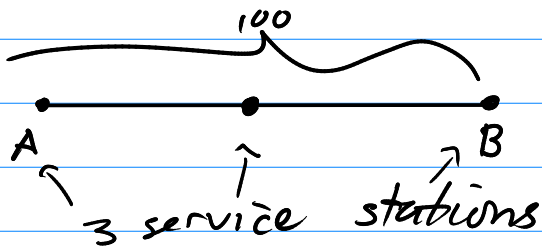


# Solutions 10

5.12



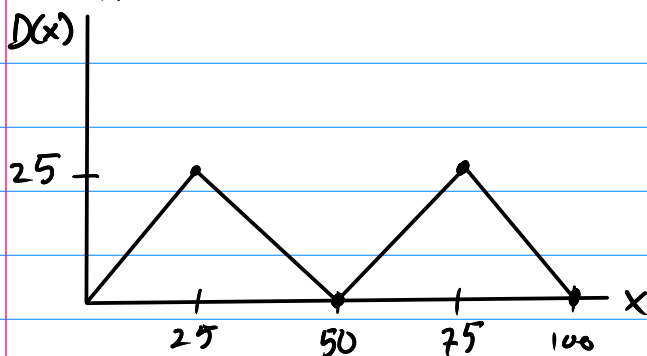
$X$  = location of breakdown  
is uniform on  $(0, 100)$

$$\text{PDF } f(x) = \begin{cases} \frac{1}{100} & 0 < x < 100 \\ 0 & \text{otherwise} \end{cases}$$

Let  $D(X)$  be the distance from location of breakdown

$$D(X) = \begin{cases} X & 0 < X \leq 25 \\ 50 - X & 25 < X \leq 50 \\ X - 50 & 50 < X \leq 75 \\ 100 - X & 75 < X < 100 \end{cases}$$

Which looks like



$$E[D(X)] = \int_0^{100} D(x) f(x) dx = \int_0^{100} D(x) \frac{1}{100} dx = \frac{1}{100} \int_0^{100} D(x) dx$$

$$= \frac{1}{100} (\text{Area under the graph of } D(x))$$

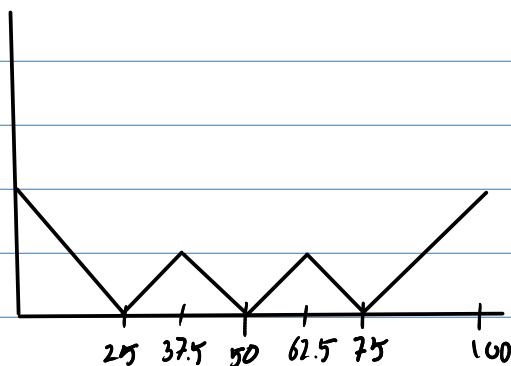
$$= \frac{1}{100} (4) \left( \frac{1}{2} \cdot 25 \cdot 25 \right) \quad (\text{break up into 4 triangles})$$

$$= \frac{1}{2} \cdot 25 = 12.5.$$

If stations are located at 25, 50, and 75,

$$D(x) = \begin{cases} 25-x & 0 < x \leq 25 \\ x-25 & 25 < x \leq 37.5 \\ 50-x & 37.5 < x \leq 50 \\ x-50 & 50 < x \leq 62.5 \\ 75-x & 62.5 < x \leq 75 \\ x-75 & 75 < x < 100 \end{cases}$$

which looks like



$$E[D(x)] = \int_0^{100} D(x) \frac{1}{100} dx$$

$$= \frac{1}{100} (\text{Area under curve})$$

$$= \frac{1}{100} \left[ 4 \cdot \left( \frac{1}{2} (12.5)^2 \right) + 2 \left( \frac{1}{2} 25^2 \right) \right]$$

$$= \frac{1}{100} [312.5 + 625] = \frac{1}{100} [937.5] = 9.375$$

Since  $9.375 < 12.5$ , the second scheme is more efficient.

5.13 You arrive a bus stop at 10.

Arrival of bus is uniformly distributed between 10 and 10:30.

(a) What is probability we have to wait longer than 10 minutes

Wait longer than 10 minutes  $\Leftrightarrow$  bus arrives between 10:10 and 10:30

$$P(10:10 < X < 10:30) = \frac{20 \text{ minutes}}{30 \text{ minutes}} = \frac{2}{3}$$

(b) If, at 10:15 bus has not arrived, find probability you must wait another 10 minutes

Wait another 10 minutes  $\Leftrightarrow$  bus arrives between 10:25 and 10:30

$$P(10:25 < X < 10:30 \mid 10:15 < X) = \frac{P(10:25 < X < 10:30)}{P(10:15 < X)}$$

$$\frac{5/30}{15/30} = \frac{5}{15} = \frac{1}{3}$$

5.15  $X$  normal with  $\mu=10$  and  $\sigma^2=36$   
compute  $\sigma=6$

let  $Z$  denote standard normal R.V.

