

P₂₄₉ - P₂₅₀

16. Let X denote the annual rainfall in a certain region and $X \sim N(40, 4^2)$
 the probability that a year has ^{not} a rainfall of over 50 inches is $p(X \leq 50) =$

$$1 - P\left(\frac{X-40}{4} > \frac{50-40}{4}\right) = 1 - \Phi(2.5) = 1 - .9938 = .0062$$

Hence, $(p(X < 50))^{10} = (.9938)^{10}$ is the final answer.

The assumption is the annual rainfalls are independent random variables.

20. Let X denote the number of people in favor. Then X is binomial with $n=100$, $p=.65$.
 Since $np(1-p) = 100(.65)(.35) > 10$. Then X is approximately normal with mean 65
 standard deviation $\sqrt{(65)(.35) \times 100} \approx 4.77$. Also let Z be a standard normal random variable

$$(a) p(X \geq 50) = p(X \geq 49.5) = p\left(\frac{(X-65)}{4.77} \geq \frac{(49.5-65)}{4.77}\right) \approx p(Z \geq -3.25) \approx .9994$$

$$(b) p(59.5 \leq X \leq 70.5) \approx p\left(\frac{(59.5-65)}{4.77} \leq Z \leq \frac{(70.5-65)}{4.77}\right) = 2\Phi(1.15) - 1 \approx .75$$

$$(c) p(X \leq 74.5) \approx p\left(Z \leq \frac{74.5-65}{4.77}\right) \approx .977$$

21. Let X denote the height, in inches, of a 25-year-old man. Then $X \sim N(71, 6.25)$

$$\text{then } p(X \geq 6 \text{ feet } 2 \text{ inches}) = p(X \geq 74) = p\left(\frac{X-71}{2.5} \geq \frac{74-71}{2.5}\right) = 1 - \Phi(1.2) = 1 - .8849 = .1151$$

$$p(X \geq 6 \text{ foot } 5 \text{ inches} \mid X \geq 6 \text{ foot}) = \frac{p(X \geq 77)}{p(X \geq 72)} = \frac{1 - \Phi(2.4)}{1 - \Phi(0.4)} =$$

$$= \frac{1 - .9918}{1 - .6554} = 0.0245$$

26. Let X denote the number of heads in the 1000 tosses.

If the coin is fair, then $X \sim \text{Binomial}(1000, .5)$ and using the normal approximation of X
 $\sim N(500, 250)$ (since $\mu = 1000 \times (.5) = 500$ and $\sigma^2 = 1000 \times (.5) \times (.5) = 250$), the probability of
 making false conclusion is $p(X \geq 525) = 1 - p(X \leq 524) \approx 1 - \Phi\left(\frac{524-500}{\sqrt{250}}\right) = 1 - \Phi(1.5179)$

$$= 1 - .9357 = 0.0643$$

If the coin is biased, then $X \sim \text{Bin}(1000, .55)$ and the normal approximation is $X \sim$
 $N(550, 1000(.55)(.45))$. Thus the probability of making a wrong conclusion is now

$$p(X \leq 524) \approx \Phi\left(\frac{524-550}{\sqrt{1000(.55)(.45)}}\right) = \Phi(-1.65) = 1 - \Phi(1.65) = 1 - .9505 = 0.0495$$

27. Let X denote the times the coin landed heads. if the coin is fair, $X \sim \text{Binomial}(1000, .5)$
 and approximately normal $(500, 250)$

$$\text{then } p(X \geq 5800) = p(X > 5799.5) = p\left(Z > \frac{799.5}{\sqrt{250}}\right) = p(Z > 15.99) \approx 0$$

so it is reasonable to say the coin is unfair.

2. (a) V is symmetric about 0 if and only if the distribution of V is the same as $-V$

$$\text{then } \forall t \quad P(V \leq t) = \int_{-\infty}^t f(v) dv = P(-V \leq t) = \int_{-t}^{\infty} f(v) dv = \int_{-\infty}^t f(-v) dv.$$

take the derivatives $f(t) = f(-t) \quad \forall t$

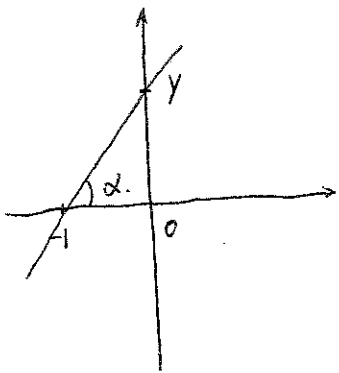
$$(b) \quad EV = \int_{-\infty}^{\infty} v \cdot f(v) dv = \int_0^{\infty} v f(v) dv + \int_{-\infty}^0 v f(v) dv = \int_0^{\infty} v f(v) dv + \int_0^{\infty} -v f(-v) dv = 0$$

$$E(V^{2m+1}) = \int_0^{\infty} v^{2m+1} f(v) dv + \int_{-\infty}^0 v^{2m+1} f(v) dv$$

$$= \int_0^{\infty} v^{2m+1} f(v) dv + \int_{\infty}^0 -v^{2m+1} f(-v) d(v)$$

$$= \int_0^{\infty} v^{2m+1} [f(v) - f(-v)] dv = 0 \quad \text{for any integer } m.$$

3



Let $\tan \alpha$ denote the slope of a line through point $(-1, 0)$ and through point $(0, y)$, then $y = \tan \alpha$ and $\alpha \sim \text{Uniform}(0, \pi)$

then $P(Y \leq y) = P(\tan \alpha \leq y) = P(\alpha \leq \arctan y) = \frac{1}{\pi} \arctan y.$

$$f(y) = \frac{d}{dy} \left(\frac{1}{\pi} \arctan y \right) = \frac{1}{\pi(1+y^2)}$$

so y has the standard Cauchy density.

