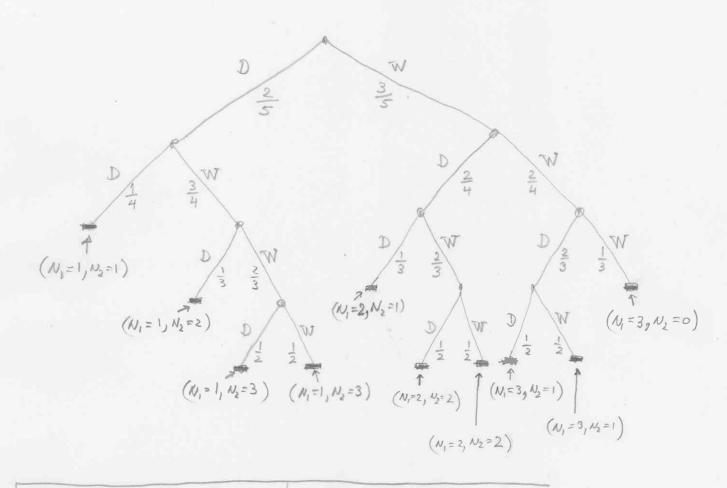
EE 342 HW7 SOLUTIONS

a CHE, PROB 6

LET W DENOTE "DRAWING A WORKING TRANSISTOR" LET D DENOTE "DRAWING A DEFECTIVE TRANSISTOR"

MAKE A PROBABILITY TREE



$$P_{N_1,N_2}(1,1) = \frac{2}{5} \cdot \frac{1}{4} = 0.1$$

$$P_{N_1,N_2}(2,1) = \frac{3}{5} \cdot \frac{2}{4} \cdot \frac{1}{3} = 0.1$$

$$P_{N_1,N_2}(1,2) = \frac{2}{5} \cdot \frac{3}{4} \cdot \frac{1}{3} = 0.1$$

$$P_{N_1,N_2}(2,2) = \frac{3}{5} \cdot \frac{2}{4} \cdot \frac{2}{3} = 0.2$$

$$P_{N_1,N_2}(1,3) = \frac{2}{5} \cdot \frac{3}{4} \cdot \frac{2}{3} = 0.2$$

$$P_{N_1, N_2}(2, 1) = \frac{3}{5} \cdot \frac{2}{4} \cdot \frac{1}{3} = 0.1$$

$$P_{N_1,N_2}(2,2) = \frac{3}{5} \cdot \frac{2}{4} \cdot \frac{2}{3} = 0.2$$

$$P_{N_1,N_2}(3,0) = \frac{3}{5} \cdot \frac{2}{4} \cdot \frac{1}{3} = 0.1$$

$$\iint_{-\infty}^{\infty} f_{X,Y}(x,y) dxdy = 1$$

$$\Rightarrow \int \int \frac{1}{x} dy dx = \int \frac{1}{x} \left[\int dy \right] dx = \int \frac{1}{x} \cdot x dx = 1$$

(a)
$$f_{Y}(y) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx = \int_{x}^{1} \frac{1}{x} dx = \ln x |_{y}^{1} = \begin{cases} \ln \frac{1}{y} & 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

(b)
$$f_{\overline{X}}(z) = \int_{-\infty}^{\infty} f_{\overline{X},Y}(x,y) dy = \int_{0}^{\infty} \pm dy = \int_{0}^{\infty} \int_{0}^{\infty} dy = \int_$$

(c)
$$E[X] = \int_{0}^{1} x \cdot f_{X}(x) dx = \int_{0}^{1} x dx = \frac{x^{2}}{2} \Big|_{0}^{1} = \frac{1}{2}$$

(a)
$$E[Y] = \iint_{X} f_{X,Y}(x,y) dx dy$$

$$= \iint_{X} f_{X,Y}(x,y) dy dx$$

$$= \iint_{X} f_{X,Y}(x,y) dy dx = \iint_{X} f_{X,Y}(x,y) dy dx$$

$$= \iint_{X} \left[\int_{X} y dy \right] dx - \int_{X} \frac{x^{2}}{x^{2}} dx = \frac{1}{2} \int_{X} x dx$$