NOTE: I USED THE TABLE ON P. 201 to look up  $\Phi(x)$

CALCULATOR MAY GIVE SLIGHTLY DIFFERENT RESULTS

$$\begin{aligned} \text{(a)} \quad P\{X > 5\} &= P\left\{\frac{X-10}{6} > \frac{5-10}{6}\right\} = P\left\{Z > -\frac{5}{6}\right\} = P\left\{Z < \frac{5}{6}\right\} \\ &= \Phi\left(\frac{5}{6}\right) \approx \Phi(.83) \approx .7967 \end{aligned}$$

$$\begin{aligned} \text{(b)} \quad P\{4 < X < 16\} &= P\left\{\frac{4-10}{6} < Z < \frac{16-10}{6}\right\} = P\{-1 < Z < 1\} \\ &= \Phi(1) - \Phi(-1) = \Phi(1) - [1 - \Phi(1)] = 2\Phi(1) - 1 \\ &= 2(.8413) - 1 = .6826 \end{aligned}$$

$$\begin{aligned} \text{(c)} \quad P\{X < 8\} &= P\left\{Z < \frac{8-10}{6}\right\} = P\left\{Z < -\frac{1}{3}\right\} = 1 - \Phi\left(\frac{1}{3}\right) \\ &= 1 - \Phi(.33) = 1 - (.6293) = .3707 \end{aligned}$$

$$(d) P\{X < 20\} = P\left\{Z < \frac{20-10}{6}\right\} = P\left\{Z < \frac{5}{3}\right\}$$

$$= \Phi\left(\frac{5}{3}\right) \approx \Phi(1.66) = .9515$$

$$(e) P\{X > 16\} = P\left\{Z > \frac{16-10}{6}\right\} = P\{Z > 1\} = 1 - \Phi(1)$$

$$= 1 - .8413 = .1587$$

5.16  $X =$  annual rainfall is normal with  $\mu = 40$ ,  $\sigma = 4$   
( $X$  is in inches)

$$P\{X > 50\} = P\left\{\frac{X-40}{4} > \frac{50-40}{4}\right\} = P\left\{\frac{X-40}{4} > 2.5\right\}$$

$$= 1 - \Phi(2.5) = 1 - (.9938) = .0062$$

The number of years it takes to have a year with over 50 inches of rain is a geometric random variable with  $p = .0062$  (Assuming the rainfalls in different years are independent)

$$\text{So } P\{\text{take over 10 years to get } > 50 \text{ inches}\}$$

$$= (1-p)^{10} = (.9938)^{10} = .9397$$

5.18  $X$  is normal with  $\mu = 5$  and unknown  $\sigma$ .  
If we know  $P\{X > 9\} = .2$ , then

$$.2 = P\left\{\frac{X-5}{\sigma} > \frac{9-5}{\sigma}\right\} = P\left\{\frac{X-5}{\sigma} > \frac{4}{\sigma}\right\} = 1 - \Phi\left(\frac{4}{\sigma}\right)$$

$$\therefore \Phi\left(\frac{4}{\sigma}\right) = 1 - .2 = .8$$

Doing a reverse lookup in the table for  $\Phi$ , we see

$$\frac{4}{\sigma} \approx .85 \Rightarrow \sigma \approx \frac{4}{.85} = 4.7$$

$$\text{Thus } \text{Var}(X) = \sigma^2 \approx 22$$

5.23 1000 independent rolls of a fair die  
 $X = \#$  of 6's rolled

Compute approximation to  $P\{150 \leq X \leq 200\}$

$X$  is binomial with  $n=1000$   $p=\frac{1}{6}$

$$\mu = np = 166.6$$

$$\sigma^2 = np(1-p) = 138.8 \quad \sigma = 11.79$$

Apply DeMoivre-Laplace Limit theorem

$$P\{150 \leq X \leq 200\} = P\{149.5 < X < 200.5\} \quad \text{continuity corrector}$$

$$\approx P\left\{ \frac{149.5 - 166.6}{11.79} < Z < \frac{200.5 - 166.6}{11.79} \right\}$$

$$\approx P\{-1.450 < Z < 2.875\}$$

$$= \Phi(2.875) - \Phi(-1.450)$$

$$\approx .9244$$

Note: I used computer for some of these steps. Rounding and using table may give slightly different answer.

