

Ross ch 3.4 Independent Events

Events E, F are indep if $P(E|F) = P(E)$

The knowledge that F occurred has no effect on $P(E)$.

$$\frac{P(E \cap F)}{P(F)}$$

$$\Rightarrow P(E \cap F) = P(E)P(F)$$

They take this as the def of 2 indep events.

4a) Pick a card at random.

$E :=$ event card is ace $P(E) = \frac{4}{52} = \frac{1}{13}$

$F :=$ event card is a spade $P(F) = \frac{13}{52} = \frac{1}{4}$

$P(E \cap F) = P(\text{card is ace \& card is spade}) = P(\text{ace of spades}) = \frac{1}{52}$

whereas $P(E) \cdot P(F) = \frac{1}{13} \cdot \frac{1}{4} = \frac{1}{52}$ ← same ↑ so events qualify as indep.

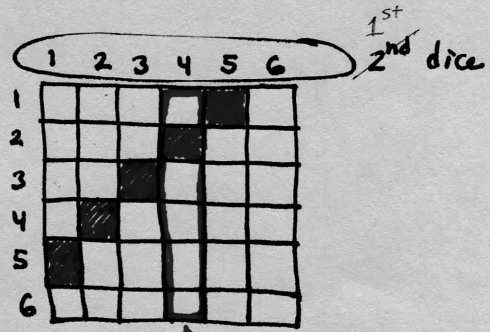
4e) Toss 2 dice $E =$ sum is 6

$F =$ 1st dice is a 4

There is only one way to sum to 6 if 1st dice is a 4. That is $\langle 4, 2 \rangle$.

$P(E \cap F) = P(\langle 4, 2 \rangle) = \frac{1}{36}$

But $P(E) = \frac{5}{36}$
 $P(F) = \frac{1}{6} = \frac{1}{6}$ $P(E) \cdot P(F) = \frac{5}{36} \cdot \frac{1}{6} = \frac{5}{216}$



↑ F restricts us to this col.

Prop 4.1 $E \& F$ indep $\Rightarrow E \& F^c$ are indep

pf. $E = (E \cap F) \cup (E \cap F^c)$
 $P(E) = P(E \cap F) + P(E \cap F^c) = P(E)P(F) + P(E \cap F^c)$ since $E \& F$ indep
 $P(E) - P(E)P(F) = P(E \cap F^c) \Rightarrow P(E)[1 - P(F)] = P(E \cap F^c)$
 $\underbrace{[1 - P(F)]}_{P(F^c)}$ □

Now consider 3 Events. E indep of F , F indep of G .
 Is E indep of $F \cap G$? NOT NECESSARILY.

4e) Toss 2 dice again
 $E :=$ event the sum is 7
 $F :=$ set of outcomes $\{(4, x) \mid x \text{ is any number on } 2^{\text{nd}}\}$
 $G := \{(y, 3) \mid y \text{ any number on } 1^{\text{st}}\}$
 E not indep of $F \cap G$ since $P(E | F \cap G) = 1$

$F \cap G = \{(4, 3)\}$
 This forces E to be true →

Def E, F, G are indep if:

- $P(E \cap F) = P(E) \cdot P(F)$
- $P(E \cap G) = P(E) \cdot P(G)$
- $P(F \cap G) = P(F) \cdot P(G)$
- $P(E \cap F \cap G) = P(E) \cdot P(F) \cdot P(G)$

and analogously for n events.

example 3g

Trial of 100 people

Medicine considered to work if 65% of people show improvement.

H_0 : Medicine has no effect. $p = 1/2$

$X = \#$ of people with improved cholesterol.

This is not a good example. Many better ones in Schuam's.

Prob of false rejection of H_0

Given that $p = 1/2$

$$P(X \geq 65) = \sum_{i=65}^{100} \binom{100}{i} \left(\frac{1}{2}\right)^i \left(\frac{1}{2}\right)^{100-i} = \left(\frac{1}{2}\right)^{100} \sum_{i=65}^{100} \binom{100}{i}$$

$\mu = Np = 100 \cdot \frac{1}{2} = 50 > 10$ so we can apply \mathcal{N} approx.

