

STAT2001 & STAT6039 Final Exam June 2016 Solutions

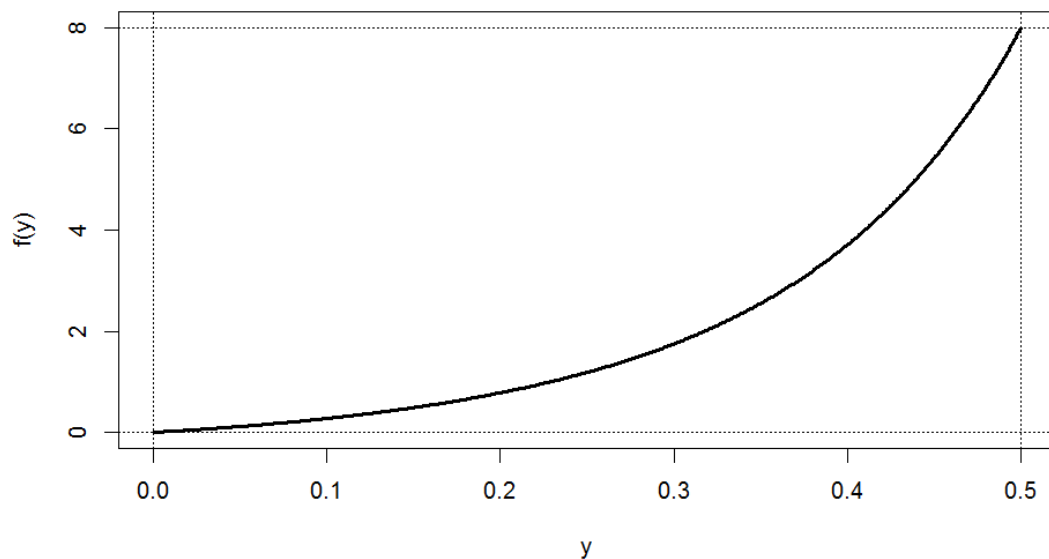
Solution to Problem 1

(a) X has pdf $f(x) = 2x, 0 < x < 1$. Also, $y = x/(x+1)$ is a strictly increasing function with inverse $x = y(1-y)^{-1}$. So, by the transformation rule,

$$\begin{aligned} f(y) &= f(x) \left| \frac{dx}{dy} \right| = 2 \times y(1-y)^{-1} \times \left| y(-1)(1-y)^{-2}(-1) + 1 \times (1-y)^{-1} \right| \\ &= \frac{2y}{1-y} \left(\frac{y}{(1-y)^2} + \frac{1}{1-y} \right). \end{aligned}$$

Thus, $\boxed{f(y) = \frac{2y}{(1-y)^3}, 0 < y < \frac{1}{2}}$, as illustrated in the sketch below.

$$\begin{aligned} \text{Also, } EY &= E\left(\frac{X}{X+1}\right) = \int_0^1 \frac{x}{x+1} \times 2x dx = 2 \int_1^2 \frac{(t-1)^2}{t} dt \quad (\text{where } t = x+1) \\ &= 2 \int_1^2 \left(t - 2 + \frac{1}{t} \right) dt = 2 \left[\frac{t^2}{2} - 2t + \log t \right]_{t=1}^2 = 2 \left(\frac{2^2}{2} - 2 \times 2 + \log 2 - \frac{1^2}{2} + 2 \times 1 - \log 1 \right) \\ &= 2 \left(2 - 4 + \log 2 - \frac{1}{2} + 2 \right) = 2 \log 2 - 1 = \boxed{0.3863}. \end{aligned}$$



(b) R has cdf $F(r) = P(R \leq r) = P(Z / U \leq r) = P(Z \leq rU)$

$$= \int_{u=0}^1 \left(\int_{z=-\infty}^{ru} \phi(z) dz \right) 1 du = \int_{u=0}^1 \Phi(ru) du. \quad (1)$$

So R has pdf $f(r) = F'(r) = \int_{u=0}^1 \phi(ru) u du = \int_0^1 \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(ru)^2} u du.$

We now substitute $t = \frac{1}{2} r^2 u^2$, so that $\frac{dt}{du} = r^2 u$ and $u du = \frac{dt}{r^2}$,

and thereby obtain $f(r) = \frac{1}{r^2 \sqrt{2\pi}} \int_0^{\frac{1}{2} r^2} e^{-t} dt.$

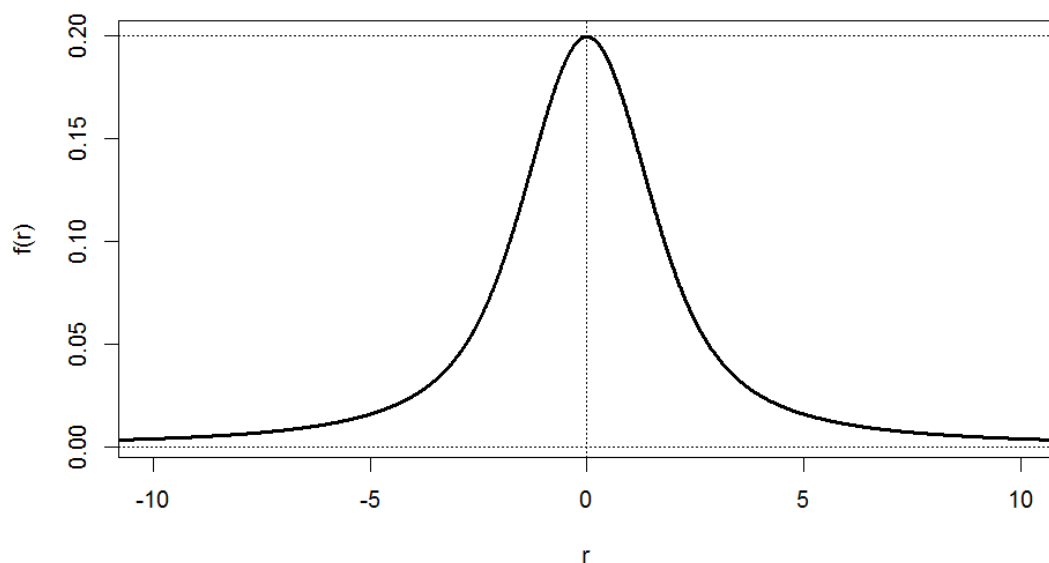
Thus $\boxed{f(r) = \frac{1}{r^2 \sqrt{2\pi}} \left(1 - e^{-\frac{1}{2} r^2} \right), \quad r \in \mathfrak{R} \quad (r \neq 0)}$, as shown in the sketch below.