Second part: Assume 6 appears exactly 200 times

Find conditional probability that 5 appears less than 150 times.

FACT: Assuming 6 appears exactly 200 times, the number of 5's which appear is a binomial Random Var. with  $n=800$  and  $p=\frac{1}{5}$

Quick intuitive argument: If 6 appears 200 times, we can set those 200 trials aside, and we have  $n=800$  trials on which a 5 might appear. Since we know a 6 cannot appear on any of these 800 trials, the probability of getting a 5 is

$$P(5 | \text{not } 6) = \frac{P(5)}{P(\text{not } 6)} = \frac{1/6}{5/6} = \frac{1}{5}, \text{ so } p = \frac{1}{5}$$

Proof with formulas let  $Y = \#$  of 5's  $X = \#$  6's

$$P(Y=k | X=200) = \frac{P(Y=k \& X=200)}{P(X=200)}$$

$$P(X=200) = \binom{1000}{200} \left(\frac{1}{6}\right)^{200} \left(\frac{5}{6}\right)^{800}$$

$$P(Y=k \& X=200) = \binom{1000}{200} \binom{800}{k} \left(\frac{1}{6}\right)^{200} \left(\frac{1}{6}\right)^k \left(\frac{4}{6}\right)^{800-k}$$

where 6's go     where 5's go      $\uparrow$  200 6's      $\uparrow$  k 5's      $\uparrow$  800-k 1,2,3,4,5

$$P(Y=k | X=200) = \frac{\binom{1000}{200} \binom{800}{k} \left(\frac{1}{6}\right)^{200} \left(\frac{1}{6}\right)^k \left(\frac{4}{6}\right)^{800-k}}{\binom{1000}{200} \left(\frac{1}{6}\right)^{200} \left(\frac{5}{6}\right)^{800}}$$

$$= \binom{800}{k} \left(\frac{1}{6}\right)^k \left(\frac{4}{6}\right)^{800-k} \left(\frac{6}{5}\right)^{800} = \binom{800}{k} 1^k 4^{800-k} \left(\frac{1}{6}\right)^{800} \frac{6^{800}}{5^{800}}$$

$$= \binom{800}{k} \left(\frac{1}{5}\right)^k \left(\frac{4}{5}\right)^{800-k}$$

which is binomial with  $n=800$   $p=\frac{1}{5}$

$P\{Y < 150 | X=200\}$  can be approximated

$$np = 160 \quad np(1-p) = 128 \quad \sqrt{np(1-p)} = 8\sqrt{2} \approx 11.31$$

$$P\{Y < 150 | X=200\} = P\{Y < 149.5 | X=200\}$$

$$= P\left\{Z < \frac{149.5 - 160}{11.31}\right\}$$

$$= P\{Z < -0.9284\} = \Phi(-0.9284) \approx 0.1766$$

5.27 In 10000 tosses coin lands on heads 5800 times

Is it reasonable to conclude coin is unfair

Suppose coin is fair: then  $X = \#$  heads is binomial/  
with  $n=10000$ ,  $p=\frac{1}{2}$ ,  $\mu=np=5000$ ,  $\sigma^2=np(1-p)=2500$   
 $\sigma=\sqrt{2500}=50$

since  $\mu = 5000$ ,  $\sigma = 50$ ,  $5800 = \mu + 16\sigma$

So 5800 heads represents a deviation from mean of 16 standard deviations

So  $P\{X \geq 5800\} = P\{Z \geq 16\} = 1 - \Phi(16)$ ,  
and this number is essentially zero.

In fact even:

$P\{X \geq 5100\} = P\{Z \geq 2\} \approx .02275 = 2.275\%$  is quite small

$P\{X \geq 5400\} = P\{Z \geq 8\} \approx 6.66 \times 10^{-16}$  is very very very small.

$P\{X \geq 5800\} = P\{Z \geq 16\}$  is so small it is outside the limits of precision on my computer.

So it is nearly certain the coin is not fair.