$$\sigma := \sqrt{Np(1-p)} = \sqrt{100 \cdot \frac{1}{2} \cdot \frac{1}{2}} = \sqrt{25} = 5$$

$$65 \leq X$$

Normalize

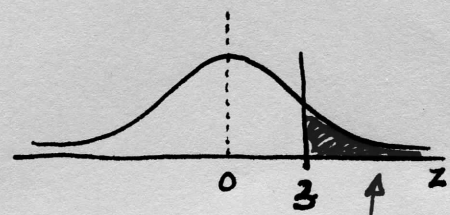
$$\frac{65 - 50}{5} \leq \frac{X - \mu}{\sigma}$$

$$\frac{15}{5} \leq Z$$

$$3 \leq Z$$

look in table p.158

$$\begin{aligned} P(Z \leq 3) &= 0.9987 \\ &= 1 - P(Z \leq -3) \\ &= 1 - 0.0013 \\ &= 0.9987 \end{aligned}$$



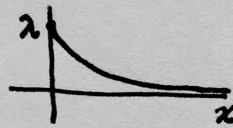
area = 0.0013

This is the prob " α " of a false rejection of H_0 :
the prob that the test results indicate we should reject H_0 when in fact H_0 is true.

Exponential RVs

pdf For $\lambda > 0$ define

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & x \text{ neg} \end{cases}$$



$$u = -\lambda x \\ du = -\lambda dx$$

cdf

$$F(a) = P(X \leq a) = \int_{-\infty}^a \lambda e^{-\lambda x} dx = \int_0^a \lambda e^{-\lambda x} dx = - \int_{u=0}^{-\lambda a} e^u du \\ = e^u \Big|_{-\lambda a}^0 = 1 - e^{-\lambda a}$$

ex 4a

Exponential RV usually represents the amount of time until some event occurs - with the 'memoryless' property that just because I waited a time t without it happening has no effect on the probability that it will happen in the next time interval (of length s , say). More on memoryless later.

Let $X =$ time in minutes until phone rings and say $\lambda = 1/10$

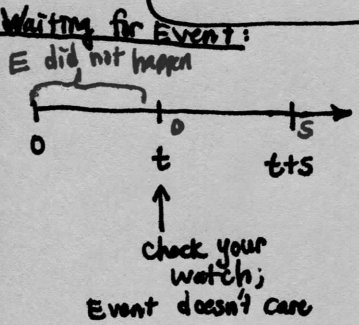
Find $P(X > 10) = \int_{10}^{\infty} \frac{1}{10} e^{-1/10 x} dx = - \int_{u=-10\lambda}^{-\infty} e^u du = \int_{-\infty}^{-10\lambda} e^u du = e^{-10\lambda} = e^{-1}$

$P(10 < X < 20) = \int_{-10\lambda}^{-20\lambda} e^u du = e^{-1} - e^{-2}$ \square

Memoryless cond

want: $P(X > t+s | X > t) = P(X > s)$

Geometric RVs are the only discrete ones with memoryless prop



$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad \text{so}$$

$$P(X > t+s | X > t) = \frac{P(X > t+s \&\& X > t)}{P(X > t)} \stackrel{X > t+s \Rightarrow X > t}{=} \frac{P(X > t+s)}{P(X > t)}$$

$$= \frac{1 - F(t+s)}{1 - F(t)} = \frac{1 - (1 - e^{-\lambda(t+s)})}{1 - (1 - e^{-\lambda t})} \\ = \frac{e^{-\lambda(t+s)}}{e^{-\lambda t}} = e^{-\lambda s} = P(X > s) \quad \square$$

Thus $\frac{P(X > t+s)}{P(X > t)} = P(X > s)$

$$P(X > t+s) = P(X > t) P(X > s)$$

This works: $e^{-\lambda(t+s)} = e^{-\lambda t} e^{-\lambda s}$

What other fns could satisfy this eq? See next page \rightarrow

We just saw the memoryless cond was equivalent to:

$$P(X > t+s) = P(X > t) \cdot P(X > s) \quad \text{Let } G(u) := P(X > u)$$

$$\Rightarrow G(t+s) = G(t) \cdot G(s)$$

What continuous fns could work here? [Ross p.166 says 'Right continuous']

Observe $G(\frac{2}{n}) = G(\frac{1}{n} + \frac{1}{n}) = G(\frac{1}{n}) \cdot G(\frac{1}{n}) = G^2(\frac{1}{n})$

then $G(1) = G(\frac{n}{n}) = G^n(\frac{1}{n}) \Rightarrow G(\frac{1}{n}) = G(1)^{\frac{1}{n}}$

then $G(\frac{m}{n}) = [G(1)]^{\frac{m}{n}}$

Any $x \in \mathbb{R}$ can be approximated by a seq of rationals $(\frac{m_i}{n_i}) \rightarrow x$

$$G(x) = G(\lim_i \frac{m_i}{n_i}) = \lim_i G(\frac{m_i}{n_i}) = \lim_i [G(1)]^{\frac{m_i}{n_i}} = G(1)^x$$