We see that $f(r)$ is symmetric around zero, with a maximum of approximately

$$\frac{1}{0.0001^2 \sqrt{2\pi}} \left(1 - e^{-\frac{1}{2} 0.0001^2} \right) = 0.1994711 \approx 0.2.$$

Note: More accurately, we can apply L'Hospital's rule to get

$$\lim_{r \rightarrow 0} \left\{ \frac{1}{r^2 \sqrt{2\pi}} \left(1 - e^{-\frac{1}{2} r^2} \right) \right\} = \lim_{r \rightarrow 0} \left\{ \frac{1}{2 \cancel{r} \sqrt{2\pi}} \left(0 - e^{-\frac{1}{2} r^2} (-\cancel{r}) \right) \right\} = \frac{1}{\sqrt{8\pi}} = 0.1994711.$$



To find $P(R > 8)$, we consider any $r > 0$ (e.g. 8) and the region under the line $z = ru$ in the u - z plane within the infinite rectangle $(0,1) \times (-\infty, \infty)$. We then write (1) as

$$\begin{aligned} F(r) &= \int_{u=0}^1 \left(\int_{z=-\infty}^{ru} \phi(z) dz \right) du = \frac{1}{2} + \int_{z=0}^r \left(\int_{u=z/r}^1 1 du \right) \phi(z) dz \\ &= \frac{1}{2} + \int_{z=0}^r \left(1 - \frac{z}{r} \right) \phi(z) dz = \frac{1}{2} + \int_0^r \phi(z) dz - \frac{1}{r} \int_0^r z \phi(z) dz \\ &= \frac{1}{2} + \left(\Phi(r) - \frac{1}{2} \right) - \frac{1}{r} I, \quad \text{where } I = \int_0^r z \phi(z) dz = \int_0^r z \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz. \end{aligned}$$

We now substitute $t = \frac{1}{2}z^2$, so that $\frac{dt}{dz} = z$ and $z dz = dt$.

Thereby we obtain $I = \frac{1}{\sqrt{2\pi}} \int_0^{\frac{1}{2}r^2} e^{-t} dt = \frac{1}{\sqrt{2\pi}} \left(1 - e^{-\frac{1}{2}r^2} \right)$,

and hence $F(r) = \Phi(r) - \frac{1}{r} \left\{ \frac{1}{\sqrt{2\pi}} \left(1 - e^{-\frac{1}{2}r^2} \right) \right\} \quad (r > 0).$ (2)

It follows that

$$\begin{aligned} P(R > 8) &= 1 - F_R(8) = 1 - \Phi(r) + \frac{1}{r} \left\{ \frac{1}{\sqrt{2\pi}} \left(1 - e^{-\frac{1}{2}r^2} \right) \right\} \quad \text{where } r = 8 \\ &\approx \frac{1}{8} \left\{ \frac{1}{\sqrt{2\pi}} \left(1 - e^{-\frac{1}{2}8^2} \right) \right\} \quad (\text{since } \Phi(8) = P(Z \leq 8) \approx 1) \\ &= \boxed{0.04987}. \end{aligned}$$

Note: Equation (2) provides an alternative way to get the pdf of R , namely as

$$\begin{aligned} f(r) = F'(r) &= \phi(r) - \cancel{\frac{1}{r}} \left\{ \frac{1}{\sqrt{2\pi}} \left(0 - e^{-\frac{1}{2}r^2} (-\cancel{r}) \right) \right\} - \left(\frac{-1}{r^2} \right) \left\{ \frac{1}{\sqrt{2\pi}} \left(1 - e^{-\frac{1}{2}r^2} \right) \right\} \\ &= \cancel{\phi(r)} - \cancel{\phi(r)} + \frac{1}{r^2 \sqrt{2\pi}} \left(1 - e^{-\frac{1}{2}r^2} \right), \quad r > 0 \text{ (as before)}. \end{aligned}$$

By symmetry, this last formula must also give the pdf of R when $r < 0$.

R Code for Problem 1 (not required)

(a)

```
yvec=seq(0,0.5,0.001); fyvec=2*yvec/(1-yvec)^3
X11(w=8,h=5); plot(yvec,fyvec,type="l",lwd=3,xlab="y",ylab="f(y)")
abline(h=c(0,8), lty=3); abline(v=c(0,0.5),lty=3)
2*log(2)-1 # 0.3862944
# Check mean via Monte Carlo
set.seed(331); xv=rbeta(100000,2,1); yv=xv/(xv+1); mean(yv) # 0.3860128 OK
```

(b)

```
rvec=seq(-12.001,11.999,0.01)
frvec= (1/ (rvec^2* sqrt(2*pi) ) ) * (1-exp(-0.5*rvec^2))
rval=0.0001; (1/ (rval^2* sqrt(2*pi) ) ) * (1-exp(-0.5*rval^2)) # 0.1994711
1/sqrt(8*pi) # 0.1994711 (exact mode)
X11(w=8,h=5); plot(rvec,frvec,type="l",lwd=3,xlab="r",ylab="f(r)", xlim=c(-10,10))
abline(h=c(0,0.2), lty=3); abline(v=0,lty=3)
rval=8; (1/rval)*(1/(sqrt(2*pi)))*(1-exp(-0.5*rval^2)) # 0.04986779
# Check probability via Monte Carlo
set.seed(768); uv=runif(100000); zv=rnorm(100000); rv=zv/uv
length(rv[rv>8])/100000 # 0.0491 OK
```

Solution to Problem 2

(a) Let N = "Number of rolls until the first swipe". Then, by a first-step analysis:

$$EN = P(1)E(N | 1) + P(2)E(N | 2) + P(3)E(N | 3) + P(0)E(N | 0)$$

where $0 = "4, 5 \text{ or } 6"$

$$\begin{aligned} &= \frac{1}{6}E(N | 1) + \frac{1}{6}E(N | 2) + \frac{1}{6}E(N | 3) + \frac{3}{6}\{EN + 1\} \\ &= \frac{3}{6}E(N | 1) + \frac{3}{6}\{EN + 1\} \quad \text{since } E(N | 1) = E(N | 2) = E(N | 3). \end{aligned}$$

Also,

$$\begin{aligned} E(N | 1) &= P(11 | 1)E(N | 1, 11) + P(12 | 1)E(N | 1, 12) \\ &\quad + P(13 | 1)E(N | 1, 13) + P(10 | 1)E(N | 1, 10) \\ &= \frac{1}{6}E(N | 11) + \frac{1}{6}E(N | 12) + \frac{1}{6}E(N | 13) + \frac{3}{6}\{EN + 2\} \\ &= \frac{1}{6}\{E(N | 1) + 1\} + \frac{2}{6}E(N | 12) + \frac{3}{6}\{EN + 2\} \quad \text{since } E(N | 13) = E(N | 12). \end{aligned}$$

Furthermore,

$$\begin{aligned} E(N | 12) &= P(121 | 12)E(N | 12, 121) + P(122 | 12)E(N | 12, 122) \\ &\quad + P(123 | 12)E(N | 12, 123) + P(120 | 12)E(N | 12, 120) \\ &= \frac{1}{6}\{E(N | 12) + 1\} + \frac{1}{6}\{E(N | 1) + 2\} + \frac{1}{6} \times 3 + \frac{3}{6}\{EN + 3\}. \end{aligned}$$

Writing $a = EN$, and $b = E(N | 1)$, $c = E(N | 12)$, these equations may be written as:

$$6a = 3b + 3a + 3 \quad \Rightarrow \quad b = a - 1 \quad (1)$$

$$6b = b + 1 + 2c + 3a + 6 \quad \Rightarrow \quad 5b = 2c + 3a + 7 \quad (2)$$

$$6c = c + 1 + b + 2 + 3 + 3a + 9 \quad \Rightarrow \quad 5c = b + 3a + 15. \quad (3)$$

$$\text{Then: } (1) \rightarrow (2) \Rightarrow 5(a - 1) = 2c + 3a + 7 \Rightarrow c = a - 6 \quad (4)$$

$$(1) \& (4) \rightarrow (3) \Rightarrow 5(a - 6) = a - 1 + 3a + 15 \Rightarrow a = \boxed{44}.$$

