

HW 9

8th April 2005

5.1-1

a) Sum out all the y's $f(x) = \frac{10+4x}{32} = \frac{5+2x}{16}$

b) Sum up all the x's $f(y) = \frac{3+2y}{32}$

c) $P(X > Y) = P(2,1) = \frac{3}{32}$

d) $P(Y = 2X) = P(1,2) + P(2,4) = \frac{3}{32} + \frac{6}{32} = \frac{9}{32}$

e) $P(X + Y = 3) = P(1,2) + P(2,1) = \frac{6}{32}$

f) $P(X \leq 3 - Y) = P(1,1) + P(1,2) + P(2,1) = \frac{8}{32}$

g) No $P(1,1) = \frac{2}{32}$ but $P(X=1)P(Y=1) = \frac{14}{32} * \frac{5}{32}$ which are not equivalent and thus X,Y are dependent.

5.1-9

$f(x) = \int f(x,y)dy$ with $0 \leq y < \infty$ which equals $-2e^{-x-y}$ evaluated at 0 and infinity. So $f(x) = -2e^{-x}$

$f(y) = \int f(x,y)dx$ with $0 \leq x \leq y$ This equals $-2e^{-x-y}$ evaluated at 0 and y.
 $f(y) = -2e^{-2y} + 2e^{-y} = 2e^{-y}(1 - e^{-y})$

X and Y are not independent. $f(x,y)$ does not equal $f(x)f(y)$.

5.1-10

a) $f(x) = \int (x+y)dy$ where y is in $[0,1]$

$f(x) = x + \frac{1}{2}$

$f(y)$ is computed the same way so $f(y) = y + \frac{1}{2}$

Obviously $f(x)f(y)$ does not equal $f(x,y)$

b) $\mu_x = \int x(x + \frac{1}{2})dx = \frac{x^3}{3} + \frac{x^2}{4}$ evaluated at 0 and 1 equals $\frac{1}{3} + \frac{1}{4} = \frac{7}{12}$

c) $\mu_y = \frac{7}{12}$ by the same reasoning

d) $E(X^2) = \int x^2(x + \frac{1}{2})dx = \frac{x^4}{4} + \frac{x^3}{6} = \frac{5}{12}$

$\sigma_x^2 = \frac{5}{12} - (\frac{7}{12})^2 = \frac{11}{144}$

$\sigma_y^2 = \frac{11}{144}$

5.2-1

$$\begin{aligned}\mu_x &= \sum x\left(\frac{2x+5}{16}\right) = \frac{7}{16} + \frac{18}{16} = \frac{25}{16} \\ \mu_y &= \sum y\left(\frac{2y+3}{32}\right) = \frac{5}{32} + \frac{14}{32} + \frac{27}{32} + \frac{44}{32} = \frac{90}{32} \\ E(X^2) &= \sum x^2\left(\frac{2x+5}{16}\right) = \frac{7}{16} + \frac{36}{16} = \frac{43}{16} \\ \sigma_x^2 &= \frac{46}{16} - \left(\frac{25}{16}\right)^2 = \frac{63}{256} \\ E(Y^2) &= \sum y^2\left(\frac{2y+3}{32}\right) = \frac{5}{32} + \frac{28}{32} + \frac{81}{32} + \frac{176}{32} = \frac{290}{32} \\ \sigma_y^2 &= \frac{290}{32} - \left(\frac{90}{32}\right)^2 = \frac{295}{256} = 1.15 \\ E(XY) &= \sum \sum xy\left(\frac{x+y}{32}\right) = \frac{2}{32} + \frac{6}{32} + \frac{12}{32} + \frac{20}{32} + \frac{6}{32} + \frac{16}{32} + \frac{30}{32} + \frac{48}{32} = \frac{140}{32} = \frac{35}{8} = 4.375 \\ Cov(X, Y) &= 4.375 - \frac{25}{16}\left(\frac{90}{32}\right) = -.0195 \\ \rho &= \frac{Cov(x, y)}{\sqrt{\sigma_x^2 \sigma_y^2}} = \frac{-.0195}{\sqrt{.2461 * 1.15}} = -.0367\end{aligned}$$

This tells us that X and Y are dependent.

6.1-10

a) First compute $P(X < 85) = P(Z < \frac{85-75}{10} = 1)$ where Z is a $N(0,1)$ r.v. $P(Z < 1.5) = .8413$. Since all X 's are independent and identical.
 $P(\max X_i < 85) = P(Z < 1)^4 = .8413^4 = .501$

b) Because of independence this equals $P(60 < X_1 < 80) * P(75 < X_3 < 90) = P(-1.5 < Z < .5) * P(0 < Z < 1.5) = (.6915 - .0668) * (.9332 - .5) = .2706$

c) because of independence and identical dist $E(X_1^2 X_2^2 X_3^2) = E(X_1^2)^3 = (100 + 75^2)^3 = 1.876 * 10^{11}$

d) $E(2X_1 + 3X_2 + X_3 + X_4) = E(7X_1) = 7 * 75 = 525$

6.2-5

$$\begin{aligned}E(Y) &= -2E(X_1) + E(X_2) = 1 \\ Var(Y) &= 4Var(X_1) + Var(X_2) = 36 + 25 = 61\end{aligned}$$