Thus $G(x) = G(1)^x \xrightarrow{G \text{ cont}} G(x) = a^x$ where $a = G(1)$ and $a \geq 0$ because $G(1) = G(\frac{1}{2})^2 \geq 0$

We know $a = e^{\ln(a)}$
 $a^x = e^{\ln(a)x}$ Define $\lambda = -\ln(a)$ $\lambda > 0$ if $a \in (0,1)$
 $a^x = e^{-\lambda x}$

$\Rightarrow G(x) = e^{-\lambda x}$ and since $G(x) = P(X > x)$
 the cdf is $F(x) = P(X \leq x) = 1 - P(X > x) = 1 - G(x) = 1 - e^{-\lambda x}$

(p. No 9) Hazard Rate fns ^{time of failure}

RV $X =$ lifetime of a component, let cdf be F and f is pdf.

Failure rate $\lambda(t) := \frac{f(t)}{1-F(t)}$

Say component has survived for t hours. what is prob it will fail in the next time increment dt ?

$$P(X \in (t, t+dt) | X > t) = \frac{P(X \in (t, t+dt))}{P(X > t)} = \frac{f(t)dt}{1-F(t)} = \lambda(t)dt$$

Now if $X \sim \exp(\lambda)$, previous survival time should make no difference

$$\lambda(t) = \frac{f(t)}{1-F(t)} = \frac{\lambda e^{-\lambda t}}{1-[1-e^{-\lambda t}]} = \frac{\lambda e^{-\lambda t}}{e^{-\lambda t}} = \lambda \text{ const. } \checkmark$$

Thm Failure rate fcn $\lambda(t)$ uniquely determines cdf F

pf we know $\lambda(t) = \frac{f(t)}{1-F(t)} \Rightarrow \frac{1}{1-F} dF = \lambda dt$ [$u = 1-F$, $du = -dF$]

$$\ln(1-F) = -\int_0^t \lambda(t)dt + k$$

plug in $t=0$ $F(0)=0$

$$\ln(1) = \int_0^0 + k \Rightarrow k=0$$

$$1-F(t) = e^{-\int_0^t \lambda(t)dt}$$

$$\Rightarrow F(t) = 1 - e^{-\int_0^t \lambda(t)dt} \quad \square$$

1.170 Gamma Distribution
 $X \sim \Gamma(t, \lambda)$ $\lambda > 0$
 $t > 0$

$$f(x) = \begin{cases} \frac{\lambda e^{-\lambda x} (\lambda x)^{t-1}}{\Gamma(t)} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

where $\Gamma(t) = \int_0^\infty e^{-y} y^{t-1} dy$
Gamma fcn

Assuming technical conds of p.129, the amount of time one has to wait for n events is given by $\Gamma(n, \lambda)$.

Using integration by parts, we can show $\Gamma(t) = (t-1)\Gamma(t-1)$
thus $\Gamma(n) = (n-1)!$ $\Gamma(1) = \int_0^\infty e^{-x} dx = 1$

$\Gamma(1/2, 1/2)$ is $\chi^2_{(1)}$ distrib

▷ Weibull distrib failure time of complex system with weakest link

$$X \sim W(n, \alpha, \beta)$$

$$F(x) = \begin{cases} 0 & x \leq n \\ 1 - e^{-\left(\frac{x-n}{\alpha}\right)^\beta} & x > n \end{cases}$$

$$f(x) = \begin{cases} 0 & x \leq n \\ \frac{\beta}{\alpha} \left(\frac{x-n}{\alpha}\right)^{\beta-1} e^{-\left(\frac{x-n}{\alpha}\right)^\beta} & x > n \end{cases}$$

▷ Cauchy distrib

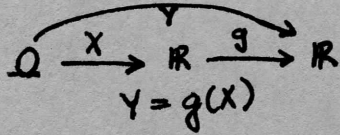
$$f(x) = \frac{1}{\pi} \frac{1}{[1+(x-\theta)^2]}$$

▷ Beta distrib

$$f(x) = \begin{cases} \frac{1}{B(a,b)} x^{a-1} (1-x)^{b-1} & 0 < x < 1 \\ 0 & \text{elsewhere} \end{cases}$$

where $B(a,b) = \int_0^1 x^{a-1} (1-x)^{b-1} dx$

Functions of a RV



- The basic steps are:
- (1) Find cdf of transformed var.
 - (2) Differentiate to get pdf
 - (3) specify region where result holds.