(b) Let A = "Number of rolls to first swipe is even". Then, by a first-step analysis:

$$\begin{aligned}
 P(A) &= P(1)P(A|1) + P(2)P(A|2) + P(3)P(A|3) + P(0)P(A|0) \\
 &= \frac{1}{6}P(A|1) + \frac{1}{6}P(A|2) + \frac{1}{6}P(A|3) + \frac{3}{6}\{1 - P(A)\} \\
 &= \frac{3}{6}P(A|1) + \frac{3}{6}\{1 - P(A)\} \quad \text{since } P(A|1) = P(A|2) = P(A|3).
 \end{aligned}$$

Also,

$$\begin{aligned}
 P(A|1) &= P(11|1)P(A|1,11) + P(12|1)P(A|1,12) \\
 &\quad + P(13|1)P(A|1,13) + P(10|1)P(A|1,10) \\
 &= \frac{1}{6}\{1 - P(A|1)\} + \frac{1}{6}P(A|12) + \frac{1}{6}P(A|13) + \frac{3}{6}P(A) \\
 &= \frac{1}{6}\{1 - P(A|1)\} + \frac{2}{6}P(A|12) + \frac{3}{6}P(A) \quad \text{since } P(A|13) = P(A|12).
 \end{aligned}$$

Furthermore,

$$\begin{aligned}
 P(A|12) &= P(121|12)P(A|12,121) + P(122|12)P(A|12,122) \\
 &\quad + P(123|12)P(A|12,123) + P(120|12)P(A|12,120) \\
 &= \frac{1}{6}\{1 - P(A|21)\} + \frac{1}{6}P(A|2) + \frac{1}{6} \times 0 + \frac{3}{6}\{1 - P(A)\} \\
 &= \frac{1}{6}\{1 - P(A|12)\} + \frac{1}{6}P(A|1) + \frac{3}{6}\{1 - P(A)\} \quad \text{since } P(A|21) = P(A|12).
 \end{aligned}$$

Writing $a = P(A)$, and $b = P(A|1)$, $c = P(A|12)$, these equations may be written as:

$$6a = 3b + 3 - 3a \quad \Rightarrow \quad b = 3a - 1 \quad (1)$$

$$6b = 1 - b + 2c + 3a \quad \Rightarrow \quad 7b = 1 + 2c + 3a \quad (2)$$

$$6c = 1 - b + 2c + 3a \quad \Rightarrow \quad 4c = 1 - b + 3a. \quad (3)$$

$$\text{Then: } (1) \rightarrow (2) \Rightarrow 21a - 7 = 1 + 2c + 3a \Rightarrow c = 9a - 4 \quad (4)$$

$$(1) \& (4) \rightarrow (3) \Rightarrow 36 - 16 = 1 - 3a + 1 + 3a \Rightarrow a = \boxed{1/2}.$$

(c) Let A_i = "Swipe on rolls $i, i + 1$ and $i + 2$ " and $Y_i = I(A_i)$.

Then, on n rolls, the number of swipes is $Y = Y_1 + \dots + Y_{n-2}$, with expectation

$$EY = EY_1 + \dots + EY_{n-2} = (n-2)EY_1 = (n-2)P(A_1) = (n-2)\frac{3}{6} \times \frac{2}{6} \times \frac{1}{6} = \frac{n-2}{36}.$$

Setting EY to 1 yields $n = \boxed{38}$.

(d) The probability of at least one swipe on six rolls is

$$p = P(A_1 \cup \dots \cup A_4) = \sum_{i=1, \dots, 4} P(A_i) - \sum_{i < j} P(A_i A_j) + \sum_{i < j < k} P(A_i A_j A_k) - P(A_1 \dots A_4).$$

Now: $P(A_1) = \frac{3!}{6^3} = P(A_2) = P(A_3) = P(A_4) = \frac{216}{d}$ where $d = 6^5 = 7776$

$$P(A_1 A_2) = \frac{3!}{6^3} \times \frac{1}{6} = P(A_{23}) = P(A_{34}) = \frac{36}{d}$$

$$P(A_1 A_3) = \frac{3!}{6^3} \times \frac{2!}{6^2} = P(A_{24}) = \frac{12}{d}$$

$$P(A_1 A_4) = \frac{3!}{6^3} \times \frac{3!}{6^3} = \frac{6}{d}$$

$$P(A_1 A_2 A_3) = \frac{3!}{6^3} \times \frac{1}{6^2} = P(A_2 A_3 A_4) = \frac{6}{d}$$

$$P(A_1 A_2 A_4) = \frac{3!}{6^3} \times \frac{1}{6} \times \frac{2!}{6^2} = P(A_1 A_3 A_4) = \frac{2}{d}$$

$$P(A_1 A_2 A_3 A_4) = \frac{3!}{6^3} \times \frac{1}{6^3} = \frac{1}{d}.$$

So $p = 4 \times \frac{216}{d} - \left(3 \times \frac{36}{d} + 2 \times \frac{12}{d} + \frac{6}{d} \right) + \left(2 \times \frac{6}{d} + 2 \times \frac{2}{d} \right) - \frac{1}{d} = \frac{741}{7776} = \frac{247}{2592}.$

So the probability of no sweeps on six rolls is $1 - p = 1 - \frac{247}{2592} = \frac{2345}{2592} = \boxed{0.9047}.$

R Code for Problem 2

(a)

```
trialfun=function(){
  n=2; v=sample(1:6,2,replace=T); cond=F; while(cond==F){
    n=n+1; v=c(v,sample(1:6,1)); ss=v[(n-2):n]
    if( length( grep(1,ss)+grep(2,ss)+grep(3,ss) ) ==1) cond=T }
  n }
set.seed(193); trialfun() # 92
date(); set.seed(284); J=10000; nv=rep(NA,J); for(j in 1:J) nv[j]=trialfun()
date() # Took 32 secs
```

```
me=mean(nv); se=sd(nv); ci=me+c(-1,1)*qnorm(0.975)*se/sqrt(J)
c(me,se,ci) # 43.88960 40.86393 43.08868 44.69052 OK
```

```
# (b) (Follows on from (a))
```

```
summary(nv)
# Min. 1st Qu. Median Mean 3rd Qu. Max.
# 3.00 14.00 32.00 43.89 60.00 374.00
for(i in 1:400) fv[i]=length(nv[nv==i])
fv # [1] 0 0 295 225 2 .... 0 0 0 0
plot(1:400,fv); sum(fv[seq(2,400,2)])/10000 # 0.502 OK
```

```
# (c)
```

```
trialfun2=function(n=38){
  v=sample(1:6,n,replace=T); y=0; for(i in 1:(n-2)){ ss=v[i:(i+2)]
    if( length( grep(1,ss)+grep(2,ss)+grep(3,ss) ) ==1) y=y+1 }
  y }
set.seed(472); trialfun2() # 3 OK
date(); set.seed(224); J=10000; yv=rep(NA,J); for(j in 1:J) yv[j]=trialfun2()
date() # Took 23 secs
me=mean(yv); se=sd(yv); ci=me+c(-1,1)*qnorm(0.975)*se/sqrt(J)
c(me,se,ci) # 1.0020000 1.1256655 0.9799374 1.0240626 OK
```

```
# (d)
```

```
date(); set.seed(264); J=100000; yv=rep(NA,J); for(j in 1:J) yv[j]=trialfun2(n=6)
date() # Took 25 secs
phat=length(yv[yv==0])/J
pci=phat+c(-1,1)*qnorm(0.975)*sqrt(phat*(1-phat)/J)
c(phat,pci) # 0.9061000 0.9042921 0.9079079 OK
```