4. if y follows the Cauchy density $c(\lambda, \mu)$ $f(y) = \frac{1}{\pi} \frac{\lambda}{\lambda^2 + (y-\mu)^2}$ $-\infty < y < \infty$.

then the density function of $\frac{1}{y} = x$ is $f(x) = \frac{1}{\pi} \frac{\lambda}{\lambda^2 + (\frac{1}{x} - \mu)^2} \cdot \frac{1}{x^2}$

$$f(x) = \frac{1}{\pi} \frac{\lambda}{\lambda^2 x^2 + (\mu x - 1)^2} = \frac{1}{\pi} \frac{\lambda}{(\lambda^2 + \mu^2) x^2 - 2\mu x + 1} = \frac{\lambda}{\pi(\lambda^2 + \mu^2)} \frac{1}{x^2 - \frac{2\mu}{\lambda^2 + \mu^2} x + \frac{\mu^2}{(\lambda^2 + \mu^2)^2} - \frac{\mu}{\lambda^2 + \mu^2} + \frac{1}{\lambda^2 + \mu^2}}$$

$$= \frac{\lambda}{\pi(\lambda^2 + \mu^2)} \frac{1}{\left(x - \frac{\mu}{\lambda^2 + \mu^2}\right)^2 + \frac{1}{\lambda^2 + \mu^2} \frac{\mu^2}{(\lambda^2 + \mu^2)^2}}$$

$$= \frac{\lambda}{\pi(\lambda^2 + \mu^2)} \frac{1}{\frac{\lambda^2}{(\lambda^2 + \mu^2)^2} + \left(x - \frac{\mu}{\lambda^2 + \mu^2}\right)^2}$$

so $\frac{1}{y}$ follows the Cauchy density $c\left(\frac{\lambda}{\lambda^2 + \mu^2}, \frac{\mu}{\lambda^2 + \mu^2}\right)$

5 (a) If X is a random variable with density f and CDF F , then. $\forall y \in (0, 1)$, $\exists y = F(x)$. we can find x .

$$P(F(X) \leq y) = P(F(X) \leq F(x)) = P(X \leq x) = F(x) = y.$$

So $F(x) \sim \text{Uniform}(0, 1)$

(b). Let $\tilde{X} = F^{-1}(u)$, u is uniform $(0, 1)$, then $\forall y \in \mathbb{R}$.

$$P(\tilde{X} \leq y) = P(F^{-1}(u) \leq y) = P(u \leq F(y)) = F(y).$$

\therefore density $f(\tilde{X}) = \frac{d}{dy} P(\tilde{X} \leq y) = f(y)$. is the same as the density function of x .

6 (a). First prove $E(X^{2k}) = (2k-1)E(X^{2k-2})$. $X \sim N(0, 1)$.

$$\begin{aligned} E(X^{2k-2}) &= \int_{-\infty}^{\infty} x^{2k-2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \left[\frac{x^{2k-1}}{2k-1} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \right]_{-\infty}^{\infty} - \int_{-\infty}^{\infty} \frac{x^{2k-1}}{2k-1} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \\ &= \int_{-\infty}^{\infty} \frac{x^{2k}}{2k-1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \frac{1}{2k-1} E(X^{2k}) \end{aligned}$$

$$\because E(X^2) = 1 \quad \therefore E(X^{2k}) = (2k-1)E(X^{2k-2}) = (2k-1)(2k-3)E(X^{2k-4})$$

$$\dots = (2k-1)!! E(X^2) = (2k-1)!! \quad \forall k \in \mathbb{N}$$

$$\begin{aligned} E(X^{2k+1}) &= \int_{-\infty}^{\infty} x^{2k+1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \int_0^{\infty} x^{2k+1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx + \int_{-\infty}^0 x^{2k+1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \\ &= \int_0^{\infty} x^{2k+1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx + \int_0^{\infty} -x^{2k+1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \\ &= \int_0^{\infty} x^{2k+1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx - \int_0^{\infty} x^{2k+1} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = 0 \end{aligned}$$

(b) Let X denote the random variable produced after the coin toss.

$$X = \begin{cases} T & \text{coin lands Head} \\ -T & \text{coin lands tail} \end{cases}$$

$$\text{if } x > 0 \quad P(X \leq x) = P(X \leq x | H)P(H) + P(X \leq x | T)P(T)$$

$$= \frac{1}{2} \cdot \int_0^x \lambda e^{-\lambda x} dx + \frac{1}{2} = \frac{1}{2} + \frac{1}{2}(1 - e^{-\lambda x}) = 1 - \frac{1}{2}e^{-\lambda x}$$

$$\text{if } x < 0 \quad P(X \leq x) = P(X \leq x | H)P(H) + P(X \leq x | T)P(T)$$

$$= P(X \leq x | T)P(T) = \frac{1}{2} \cdot \int_{-\infty}^x \lambda e^{-\lambda x} dx = \frac{1}{2} e^{\lambda x}$$

$$\text{the density function } f(x) = \begin{cases} \frac{1}{2} \lambda e^{-\lambda x} & x > 0 \\ \frac{1}{2} \lambda e^{\lambda x} & x < 0 \end{cases}$$