6.2-6

$$\begin{aligned}M_Y(t) &= E(e^{tY}) = E(e^{t(X_1+X_2)}) = E(e^{tX_1})E(e^{tX_2}) \\ \text{Remembering that the moment generatin function for a binomial equals } (1 - p + pe^t)^n \\ M_Y(t) &= (1 - p + pe^t)^{(n_1+n_2)} \\ \text{This tells us that Y is distributed as a } b(n_1 + n_2, p)\end{aligned}$$

6.2-7

a) Using logic from problem 6
 $M_y(t) = M_{X_1}(t) * M_{X_2}(t) * M_{X_3}(t) = e^{2(e^t-1)} e^{e^t-1} e^{4(e^t-1)}$

$$= e^{7(e^t - 1)}$$

b) Which tells us Y is poisson(7)

$$c) P(3 \leq Y \leq 9) = .83 - .03 = .8$$

6.2-9

$M_Y(t) = E(e^{tY}) = E(e^{tX_1})^5$ since each X_i has the same value for p.

$$M_X(t) = \frac{pe^t}{1-(1-p)e^t} \text{ since } p=1/3 \quad M_Y(t) = \frac{(pe^t)^5}{(1-(1-p)e^t)^5}$$

This is just the MGF of a negative binomial with parameters $p=1/3$ and $r=5$

6.2-10

$$M_Y(t) = E(e^{tY}) = E(e^{tX})^h = \left(\frac{1}{1-\theta t}\right)^h \text{ This is the MGF of a Gamma}(h, \theta)$$

The mean of a gamma is $h\theta$

6.2-11

$$M_Y(t) = E(e^{tY}) = E(e^{tX_1})^3 = \left(\frac{1}{1-5t}\right)^3 = \frac{1}{(1-5t)^3}$$

This is distributed as a gamma(21, 5)

6.2-14

X is some discrete random distribution that takes on values $\{0, 1, 2, \dots\}$ and Y is another discrete random distribution that is independent of X and can take on the same values. X is distributed according to the function $f(x)$ and Y is distributed according to $g(y)$.

We now can create a set $A = \{(x, y) : x+y=w\}$

$A = \{(0, w), (1, w-1), \dots, (w, 0)\}$ This set consists of $w+1$ mutually exclusive events. Now we just need to find the probability of these $w+1$ events.

$P(x, w-x)$ is simply $f(x) * g(w-x)$ because of independence. Summing every event in A up will give the $h(w)$ in the text.

$$h(w) = \sum f(x)g(w-x)$$

6.2-19

X_1 is the random variable where we see how many rolls until we get one new face. Since we have not observed any rolls yet $X_1 = 1$ always. This fits with a geometric(1). X_2 is the r.v where we see how many rolls it will take to get a different face besides the one that we already saw. There is a $3/4$ chance that we get a different face than the first roll. We are going to roll until we get that new face. This is equivalent to a geometric($3/4$). X_3 is the r.v where we see how

many rolls it takes to see a different face than the 2 other faces we have seen. It is distributed as a geometric(1/2). X_4 is the r.v where we roll till we observe the last face and is distributed as a geometric(1/4). X_2, X_3, X_4 can take on all positive integer values.

$$b) E(Y) = E(X_1) + E(X_2) + E(X_3) + E(X_4)$$

Since each is geometric(p) the expected value of a geometric r.v is $\frac{1}{p}$ and its variance is $\frac{1-p}{p^2}$.

$$E(Y) = 1 + \frac{4}{3} + 2 + 4 = \frac{25}{3}$$

Since the X's are independent, $Var(Y) = Var(X_1) + Var(X_2) + Var(X_3) + Var(X_4)$

$$= 0 + \frac{.25}{.75^2} + \frac{.5}{.5^2} + \frac{.75}{.25^2} = \frac{130}{9} = 14.44$$

c) To calculate these probabilities we need to look at all the combinations of each X that will add up to the Y. We know that X_1 must equal 1. We then combine this and look at the distribution function for each X, which is $f(x) = (1-p)^{1-x}p$

$$P(Y = 4) = P(1,1,1,1) = 1 * \frac{3}{4} * \frac{1}{2} * \frac{1}{4} = \frac{3}{32}$$

$$P(Y = 5) = P(1,1,1,2) + P(1,1,2,1) + P(1,2,1,1) = \frac{144}{1024}$$

$$P(Y = 6) = P(1,1,1,3) + P(1,1,3,1) + P(1,3,1,1) + P(1,1,2,2) + P(1,2,1,2) + P(1,2,2,1) = \frac{150}{1024}$$

$$P(Y = 7) = P(1,1,1,4) + P(1,1,4,1) + P(1,4,1,1) + P(1,1,2,3) + P(1,2,1,3) + P(1,2,3,1) + P(1,1,3,2) + P(1,3,2,1) + P(1,3,1,2) + P(2,2,2,2) = \frac{135}{1024}$$