Solution to Problem 3

(a) Let N be the total number of persons in the sample, and let Y be the number of persons in the sample with criplea. Then:

$$N \sim \text{NegBin}(w, q) \text{ with mean } EN = \frac{w}{q}, \text{ variance } VN = \frac{w(1-q)}{q^2}$$

$$\text{and density } f(n) = \binom{n-1}{w-1} q^w (1-q)^{n-w}, n = w, w+1, w+2, \dots$$

$$(Y | n) \sim \text{Bin}(n, p) \text{ with mean } E(Y | n) = np, \text{ variance } V(Y | n) = np(1-p)$$

$$\text{and density } f(y | n) = \binom{n}{y} p^y (1-p)^{n-y}, y = 0, 1, \dots, n.$$

$$\text{So: } EY = EE(Y | N) = E(Np) = pEN = \boxed{p \frac{w}{q}}$$

$$VY = EV(Y | N) + VE(Y | N) = E\{Np(1-p)\} + V\{Np\}$$

$$= p(1-p)EN + p^2VN = p(1-p)\frac{w}{q} + p^2 \frac{w(1-q)}{q^2} = \boxed{p \frac{w}{q} \left(1 + \frac{p}{q} - 2p\right)}.$$

With $w = 200$, $q = 0.6$ and $p = 0.05$ these formulae yield $EY = \boxed{16.67}$ and $VY = \boxed{16.39}$.

$$(b) f(n | y) = \frac{f(n)f(y | n)}{f(y)} = cf(n)f(y | n) \text{ where } c = \frac{1}{f(y)} \text{ does not depend on } n$$

$$= c \binom{n-1}{1-1} q^1 (1-q)^{n-1} \binom{n}{0} p^0 (1-p)^{n-0} \text{ since } w = 1 \text{ and } y = 0$$

$$= dt^{n-1}, n = 1, 2, 3, \dots$$

where d does not depend on n and where $t = (1-q)(1-p) = 0.637$.

By considering a list of well-known discrete distributions, we see that N given $Y = y$ has a geometric distribution with parameter $k = 1 - t$. Thus:

$$\boxed{(N | y) \sim \text{Geo}(k), \text{ where } k = 0.363}$$

$$\boxed{f(n | y) = (1-k)^{n-1} k, n = 1, 2, 3, \dots}$$

$$\boxed{E(N | y) = 1/k = 2.755}.$$

(c) With $q = 0.7$ $n = 5$ and $y = 2$, the joint density of N and Y equals

$$f(n, y) = f(n)f(y | n) = \binom{5-1}{w-1} \left(\frac{7}{10}\right)^w \left(1 - \frac{7}{10}\right)^{5-w} \times \binom{5}{2} p^2 (1-p)^{5-2}.$$

So the likelihood function is

$$L(w, p) = \binom{4}{w-1} 7^w 3^{5-w} \times p^2 (1-p)^3, \quad 0 \leq p \leq 1, \quad w = 1, \dots, 5.$$

We see that the MLE of p is $\frac{2}{5}$, because this value maximises $p^2(1-p)^3$.

Now, $\binom{4}{w-1} 7^w 3^{5-w} = 567, 5292, 18522, 28812, 16807$ at $w = 1, 2, 3, 4, 5$,

respectively, with a maximum at $w = 4$. It follows that the MLE of w is $\boxed{4}$.

R Code for Problem 3

(a)

```
w=200; q=0.6; p=0.05; me=p*w/q; va=me*(1-2*p+p/q)
```

```
c(me,va) # 16.66667 16.38889
```

(b)

```
q=0.35; p=0.02; t=(1-q)*(1-p); c(t,1-t,1/(1-t)) # 0.637000 0.363000 2.754821
```

(c)

```
wv=1:5; choose(4, (1:5)-1) * 7^wv * 3^(5-wv) # 567 5292 18522 28812 16807
```

Solution to Problem 4

(a) We need to solve the equation $\mu = cEs$ for c , where

$$Es = E\left(\frac{\sigma}{\sqrt{n-1}}\sqrt{\frac{(n-1)s^2}{\sigma^2}}\right) = \frac{\mu}{\sqrt{n-1}}EU^{1/2} \quad \text{where } U = \frac{(n-1)s^2}{\sigma^2} \sim \chi^2(n-1).$$

$$\text{Now, } EU^{1/2} = \int_0^\infty u^{1/2} \frac{u^{\left(\frac{n-1}{2}\right)-1} e^{-\frac{u}{2}}}{\Gamma\left(\frac{n-1}{2}\right) 2^{\left(\frac{n-1}{2}\right)}} du = \frac{\Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right) 2^{\frac{1}{2}}} \int_0^\infty \frac{u^{\left(\frac{n}{2}\right)-1} e^{-\frac{u}{2}}}{\Gamma\left(\frac{n}{2}\right) 2^{\left(\frac{n}{2}\right)}} du = \frac{\Gamma\left(\frac{n}{2}\right)\sqrt{2}}{\Gamma\left(\frac{n-1}{2}\right)}.$$

$$\text{Therefore } Es = \frac{\mu}{k}\sqrt{\frac{2}{n-1}} \quad \text{where } k = \frac{\Gamma\left(\frac{n-1}{2}\right)}{\Gamma\left(\frac{n}{2}\right)}. \text{ So } \mu = cEs = c\frac{\mu}{k}\sqrt{\frac{2}{n-1}}.$$

$$\text{Thus } \boxed{c = k\sqrt{\frac{n-1}{2}}, \text{ where } k = \frac{\Gamma\left(\frac{n-1}{2}\right)}{\Gamma\left(\frac{n}{2}\right)}}.$$

$$\text{For the case } n = 6: \quad \Gamma\left(\frac{n-1}{2}\right) = \Gamma\left(\frac{5}{2}\right) = \frac{3}{2}\Gamma\left(\frac{3}{2}\right) = \frac{3}{2} \times \frac{1}{2}\Gamma\left(\frac{1}{2}\right) = \frac{3}{4}\sqrt{\pi}$$

$$\Gamma\left(\frac{n}{2}\right) = \Gamma(3) = 2! = 2, \quad k = \frac{3\sqrt{\pi}/4}{2} = \frac{3\sqrt{\pi}}{8} = 0.66467$$

$$c = k\sqrt{\frac{n-1}{2}} = \frac{3\sqrt{\pi}}{8}\sqrt{\frac{6-1}{2}} = \frac{3}{8}\sqrt{\frac{5\pi}{2}} = \boxed{1.050936}.$$

