

## \*Random Variables

Friday, September 1, 2017 1:14 PM

- Measurable map: [KS Def 1.10]

$$f: (\Omega, \mathcal{F}) \rightarrow (S, \mathcal{S})$$

such that  $\forall A \in \mathcal{S}, f^{-1}(A) \in \mathcal{F}$ .

- A random variable is a measurable map on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$

$$X: (\Omega, \mathcal{F}) \rightarrow (\mathbb{R}, \mathcal{B})$$

$$X: (\Omega, \mathcal{F}) \rightarrow (\mathbb{R}^d, \mathcal{B}^d) \text{ random vector}$$

\* Could be other set,  $[0, 1]$ , graphs e.t.c.

We write  $\mathbb{P}[X \in A] := \mathbb{P}(\{X^{-1}(A)\})$

Lemma [D Thm 1.3.1]

- If  $f: (\Omega, \mathcal{F}) \rightarrow (S, \mathcal{S})$  and  $\mathcal{A} \subset \mathcal{S}$  generates  $\mathcal{S}$ ,  $\forall A \in \mathcal{A} \{X \in A\} \in \mathcal{F}$  then  $f$  is measurable.

Pf:  $\mathcal{D} = \{D: \{X \in D\} \in \mathcal{F}\}$

$$\{X \in \bigcup_i D_i\} = \bigcup_i \{X \in D_i\} \in \mathcal{F}$$

$$\{X \in D^c\} = \{X \in D\}^c \in \mathcal{F}.$$

So  $\mathcal{D}$   $\sigma$ -algebra,  $\mathcal{F} = \sigma(A) \subseteq \mathcal{D}$ .

$\Rightarrow f$  is measurable

Lem [D Thm 1.3.2]

If  $f: (\Omega, \mathcal{F}) \rightarrow (S, \mathcal{S})$ ,  $g: (S, \mathcal{S}) \rightarrow (S', \mathcal{S}')$   
measurable then  $g(f(\omega))$  is measurable.  
[D Thm 1.3.3/1.3.4]

Lemma: If  $X_1, \dots, X_n$  are random variables  
on  $(\Omega, \mathcal{F}, \mathbb{P})$  then so are

$X_1 + X_2, X_1 X_2, e^{X_1} + 2 \cos(X_2), \dots$  etc.

P.f.  $X_1, \dots, X_n$  is measurable since

$$\begin{aligned} & \{ (X_1, \dots, X_n) \in [a_1, b_1] \times \dots \times [a_n, b_n] \} \\ &= \bigcap_{i=1}^n \{ X_i \in [a_i, b_i] \} \in \mathcal{F}. \end{aligned}$$

$g(x_1, x_2) = x_1 + x_2$  is measurable

check  $\{ g^{-1}((-\infty, a]) \} = \{ (x_1, x_2) : x_1 + x_2 \leq a \}$   
open and so in  $\mathcal{B}(\mathbb{R}^2)$ .

Almost any function you can think of  
on  $\mathbb{R}^n$  is measurable.

[D Thm 1.3.5]

Lemma: If  $X_1, \dots$  are R.V.

$\inf_n X_n, \sup_n X_n, \liminf X_n, \limsup X_n$   
are random variables.

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\* Review: sup or supremum is like the maximum, but it may not be achieved

$\sup_n x_n$  is smallest  $a$  such that  
 $\forall n \quad x_n \leq a.$

$\inf_n x_n$  is largest  $a$  s.t.  $\forall n \quad x_n \geq a$

Pf:  $\{\sup_n x_n < a\} = \bigcap_{n=1}^{\infty} \{x_n < a\} \in \mathcal{F}.$

$$\limsup_n x_n := \lim_{n \rightarrow \infty} \sup_{m \geq n} x_m$$

$$= \inf_n \sup_{m \geq n} x_m \quad \text{measurable.}$$

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Distributions

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## Discrete

Uniform  $\{1, \dots, n\}$

$$P[X=k] = 1/n$$

Binomial  $\text{Bin}(n, p)$

$n$  trials, success  $p$

$$P[X=k]$$

$$= \binom{n}{k} p^k (1-p)^{n-k}$$

$0 \leq k \leq n$

Geometric  $\text{Geom}(p)$

$$P[X=k] = (1-p)^k p$$

Poisson  $\text{Pois}(\lambda)$

$$P[X=k] = \frac{\lambda^k e^{-\lambda}}{k!}$$

## Continuous

Uniform  $U[a, b]$

$$\text{Density } f(x) = \begin{cases} \frac{1}{b-a} & a < x < b \\ 0 & \text{o.w.} \end{cases}$$