Let $Y = aX + b$ where $a > 0$

$$F_Y(y) := P(Y \leq y) = P(aX + b \leq y) = P(X \leq \frac{y-b}{a}) = F_X(\frac{y-b}{a})$$

A priori, $f_Y(y) = \frac{d}{dy} F_Y(y)$ and we have $F_X(s) = \int_{-\infty}^s f_X(t) dt$ Let $s = s(y) = \frac{y-b}{a}$

$$f_Y(y) = \frac{d}{dy} \int_{-\infty}^{s(y)} f_X(t) dt \xrightarrow{\text{Leibniz Integration rule}} f_X(s(y)) s'(y) = \frac{1}{a} f_X(\frac{y-b}{a})$$

Holds for any X.

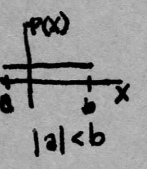
In Rice's example, take $X \sim \mathcal{N}(\mu, \sigma)$

$$f_Y(y) = \frac{1}{a} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{y-b}{a} - \mu \right)^2 / \sigma^2}$$

so we see $Y \sim \mathcal{N}(\mu a + b, a\sigma)$ just like on sheet 14

For $a < 0$, we'd get $P(X \geq \frac{y-b}{a}) = 1 - P(X < \frac{y-b}{a})$

Let $X \sim \text{Unif}(a, b)$ $Y = X^2$



$F_Y(y) = P(X^2 \leq y)$ if $y < 0$, no values of $X \in \mathbb{R}$ satisfy $X^2 \leq y$ so $P(X^2 \leq y) = 0$
if $y \geq 0$ $X^2 \leq y \iff |X| \leq \sqrt{y} \iff -\sqrt{y} \leq X \leq \sqrt{y}$ (X may be neg, not X^2)

$$F_Y(y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq \sqrt{y}) = F_X(\sqrt{y}) - F_X(-\sqrt{y}) = \int_a^{\sqrt{y}} \frac{1}{b-a} dx - \int_a^{-\sqrt{y}} \frac{1}{b-a} dx$$

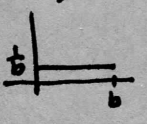
(provided $\sqrt{y} < b$)

$$= \frac{1}{b-a} (\sqrt{y} - a - (-\sqrt{y} - a)) = \frac{2}{b-a} \sqrt{y}$$

Thus

$$F_Y(y) = \begin{cases} 0 & y < 0 \\ \frac{2}{b-a} \sqrt{y} & y < a^2 < b^2 \\ \frac{1}{b-a} (\sqrt{y} - a) & a^2 < y < b^2 \\ 1 & y > b \end{cases} \implies f_Y(y) = \begin{cases} 0 & \\ \frac{1}{b-a} y^{-1/2} & \\ 0 & \end{cases}$$

Let $X \sim \text{Unif}(0, b)$ $Y = e^X$



$$f_X(x) = \begin{cases} \frac{1}{b} & x \in (0, b) \\ 0 & \text{otherwise} \end{cases}$$

$$F_Y(y) = P(Y \leq y) = P(e^X \leq y) = P(X \leq \ln y) = F_X(\ln y) = \int_{-\infty}^{\ln y} \frac{1}{b} dx = \frac{1}{b} \ln y$$

provided $\ln y \leq b$

observe e^x never neg, so $y \geq 0$. Also domain $[0, b] \xrightarrow{e^x} [1, e^b]$

$$\implies F_Y(y) = \begin{cases} 0 & y < 1 \\ \frac{1}{b} \ln y & 1 < y < e^b \\ 1 & e^b < y \end{cases} \implies f_Y(y) = \begin{cases} \frac{1}{b} \frac{1}{y} & 1 \leq y \leq e^b \\ 0 & \text{otherwise} \end{cases}$$

Let $X \sim \mathcal{N}(0,1)$ $\mu=0$ $\sigma=1$ $Y = X^2$

$$F_Y(y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq \sqrt{y}) = \Phi(\sqrt{y}) - \Phi(-\sqrt{y})$$

$$\text{Then } \frac{d}{dy} F_Y(y) = \frac{d}{dy} \int_{-\infty}^{\sqrt{y}} \varphi(t) dt - \frac{d}{dy} \int_{-\infty}^{-\sqrt{y}} \varphi(t) dt \quad \text{where } \varphi(t) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}t^2}$$

Leibniz integration
rule

$$= \varphi(y^{1/2}) \frac{1}{2} y^{-1/2} - \varphi(-y^{1/2}) \left(-\frac{1}{2} y^{-1/2}\right) = \boxed{\frac{1}{\sqrt{y}} \varphi(\sqrt{y})}$$

Fisz p. 60-61

$Z = XY$

This is now 2-dim

$Z = f(x,y)$ should be in next chapter writeup