(b) Using results in (a), the variance of $\hat{\sigma}$ (as an unbiased estimator of $\sigma = \mu$) is

$$\begin{aligned} V\hat{\sigma} &= c^2Vs = c^2\{Es^2 - (Es)^2\} = c^2\left\{\mu^2 - \left(\frac{\mu}{k}\sqrt{\frac{2}{n-1}}\right)^2\right\} = \mu^2c^2\left(1 - \frac{2}{k^2(n-1)}\right) \\ &= \mu^2k^2\left(\frac{n-1}{2}\right)\left(1 - \frac{2}{k^2(n-1)}\right) = \mu^2\left\{\left(\frac{n-1}{2}\right)k^2 - 1\right\}. \end{aligned}$$

Also $V\bar{y} = \frac{\sigma^2}{n} = \frac{\mu^2}{n}$. So, the efficiency of \bar{y} relative to $\hat{\sigma}$ is

$$r = \frac{V\hat{\sigma}}{V\bar{y}} = n \left(\left(\frac{n-1}{2} \right) k^2 - 1 \right) \text{ where } k = \frac{\Gamma\left(\frac{n-1}{2}\right)}{\Gamma\left(\frac{n}{2}\right)}.$$

By results in (a) with $n = 6$ we get $r = 6 \left(\left(\frac{6-1}{2} \right) \frac{9\pi}{64} - 1 \right) = 6 \left(\frac{45\pi}{128} - 1 \right) = \boxed{0.6268}$.

Note: The formula for k can be expressed in various other ways, for example using the results that $\Gamma\left(\frac{m}{2}\right) = \left(\frac{m}{2} - 1\right)!$ if m is even and $\Gamma\left(\frac{m}{2}\right) = \frac{(m-2)!!\sqrt{\pi}}{2^{(m-1)/2}}$ if m is odd.

Here, $!!$ denotes the double factorial function such that:

$$m!! = m \times (m-2) \times \dots \times 4 \times 2 \text{ if } m \text{ is even}$$

$$m!! = m \times (m-2) \times \dots \times 3 \times 1 \text{ if } m \text{ is odd.}$$

(c) The joint density is

$$f(y) = \prod_{i=1}^n \frac{1}{\mu\sqrt{2\pi}} \exp\left\{-\frac{1}{2\mu^2}(y_i - \mu)^2\right\} = \mu^{-n} (2\pi)^{-n/2} \exp\left\{-\frac{1}{2\mu^2} \sum_{i=1}^n (y_i - \mu)^2\right\},$$

and therefore the likelihood function is $L(\mu) = \mu^{-n} \exp\left\{-\frac{1}{2\mu^2} \sum_{i=1}^n (y_i - \mu)^2\right\}$.

So the log-likelihood function is $l(\mu) = -n \log \mu - \frac{1}{2} \mu^{-2} \sum_{i=1}^n (y_i - \mu)^2$.

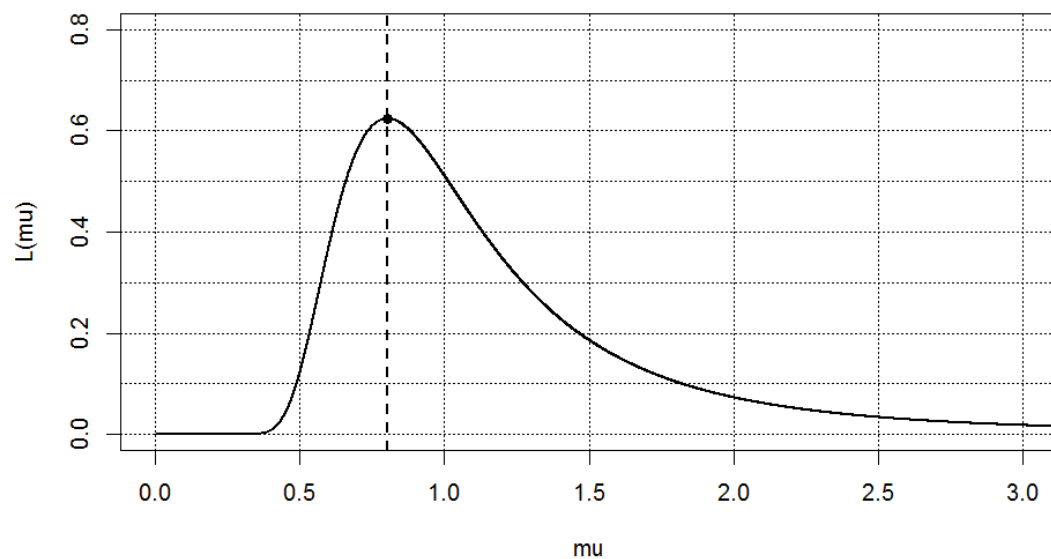
$$\begin{aligned} \text{Then, } l'(\mu) &= -\frac{n}{\mu} - \frac{1}{2} \left\{ \mu^{-2} 2 \sum_{i=1}^n (y_i - \mu)^1 (-1) + (-2\mu^{-3}) \sum_{i=1}^n (y_i - \mu)^2 \right\} \\ &= -\frac{n}{\mu} + \left\{ \mu^{-2} (n\bar{y} - n\mu) + \mu^{-3} (na - 2n\bar{y}\mu + n\mu^2) \right\} \text{ where } a = \frac{1}{n} \sum_{i=1}^n y_i^2 \\ &= \cancel{-\frac{n}{\mu}} + \left\{ \mu^{-2} n\bar{y} - \mu^{-2} n\mu + \mu^{-3} na - \mu^{-3} 2n\bar{y}\mu + \cancel{\mu^{-3} n\mu^2} \right\} \\ &= -\mu^{-2} n\bar{y} - \mu^{-2} n\mu + \mu^{-3} na = -n\mu^{-3} (\mu^2 + \mu\bar{y} - a). \end{aligned}$$

Setting $l'(\mu)$ to zero defined a quadratic equation whose solution yields the MLE,

$$\hat{\mu} = \frac{-\bar{y} + \sqrt{\bar{y}^2 + 4a}}{2}, \text{ where } a = \frac{1}{n} \sum_{i=1}^n y_i^2 \quad (\text{since } \mu > 0).$$

If $(y_1, \dots, y_n) = (1.2, 1.7, 0.1)$ then $n = 3$, $\bar{y} = 1$, $a = 1.44667$ and $\hat{\mu} = \boxed{0.8026}$.

The figure below shows $L(\mu)$ and shows the MLE with a vertical dashed line.



(d) The probability of a Type I error is

$$\begin{aligned} 0.05 &= P(S^2 > k) = P\left(\frac{(n-1)S^2}{\sigma^2} > \frac{(2-1)k}{3^2}\right) = P\left(\chi^2(1) > \frac{k}{9}\right) \\ &= 2P\left(Z > \frac{\sqrt{k}}{3}\right) \quad \text{where } Z \sim N(0,1) \text{ since } Z^2 \sim \chi^2(1). \end{aligned}$$

But $0.05 = 2P(Z > 1.96)$. So we equate $\frac{\sqrt{k}}{3} = 1.96$ to get $\sqrt{k} = 5.88$ and $k = \boxed{34.57}$.

Note: We could also get this value of k by looking up the chi-square tables with one degree of freedom, locating 3.84146 and multiplying this upper 0.05-quantile by 9.

