

We've seen that weak convergence is weaker than a.s. convergence,  $L^p$  convergence, convergence in probability. There is a connection to a.s. convergence that is useful in many contexts, however.

Theorem: (Skorohod)

Let  $S$  be a separable metric space, and  $\mu_n, \mu \in \text{Prob}(S, \mathcal{B}(S))$ . If  $\mu_n \rightarrow_w \mu$ , then there exists a probability space  $(\Omega, \mathcal{F}, P)$  and random variables  $Y_n, Y : (\Omega, \mathcal{F}) \rightarrow (S, \mathcal{B}(S))$  with  $Y_n^* P = \mu_{Y_n} = \mu_n$ ,  $Y^* P = \mu_Y = \mu$ , and  $Y_n \rightarrow Y$  a.s.

(The proof is quite involved. The probability space can be chosen to be  $\Omega = (0, 1) \times S^{\mathbb{N}}$ ,  $\mathcal{F} = \mathcal{B}((0, 1) \times S^{\mathbb{N}})$ , and  $P$  a well-chosen infinite product measure (that introduces lots of complicated correlations between the  $Y_n$ s).

We'll prove Skorohod's Theorem in the case  $S = \mathbb{R}$ .

## "Inverting" the CDF

Suppose  $F: \mathbb{R} \rightarrow [0, 1]$  is a CDF

If  $F$  is strictly increasing,

meaning  $\mu(a, b) > 0 \forall a < b$ , then  $F$  is an invertible function,  
 $F^{-1}: (0, 1) \rightarrow \mathbb{R}$

Equip  $\mathbb{R}$  with Lebesgue measure  $\lambda$ ; then  $F^{-1}$  becomes a random variable.

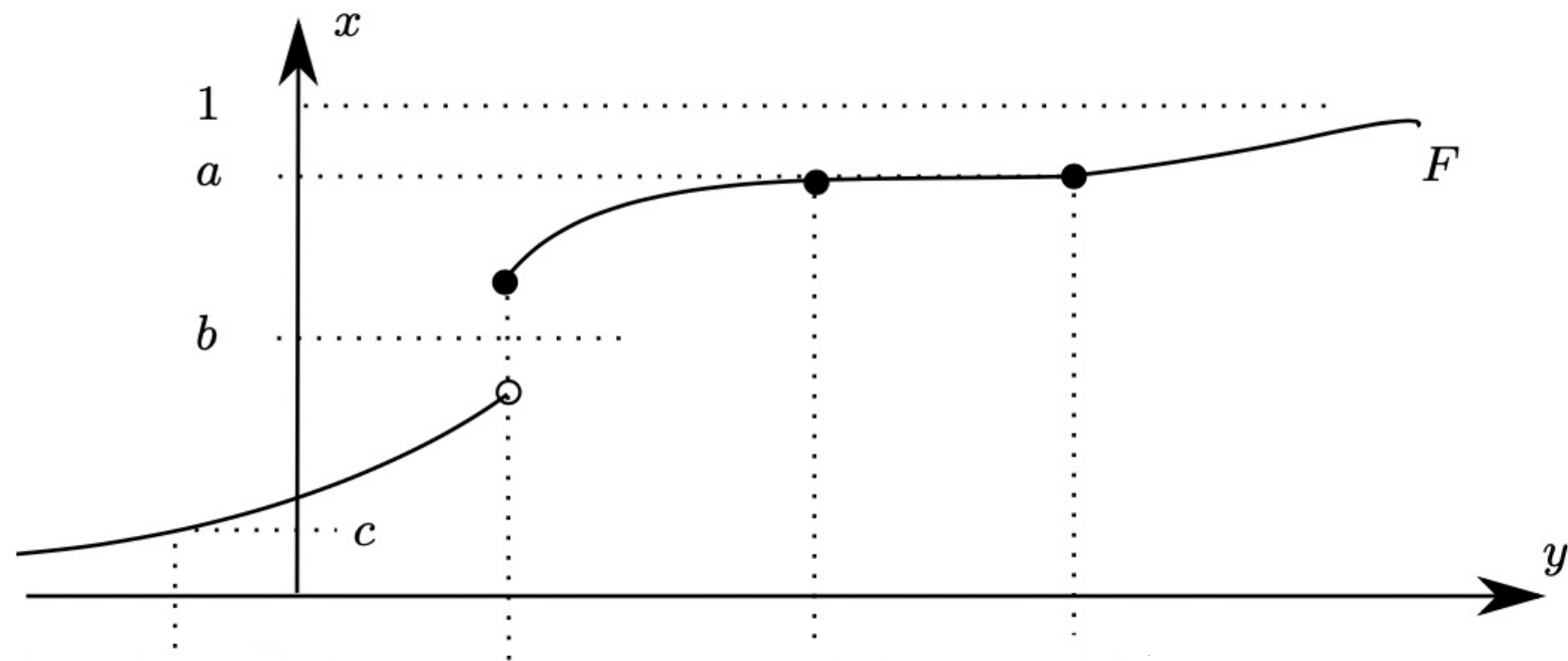
Thus,  $\mu_Y =$

It turns out we can make this work even if  $F$  is not strictly increasing.