P.227 - 228

5.2 Statement:  $E[Y] = \int_0^{\infty} P\{Y > y\} dy - \int_0^{\infty} P\{Y < -y\} dy$

Step 1:  $\int_0^{\infty} P\{Y > y\} dy = \int_0^{\infty} x f_Y(x) dx$

(this is the same proof as in class)

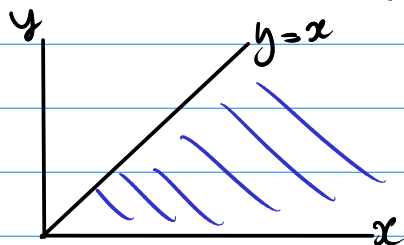
$$P\{Y > y\} = \int_y^{\infty} f_Y(x) dx$$



$$\text{So } \int_0^{\infty} P\{Y > y\} dy = \int_0^{\infty} \int_y^{\infty} f_Y(x) dx dy$$

The integral is over the region  $\{(x, y) : 0 < y < \infty, y < x < \infty\}$

Picture:



$$= \{(x, y) : 0 < x < \infty, 0 < y < x\}$$

Change order of integration:  $= \int_0^{\infty} \int_0^x f_Y(x) dy dx$

$$= \int_0^{\infty} [f_Y(x) y]_{y=0}^{y=x} dx = \int_0^{\infty} x f_Y(x) dx \quad \text{Done with Step 1.}$$

Step 2  $\int_0^{\infty} P\{Y < -y\} dy = \int_{-\infty}^0 x f_Y(x) dx$

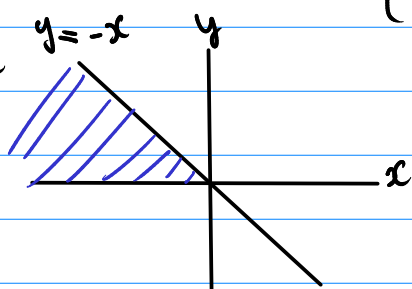
$$P\{Y < -y\} = \int_{-\infty}^{-y} f_Y(x) dx$$

$$\text{So } \int_0^{\infty} P\{Y < -y\} dy = \int_0^{\infty} \int_{-\infty}^{-y} f_Y(x) dx dy$$

Region of integration  $\{(x, y) : 0 < y < \infty, -\infty < x < -y\}$

$$= \{(x, y) : -\infty < x < 0, 0 < y < -x\}$$

Picture



Switch order of integration

$$= \int_{-\infty}^0 \int_0^{-x} f_Y(x) dy dx = \int_{-\infty}^0 \left[ y f_Y(x) \right]_{y=0}^{y=-x} dx$$

$$= \int_{-\infty}^0 (-x) f_Y(x) dx = - \int_{-\infty}^0 f_Y(x) dx \quad \text{Done with Step 2}$$

Finally,  
Step 3:  $E[Y] = \int_{-\infty}^{\infty} x f_Y(x) dx$  by definition

$$= \int_{-\infty}^0 x f_Y(x) dx + \int_0^{\infty} x f_Y(x) dx$$

$$= - \int_0^{\infty} P\{Y < -y\} dy + \int_0^{\infty} P\{Y > y\} dy \quad \text{QED}$$

5.3 Statement: Let  $Y = g(X)$  then

$$E[Y] = E[g(X)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

Using 5.2, we have

$$E[Y] = \int_0^{\infty} P\{Y > y\} dy - \int_0^{\infty} P\{Y < -y\} dy$$

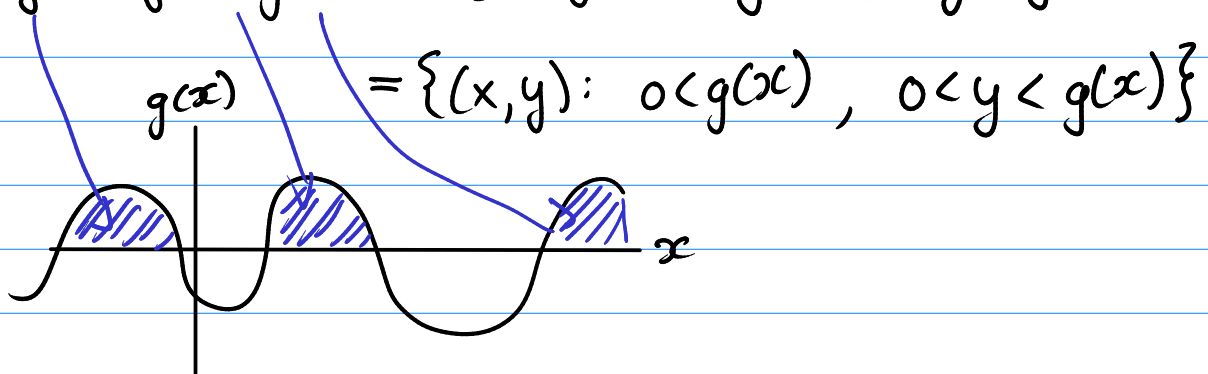
$$= \int_0^{\infty} P\{g(X) > y\} dy - \int_0^{\infty} P\{g(X) < -y\} dy$$