The probability of a Type II error is

$$\begin{aligned}\beta(\mu) &= P(S^2 < k) = P\left(\frac{(n-1)S^2}{\sigma^2} < \frac{(2-1)k}{\mu^2}\right) = 1 - P\left(\chi^2(1) > \frac{k}{\mu^2}\right) \\ &= 1 - 2P\left(Z > \frac{\sqrt{k}}{\mu}\right) \quad (\mu > 0, \mu \neq 3).\end{aligned}$$

So the power function is $\boxed{Power(\mu) = 2P\left(Z > \frac{\sqrt{k}}{\mu}\right), \mu > 0.}$

Thus: $Power(2) = 2P\left(Z > \frac{5.88}{2}\right) = 2P(Z > 2.94) = 2(0.0016) = \boxed{0.0032}$

$$Power(3) = 2P\left(Z > \frac{5.88}{3}\right) = \alpha = \boxed{0.0500}$$

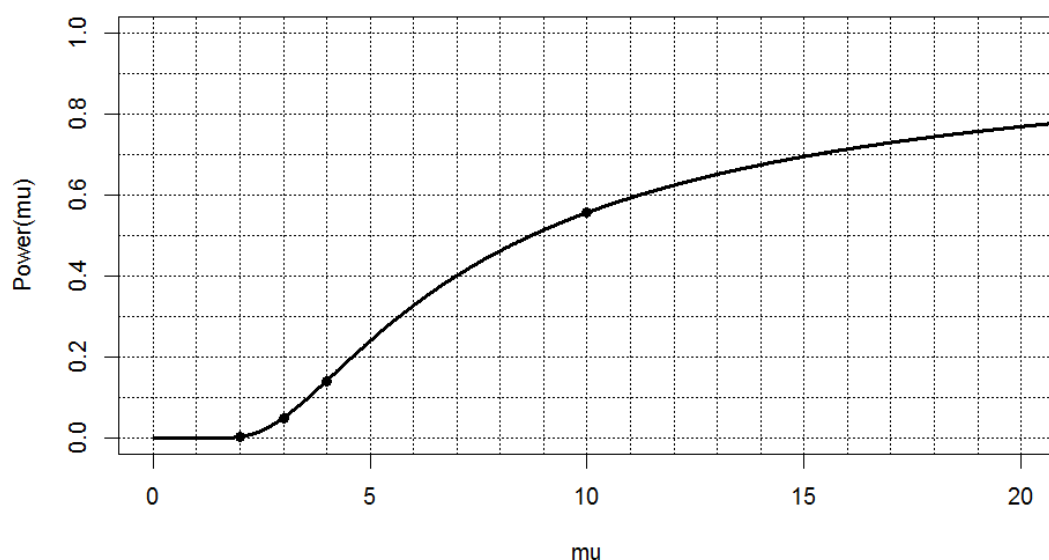
$$Power(4) = 2P\left(Z > \frac{5.88}{4}\right) = 2P(Z > 1.47) = 2(0.0708) = \boxed{0.1416}$$

$$Power(10) = 2P\left(Z > \frac{5.88}{10}\right) = 2P(Z > 0.588) = 2(0.2776) = \boxed{0.5552}.$$

Note: These values were obtained using tables that are accurate to two decimals.

The exact values correct to four decimals are 0.0033, 0.0500, 0.1416 and 0.5566.

Below is a sketch of the power function with points marked at $\mu = 2, 3, 4$ and 10.



R Code for Problem 4

(a)

```
n=6; k=gamma((n-1)/2)/gamma(n/2); c=k*sqrt((n-1)/2)
```

```
c(k,c) # 0.6646702 1.0509359
```

(b)

```
n*( ((n-1)/2)*k^2-1 ) # 0.626797
```

Checking via Monte Carlo

```
mu=10; J=10000; ybarvec=rep(NA,J); sighatvec=rep(NA,J); set.seed(294); for(j in 1:J){
```

```
  yv=rnorm(n,mu,mu); ybarvec[j]=mean(yv); sighatvec[j]=c*sd(yv) }
```

```
c(mean(ybarvec), mean(sighatvec)) # 10.01976 10.02061 OK
```

```
var(sighatvec)/var(ybarvec) # 0.6226136 OK
```

(c)

```
Lfun=function(mu,y){ n=length(y); mu^(-n)*exp(-0.5*sum((y-mu)^2)/mu^2) }
```

```
X11(w=8,h=5); plot(c(0,3),c(0,0.8),type="n",xlab="mu",ylab="L(mu)");
```

```
  abline(h=seq(0,1,0.1),lty=3); abline(v=seq(0,5,0.5),lty=3)
```

```
muv=seq(0,10,0.001); m=length(muv); Lv=muv
```

```
y=c(1.2,1.7, 0.1); ybar=mean(y); a=mean(y^2); c(ybar,a) # 1.000000 1.446667
```

```
for(i in 1:m) Lv[i]=Lfun(mu=muv[i],y=y); lines(muv,Lv,lty=1,lwd=2)
```

```
mle=0.5*(-ybar+sqrt(ybar^2+4*a)); mle # 0.8025616
```

```
points(mle, Lfun(mle,y), pch=16); abline(v=mle,lty=2,lwd=2)
```

Another example

```
y=c(-0.3,-0.5,0.1); ybar=mean(y); a=mean(y^2); c(ybar,a) # -0.2333333 0.1166667
```

```
for(i in 1:m) Lv[i]=Lfun(mu=muv[i],y=y); lines(muv,Lv,lty=3,lwd=2)
```

```
mle=0.5*(-ybar+sqrt(ybar^2+4*a)); mle # 0.4776068
```

```
points(mle, Lfun(mle,y), pch=16); abline(v=mle,lty=2,lwd=2) # OK
```

(d)

```
k=(3*qnorm(0.975))^2; k # 34.57313
```

```
2*(1-pnorm(sqrt(k)/c(2,3,4,10))) # 0.003282695 0.050000000 0.141569070 0.556539545
```

```
X11(w=8,h=5); plot(c(0,20),c(0,1),type="n",xlab="mu",ylab="Power(mu)");
  abline(h=seq(0,1,0.1),lty=3); abline(v=seq(0,20,1),lty=3)
muvec=seq(0.01,21,0.01); betavec=2*(1-pnorm(sqrt(k)/muvec))
lines(muvec,betavec,lty=1,lwd=3)
points(c(2,3,4,10), 2*(1-pnorm(sqrt(k)/c(2,3,4,10))), pch=16)
```

Solution to Problem 5

(a) $f_X(x) = \frac{f_Y(x)}{P(Y > 1)}$, where $P(Y > 1) = 1 - P(Y \leq 1) = 1 - \frac{2}{3} - \frac{1}{3} \times \frac{1}{2} = \frac{1}{6}$.