Normal / Gaussian

$$N(\mu, \sigma^2)$$

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

Exponential  $\text{Exp}(\lambda)$

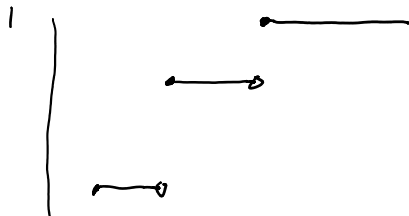
$$f(x) = \lambda e^{-\lambda x}$$

The distribution function of  $X$  is

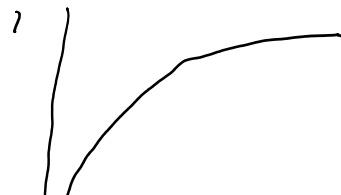
$$F(x) = F_X(x) = P[X \leq x]$$

\* Defined for any R.V.

Discrete



Continuous





Properties: [D Thm 1.2.1]

- i) Increasing:  $x < y \Rightarrow F(x) \leq F(y)$
  - ii)  $\lim_{x \rightarrow -\infty} F(x) = 0, \quad \lim_{x \rightarrow \infty} F(x) = 1$
  - iii) Right Continuous:  $\lim_{y \downarrow x} F(y) = F(x)$
  - iv)  $F(x-) := \lim_{y \uparrow x} F(y) = \mathbb{P}[X < x]$
  - v)  $\mathbb{P}[X = x] = F(x) - F(x-)$ .
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Lemma: If  $F(x)$  satisfies (i)–(iii) then it is the distribution function of a R.V.

Pf: 2 Proof

[KS Sec 3.2]

- a) The function  $\mu((a, b]) = F(b) - F(a)$  is additive so by Caratheodory extends to a measure  $\mu$  on  $(\mathbb{R}, \mathcal{B})$ .

We say that  $F$  induces this measure on  $\mathbb{R}$ .

$$\bullet \mu(\mathbb{R}) = \lim_{\substack{a \rightarrow -\infty \\ b \rightarrow \infty}} F(b) - F(a) = 1.$$

Let  $X(x) = x$  be identity on  $\mathbb{R}$ .

$$\mathbb{P}[X \leq x] =$$

$$\mu(X \leq x) = \lim_{a \rightarrow 0} F(x) - F(a) = F(x),$$

[D Thm 1.2.2]

b) For Lebesgue measure on  $[0, 1]$ ,

$U(x) = x$  is a uniform R.V.

Let  $X(x) = \sup \{y : F(y) \leq x\}$ .

Homework: Show that  $F$  is the distribution function of  $X$ .

\* This is useful as an algorithm to construct a R.V. with distribution  $F$ .

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- If  $X, X'$  have the same distribution function we say they are identically distributed and write  $X \stackrel{d}{=} X'$ .

Then  $\mathbb{P}[X \in A] = \mathbb{P}[X' \in A]$ .

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Distributions from densities

If  $f: \mathbb{R} \rightarrow \mathbb{R}$  is non-negative, integrable &

If  $f: \mathbb{R} \rightarrow \mathbb{R}$  is non-negative, integrable &  
 $\int_{\mathbb{R}} f(x) dx = 1$

Then

$F(x) = \int_{-\infty}^x f(y) dy$  is a distribution function.

Ex: Exp(1)  $f(x) = \begin{cases} e^{-x} & x > 0 \\ 0 & x \leq 0 \end{cases}$

$$F(x) = \int_0^x e^{-x} dx = 1 - e^{-x} \quad \text{when } x \geq 0.$$

Properties: •  $F'(x) = f(x)$

- $F$  is continuous
- We interpret  $f(x) \cdot dx \approx \mathbb{P}[X \in (x, x+dx)]$
- $\mathbb{P}[X \in A] = \int_A f(x) dx = \int f(x) I(x \in A) dx$