Analyze  $\int_0^{\infty} P\{g(x) > y\} dy$

$$\left( P\{g(x) > y\} = \int_{\substack{x \text{ such} \\ \text{that } g(x) > y}} f_X(x) dx \right)$$

$$= \int_0^{\infty} \int_{\substack{x \text{ such that} \\ g(x) > y}} f_X(x) dx dy$$

Region of integration:  $\{(x, y) : 0 < y < \infty, y < g(x)\}$



$$= \int_{\substack{x \text{ such} \\ \text{that } g(x) > 0}} \int_0^{g(x)} f_X(x) dy dx$$

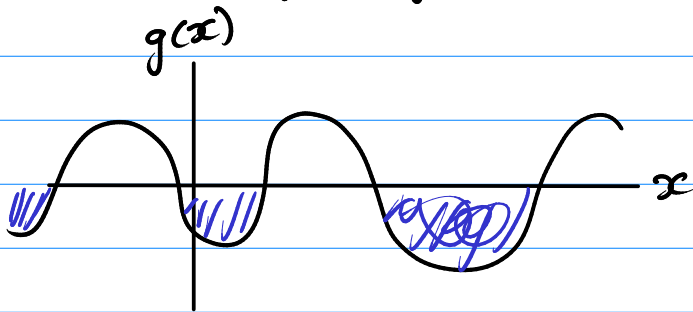
$$= \int_{\substack{x \text{ such that} \\ g(x) > 0}} g(x) f_X(x) dx$$

$$\text{Analyze } - \int_0^{\infty} P\{g(x) < -y\} dy$$

$$= - \int_0^{\infty} \int_{x \text{ such that } g(x) < -y} f_X(x) dx dy$$

$$\text{Region: } \{(x, y) : 0 < y < \infty, g(x) < -y\}$$

$$\{(x, y) : g(x) < 0, 0 < y < -g(x)\}$$



$$= - \int_{x \text{ such that } g(x) < 0} \int_0^{-g(x)} f_X(x) dx = - \int_{x \text{ such that } g(x) < 0} -g(x) f_X(x) dx$$

$$= \int_{x \text{ such that } g(x) < 0} g(x) f_X(x) dx$$

$$E[g(x)] = \int_{x \text{ such that } g(x) > 0} g(x) f_X(x) dx + \int_{x \text{ such that } g(x) < 0} g(x) f_X(x) dx$$

$$\text{Now } \int_{x \text{ such that } g(x) = 0} g(x) f_X(x) dx = 0$$

So we can neglect the set where  $g(x)=0$ , and we get

$$E[g(x)] = \int_{\text{all } x} g(x) f_X(x) dx = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

( $g(x) > 0, g(x) < 0, \text{ or } g(x) = 0$ ) QED.

5.9  $Z$  standard normal random variable, and  $x > 0$

(a)  $P\{Z > x\} = P\{Z < -x\}$

In pictures

$$P\{Z > x\} = \text{[Normal curve with area to the right of } x \text{ shaded]} = \text{[Normal curve with area to the left of } -x \text{ shaded]} = P\{Z < -x\}$$

With integrals

$$\begin{aligned} P\{Z > x\} &= \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-y^2/2} dy \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{-x} e^{-u^2/2} (-du) \quad \left. \begin{array}{l} u = -y \\ du = -dy \end{array} \right\} \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{-x} e^{-u^2/2} du = P\{Z < -x\} \end{aligned}$$

$$(b) P\{|Z| > x\} = 2P\{Z > x\}$$

Proof  $|Z| > x \iff Z > x \text{ or } -Z > x$   
 $\iff Z > x \text{ or } Z < -x$

$$\begin{aligned} P\{|Z| > x\} &= P\{Z > x\} + P\{Z < -x\} \\ &= P\{Z > x\} + P\{Z > x\} \quad \text{by part (a)} \\ &= 2P\{Z > x\} \quad \text{QED} \end{aligned}$$

$$(c) P\{|Z| < x\} = 2P\{Z < x\} - 1$$

Proof:  $P\{|Z| < x\} = 1 - P\{|Z| > x\}$   
 $= 1 - 2P\{Z > x\}$  by part (b)  
 $= 1 - 2[1 - P\{Z < x\}]$   
 $= 1 - 2 + 2P\{Z < x\}$   
 $= 2P\{Z < x\} - 1 \quad \text{QED}$