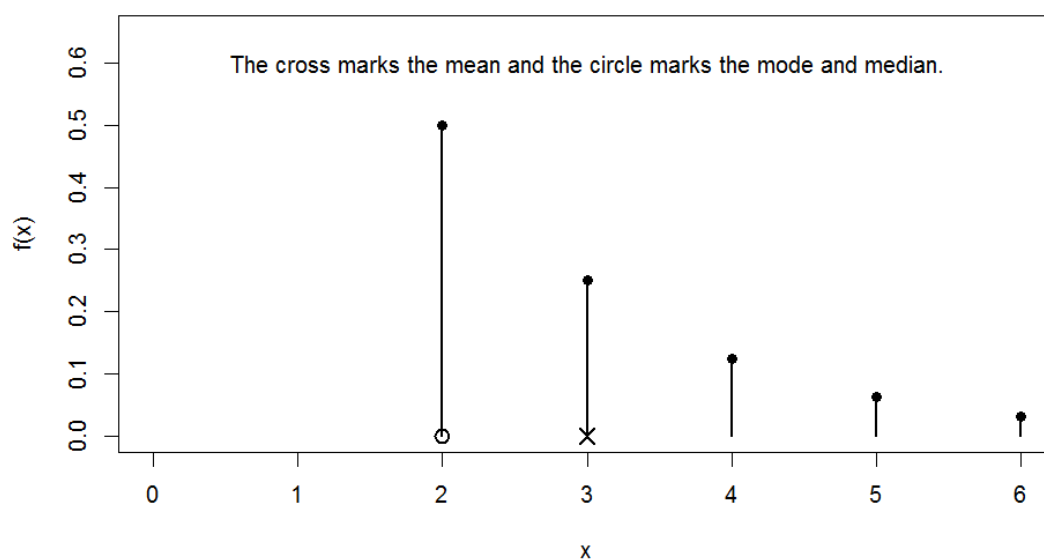
So $f_X(x) = 6 \times \frac{1}{3} \times \frac{1}{2^x} = \left(\frac{1}{2}\right)^{x-1}$, $x = 2, 3, 4, \dots$ We see that $M = m = \boxed{2}$.

We may write $X = T + 1$, where $T \sim \text{Geo}(1/2)$ with mean $ET = \frac{1}{1/2} = 2$,

variance $VT = \frac{1 - 1/2}{(1/2)^2} = 2$ and second raw moment $ET^2 = VT + (ET)^2 = 6$.

Thus $\mu = EX = ET + 1 = 2 + 1 = \boxed{3}$, and $\sigma^2 = VX = VT = \boxed{2}$.

(a) $X = (Y | Y > 1)$



$$(b) f_X(x) = \begin{cases} P(Y \leq 1), & x = 0 \\ f_Y(x), & x = 2, 3, 4, \dots \end{cases} \text{ where } P(Y \leq 1) = \frac{2}{3} + \frac{1}{3} \times \frac{1}{2} = \frac{5}{6}.$$

$$\text{So } f_X(x) = \begin{cases} 5/6, & x = 0 \\ \frac{1}{3} \left(\frac{1}{2} \right)^x, & x = 2, 3, 4, \dots \end{cases}. \text{ We see that } M = m = \boxed{0}.$$

$$\text{Also: } \mu = 0 \times \frac{5}{6} + \frac{1}{3} \sum_{x=2}^{\infty} x \left(\frac{1}{2} \right)^x = \frac{1}{3} \sum_{t=1}^{\infty} (t+1) \left(\frac{1}{2} \right)^{t+1} \text{ where } t = x - 1$$

$$= \frac{1}{6} \sum_{t=1}^{\infty} (t+1) \left(\frac{1}{2} \right)^t = \frac{1}{6} E(T+1) \text{ where } T \sim Geo(1/2)$$

$$= \frac{1}{6} (ET + 1) = \frac{1}{6} (2 + 1) = \boxed{1/2}$$

$$EX^2 = 0^2 \times \frac{5}{6} + \frac{1}{3} \sum_{x=2}^{\infty} x^2 \left(\frac{1}{2} \right)^x = \frac{1}{3} \sum_{t=1}^{\infty} (t+1)^2 \left(\frac{1}{2} \right)^{t+1}$$

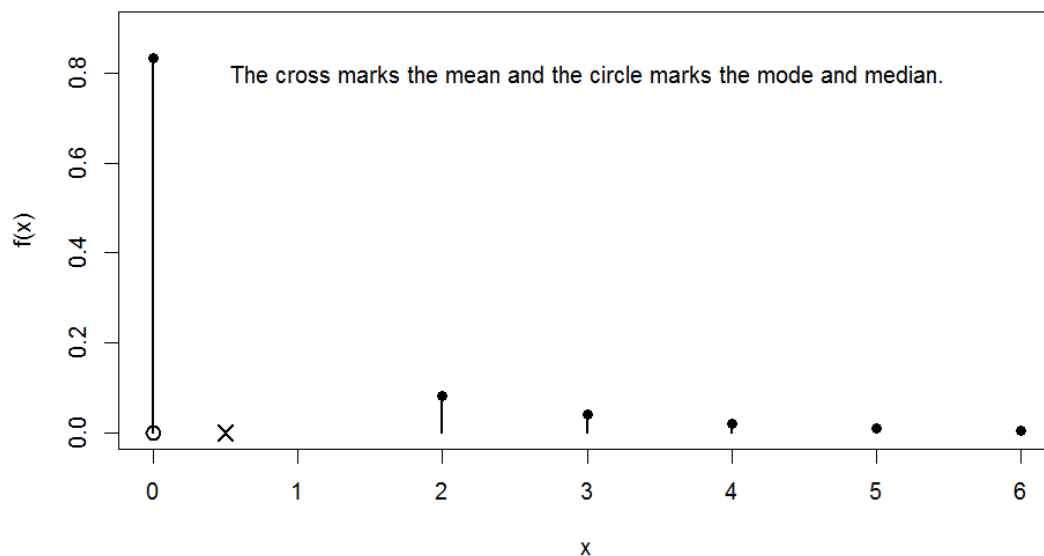
$$= \frac{1}{6} \sum_{t=1}^{\infty} (t+1)^2 \left(\frac{1}{2} \right)^t = \frac{1}{6} E\{(T+1)^2\} = \frac{1}{6} (ET^2 + 2ET + 1)$$

$$= \frac{1}{6} (VT + (ET)^2 + 2ET + 1)$$

$$= \frac{1}{6} (2 + 2^2 + 2 \times 2 + 1) = 11/6$$

$$VX = EX^2 - (EX)^2 = \frac{11}{6} - \left(\frac{1}{2} \right)^2 = \boxed{19/12}.$$

(b) $X = Y * I(Y > 1)$



(c) $X = |Y - 1|$ has pdf given by: $f_X(0) = P(Y = 1) = 1/6$

$$f_X(1) = P(Y = 0) + P(Y = 2) = \frac{1}{3} + \frac{1}{12} = 5/12$$

$$f_X(2) = P(Y = -1) + P(Y = 3) = \frac{1}{6} + \frac{1}{24} = \frac{1}{2} \left(\frac{5}{12} \right), \text{ etc.}$$

