$

and similarly for  $\widehat{Y}$ .

ullet The name monotone coupling comes from the observation that if  $p \leq q$ , then

$$\widehat{X} \leq \widehat{Y}$$
 determistically.

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# An application of monotone coupling

- Let  $(P_t)_{t\geq 0}$  denote a simple random walk starting from 0 where the probability of taking a step to the right is p.
- Let  $(Q_t)_{t\geq 0}$  denote a simple random walk starting from 0 where the probability of taking a step to the right is q.
- Then, if  $p \leq q$ , intuitively,

$$\mathbb{P}[Q_t \leq z] \leq \mathbb{P}[P_t \leq z] \quad \forall t \geq 0, z \in \mathbb{Z}.$$

• Monotone coupling lets us see this directly.

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# An application of monotone coupling

- Let  $(\widehat{X}, \widehat{Y})$  denote the monotone coupling of Ber(p) and Ber(q).
- Let  $(\widehat{X}_t, \widehat{Y}_t)_{t \geq 1}$  denote iid copies of  $(\widehat{X}, \widehat{Y})$ .
- Let  $\widehat{P}_t = \sum_{i=1}^t (2\widehat{X}_i 1)$  and  $\widehat{Q}_t = \sum_{i=1}^t (2\widehat{Y}_i 1)$ .
- Then, by construction,  $\widehat{P}_t \leq \widehat{Q}_t$  for all t.
- Moreover,  $P_t \sim \widehat{P}_t$  and  $Q_t \sim \widehat{Q}_t$  for all t.
- So, for any  $z \in \mathbb{Z}$  and any  $t \ge 0$ ,

$$\mathbb{P}[Q_t \leq z] = \mathbb{P}[\widehat{Q}_t \leq z] \leq \mathbb{P}[\widehat{P}_t \leq z] = \mathbb{P}[P_t = z].$$

# Coupling and total variation

- Let  $\mu$  and  $\nu$  be two probability distributions on  $\Omega$ .
- The coupling lemma asserts that

$$\mathsf{TV}(\mu,\nu) = \inf\{\mathbb{P}[X \neq Y] : (X,Y) \text{ is a coupling of } \mu \text{ and } \nu\}.$$

- Example: let  $\mu = \mathsf{Ber}(p)$  and  $\nu = \mathsf{Ber}(q)$  with  $0 \le p \le q \le 1$ .
- Then, by direct computation,

$$\mathsf{TV}(\mathsf{Ber}(p),\mathsf{Ber}(q)) = rac{1}{2}(|q-p|+|1-q-1+p|) = q-p.$$

• For the monotone coupling  $(\widehat{X}, \widehat{Y})$ , we have

$$\mathbb{P}[\widehat{X} \neq \widehat{Y}] = \mathbb{P}[1 - q \le r \le 1 - p] = q - p.$$

 The above characterization shows that the monotone coupling is an optimal coupling.

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# Coupling and total variation

- $\mathsf{TV}(\mu, \nu) = \inf\{\mathbb{P}[X \neq Y] : (X, Y) \text{ is a coupling of } \mu \text{ and } \nu\}.$
- Easy direction: ≤. Why?
- Let (X, Y) be any coupling of  $\mu, \nu$ . Let  $A \subseteq \Omega$ . Then,

$$\mu(A) - \nu(A) = \mathbb{P}[X \in A] - \mathbb{P}[Y \in A]$$

$$= \mathbb{P}[X \in A] - \mathbb{P}[X \in A, Y \in A] - \mathbb{P}[X \notin A, Y \in A]$$

$$= \mathbb{P}[X \in A, Y \notin A] - \mathbb{P}[X \notin A, Y \in A]$$

$$\leq \mathbb{P}[X \in A, Y \notin A]$$

$$\leq \mathbb{P}[X \neq Y].$$

• The reverse inequality is a starred problem on HW7.