Def. Let  $F: \mathbb{R} \rightarrow [0, 1]$  be a CDF. Define

$$F^{\leftarrow}: (0, 1) \rightarrow \mathbb{R} \quad F^{\leftarrow}(x) := \sup \{y \in \mathbb{R} : F(y) < x\}$$

$$F^{\rightarrow}: (0, 1) \rightarrow \mathbb{R} \quad F^{\rightarrow}(x) := \inf \{y \in \mathbb{R} : F(y) > x\}$$



1.  $F^{\leftarrow}(x) \leq F^{\rightarrow}(x)$   
 $<$  iff

2.  $E = \{x \in (0, 1) : F^{\leftarrow}(x) < F^{\rightarrow}(x)\}$  is countable.

3.  $\lim_{y \uparrow F^{\leftarrow}(x)} F(y) \leq x \leq F(F^{\leftarrow}(x)) \quad \forall x \in (0, 1)$

Lemma: Let  $\mu \in \text{Prob}(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  with CDF  $F_\mu = F$ .

Define  $Y: (0,1) \rightarrow \mathbb{R}$  to be  $Y = F^\leftarrow$

$$Y(x) = \sup \{ y \in \mathbb{R} : F(y) < x \}$$

Then  $Y: (0,1), \mathcal{B}(0,1) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R}))$  is measurable,  
and wrt Lebesgue measure  $\lambda$  on  $\mathbb{R}$ ,

$$Y^* \lambda = \mu_Y = \mu.$$

Pf. For  $t \in \mathbb{R}$ , suppose  $Y(x) \leq t$ . Then  $x \leq F(Y(x))$

In fact:  $\{ x \in (0,1) : Y(x) \leq t \} =$

Cor: If  $\mu \in \text{Prob}(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  with  $F_\mu = F$ , and  $U \stackrel{d}{=} \text{Unif}([0,1])$ ,  
then  $F^\leftarrow(U) \stackrel{d}{=} \mu$ .

## Baby Skorohod Theorem

Let  $\mu_n, \mu \in \text{Prob}(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  with  $\mu_n \rightarrow_w \mu$ .

Let  $Y_n, Y$  be the random variables

$$Y_n = F_{\mu_n}^{\leftarrow}, \quad Y = F_{\mu}^{\leftarrow}$$

on  $([0,1], \mathcal{B}([0,1]), \lambda)$ . Then  $Y_n \stackrel{d}{=} \mu_n$ ,  $Y \stackrel{d}{=} \mu$ , and  $Y_n \rightarrow Y$  a.s.

Pf. To complete the proof, we will show that  $Y_n(x) \rightarrow Y(x)$

for  $x \notin E = \{t \in (0,1) : F_{\mu}^{\leftarrow}(t) < F_{\mu}^{\rightarrow}(t)\}$

• If  $y \in \text{Cont}(F_{\mu})$  with  $y < Y(x)$

• If  $y \in \text{Cont}(F_{\mu})$  with  $y > Y(x) = F_{\mu}^{\leftarrow}(x) = F_{\mu}^{\rightarrow}(x)$

Cor: (Continuous mapping theorem)

Let  $f: \mathbb{R} \rightarrow \mathbb{R}$  be Borel measurable.

Let  $X_n \rightarrow_w X$ , and suppose  $P(X \in \text{Disc}(f)) = 0$ .

Then  $f(X_n) \rightarrow_w f(X)$ . If in addition  $f$  is bounded, then  $\mathbb{E}[f(X_n)] \rightarrow \mathbb{E}[f(X)]$ .

Pf. Replace  $X_n, X$  with  $Y_n, Y$  on a probability space where  $Y_n \stackrel{d}{=} X_n$ ,  $Y \stackrel{d}{=} X$ , and  $Y_n \rightarrow Y$  a.s.

Let  $g \in C_b(\mathbb{R})$ ; then