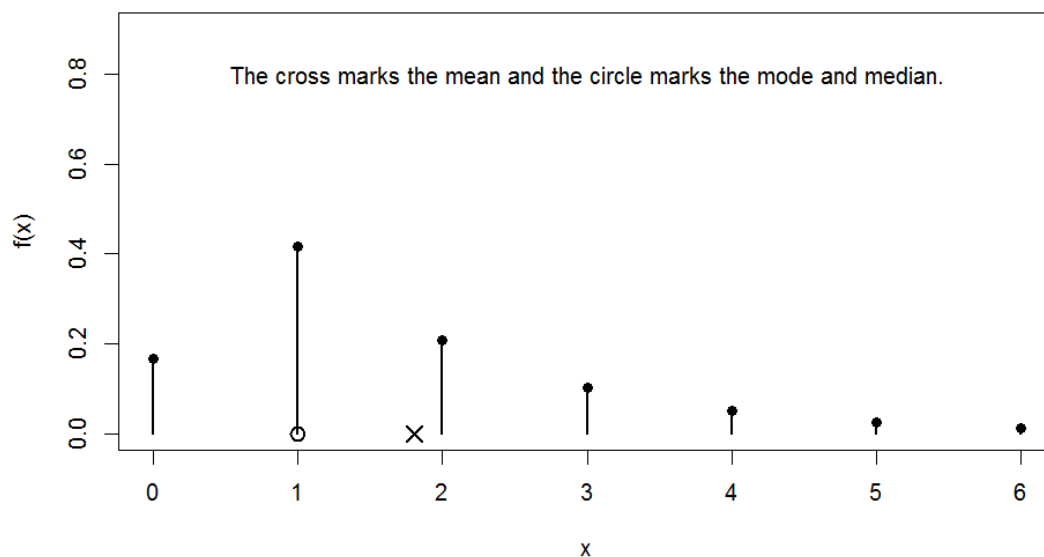
We see that
$$f_X(x) = \begin{cases} 1/6, & x = 0 \\ \frac{5}{12} \left(\frac{1}{2} \right)^{x-1}, & x = 1, 2, 3, \dots \end{cases} \text{ and } M = m = \boxed{1}.$$

Also:
$$\begin{aligned} \mu = EX &= 0 \times \frac{1}{6} + 1 \times \frac{5}{12} + 2 \times \frac{5}{12} \left(\frac{1}{2} \right) + 3 \times \frac{5}{12} \left(\frac{1}{2} \right)^2 + \dots \\ &= \frac{5}{6} \left\{ 1 \left(\frac{1}{2} \right) + 2 \left(\frac{1}{2} \right)^2 + 3 \left(\frac{1}{2} \right)^3 + \dots \right\} = \frac{5}{12} ET \text{ where } T \sim Geo(1/2) \\ &= \frac{5}{12} \times 2 = \boxed{5/6} \end{aligned}$$

$$\begin{aligned} EX^2 &= 0^2 \times \frac{1}{6} + 1^2 \times \frac{5}{12} + 2^2 \times \frac{5}{12} \left(\frac{1}{2} \right) + 3^2 \times \frac{5}{12} \left(\frac{1}{2} \right)^2 + \dots \\ &= \frac{5}{6} \left\{ 1^2 \left(\frac{1}{2} \right) + 2^2 \left(\frac{1}{2} \right)^2 + 3^2 \left(\frac{1}{2} \right)^3 + \dots \right\} = \frac{5}{12} ET^2 = \frac{5}{12} \times 6 = \frac{5}{2} \end{aligned}$$

$$VX = EX^2 - (EX)^2 = \frac{5}{2} - \left(\frac{5}{6} \right)^2 = \boxed{65/36}.$$

(c) $X = |Y - 1|$



Summary table (not required):

X	μ	M	m	σ^2
$(Y Y > 1)$	3	2	2	2
$YI(Y > 1)$	1/2	0	0	19/12
$ Y - 1 $	5/6	1	1	65/36

R Code for Problem 5

X11(w=8,h=5)

(a)

```
xvec=2:6; plot(c(0,6),c(0,0.65),type="n",xlab="x",ylab="f(x)",
  main="(a) X = ( Y | Y > 1)")
fvec=(1/2)^(xvec-1); for(i in 1:length(xvec)){
  x=xvec[i]; lines(c(x,x),c(0,fvec[i]),lwd=2); points(x,fvec[i],pch=16) }
points(c(2,3),c(0,0),pch=c(1,4),cex=1.5,lwd=2)
text(3,0.6,"The cross marks the mean and the circle marks the mode and median.")
```

(b)

```
xvec=c(0,2:6); plot(c(0,6),c(0,0.9),type="n",xlab="x",ylab="f(x)",
  main="(b) X = Y * I ( Y > 1 )")
fvec=c( 5/6, (1/3)*(1/2)^(2:6) ); for(i in 1:length(xvec)){
  x=xvec[i]; lines(c(x,x),c(0,fvec[i]),lwd=2); points(x,fvec[i],pch=16) }
points(c(0,1/2),c(0,0),pch=c(1,4),cex=1.5,lwd=2)
text(3,0.8,"The cross marks the mean and the circle marks the mode and median.")
```

(c)

```
xvec=c(0:6); plot(c(0,6),c(0,0.9),type="n",xlab="x",ylab="f(x)",
  main="(c) X = | Y - 1 |")
fvec=c( 1/6, (5/12)*(1/2)^(0:5) ); for(i in 1:length(xvec)){
  x=xvec[i]; lines(c(x,x),c(0,fvec[i]),lwd=2); points(x,fvec[i],pch=16) }
points(c(1,65/36),c(0,0),pch=c(1,4),cex=1.5,lwd=2)
text(3,0.8,"The cross marks the mean and the circle marks the mode and median.")
```

Solution to Problem 6

(a) Here: $\bar{x} = 2/3$, $\bar{y} = 8/3$,

$$S_{xx} = \sum x_i^2 - n\bar{x}^2 = 2 - 3(2/3)^2 = 2/3$$

$$S_{xy} = \sum x_i y_i - n\bar{x}\bar{y} = 7 - 3(2/3)(8/3) = 5/3,$$

$$b = S_{xy} / S_{xx} = \boxed{2.5}$$

$$a = \bar{y} - b\bar{x} = \boxed{1}$$

$$\hat{m} = a + 0.5b = \boxed{2.25}.$$

Next, $V\hat{m} = 1 \left(1 + \frac{1}{3} + \frac{(0.5 - 2/3)^2}{2/3} \right) = \frac{11}{8}$, and so a 95% prediction interval for m is

$$\left(\hat{m} \pm 1.96\sqrt{V\hat{m}} \right) = \left(2.25 \pm 1.96\sqrt{11/8} \right) = \left(2.25 \pm 2.298 \right) = \boxed{(-0.048, 4.548)}.$$

(b) The sum of squares for error is $SSE = \sum (y_i - au_i - bx_i)^2$. This has derivatives:

$$\frac{\partial SSE}{\partial a} = \sum 2(y_i - au_i - bx_i)^1(-u_i) = -2\{T_{yu} - aT_{uu} - bT_{xu}\}$$

where $T_{yu} = \sum y_i u_i$, $T_{uu} = \sum u_i u_i = \sum u_i^2$, etc. (T stands for "Total")

$$\frac{\partial SSE}{\partial b} = \sum 2(y_i - au_i - bx_i)^1(-x_i) = -2\{T_{yx} - aT_{ux} - bT_{xx}\}.$$

Setting these derivatives to zero yields $b = \frac{T_{yu} - aT_{uu}}{T_{xu}}$ and $b = \frac{T_{yx} - aT_{ux}}{T_{xx}}.$

Equating these two expressions for b , we obtain $a = \frac{T_{xx}T_{uy} - T_{ux}T_{xy}}{T_{uu}T_{xx} - T_{ux}^2}.$

For the given data, we get $a = \frac{2 \times 9 - 2 \times 7}{5 \times 2 - 2^2} = \boxed{2/3}$ and $b = \frac{7 - (2/3)2}{2} = \boxed{17/6}.$