C CH 6, PROB 20

$$f_{X,Y}(\alpha,y) = \begin{cases} xe^{-(x+y)} & x>0, y>0 \\ 0 & \text{otherwise} \end{cases}$$

if
$$x>0$$

$$f_{X}(x) = \int xe^{-(x+y)} dy = xe^{-x} \int e^{-y} dy = xe^{-x} \left[-e^{-y} \Big|_{0}^{\infty} \right] = xe^{-x}$$

$$= xe^{-x} \int xe^{-x} = xe^{-x} \int xe^{-x} = xe^{-x} \left[-e^{-y} \Big|_{0}^{\infty} \right] = xe^{-x}$$

$$= xe^{-x} \int xe^{-x} = xe^{-x} \int xe^{-x} = xe^{-x} \int xe^{-x} = xe^{-x} \left[-e^{-y} \Big|_{0}^{\infty} \right] = xe^{-x}$$

$$= xe^{-x} \int xe^{-x} = xe$$

if
$$y>0$$

$$f_{Y}(y) = \int x e^{ix+y} dx = e^{-y} \int x e^{-x} dx = e^{-y} \left[(-xe^{-x} - e^{-x}) \right]_{0}^{\infty} = e^{-y}$$

$$\Rightarrow f_{Y}(y) = \begin{cases} e^{-y} & y>0 \\ 0 & \text{otherwise} \end{cases}$$

Clearly
$$f_{X,Y}(x,y) = f_{X}(x) \cdot f_{Y}(y)$$
 So $X \& Y \text{ or } e$ independent

2°)
$$f_{X,Y}(x,y) = \begin{cases} 2 & 0 < x < y < 1 \\ 0 & otherwise \end{cases}$$

for occes |
$$f_{X}(x) = \int_{2}^{2} dy = 2 \cdot y |_{x}^{2} = 2(1-x)$$

for $xy < 1$
 $f_{Y}(y) = \int_{2}^{2} dx = 2y$

$$\begin{cases}
f_{X}(x) \cdot f_{Y}(y) = 2(1-x) \cdot 2y \\
\neq 2 = f_{X,Y}(x,y)
\end{cases}$$

> X, Y ARE NOT |

$$f_{X,Y}(x,y) = \begin{cases} 12 xy(1-x) & 0 < x < 1, 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

(a)
$$f_{\mathbf{X}}(\mathbf{x}) = \int_{0}^{1} f_{\mathbf{X},\mathbf{Y}}(\mathbf{x},y) dy = \int_{0}^{1} 12 x y(1-x) dy$$
 for $0 < x < 1$

$$= 12 x(1-x) \int_{0}^{1} y dy$$

$$= 12 x(1-x) \cdot \frac{1}{2} \qquad \Rightarrow f_{\mathbf{X}}(\mathbf{x}) = \begin{cases} 6x(1-x) & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$= 6 x (1-x)$$

$$f_{Y}(y) = \int_{0}^{1} f_{X,Y}(x,y) dx = \int_{0}^{1} 12 x y(1-x) dx$$

$$= 12y \int_{0}^{1} x(1-x) dx = 12y \left(\frac{x^{2}}{2} - \frac{x^{3}}{3}\right) \Big|_{0}^{1}$$

$$= 12 \cdot 9 \left[\frac{1}{2} - \frac{4}{3} \right]$$

Clearly
$$f_{\overline{X}}(x)f_{\overline{Y}}(y) = 6x(1-x)\cdot 2y = 12xy(1-x) = f_{\overline{X},\overline{Y}}(x,y)$$

So $X \otimes Y$ are independent

(4)

(b)
$$E[X] = \int_{0}^{1} x \cdot f_{X}(x) dx = \int_{0}^{1} x \cdot 6x(1-x) dx$$

$$= \left(6 \cdot \frac{x^3}{3} - 6 \cdot \frac{x^4}{4}\right) \left[\frac{1}{4}\right]$$

$$= \frac{6}{3} - \frac{6}{4} = 6 \cdot \left[\frac{1}{3} - \frac{1}{4} \right] = 6 \left[\frac{4-3}{12} \right] = \frac{1}{2} = F[X]$$

(c)
$$E[Y] = \int_{0}^{1} f_{Y}(y) dy = \int_{0}^{1} y \cdot 2y dy$$

$$=2\cdot\frac{4^3}{3}\begin{vmatrix}1\\3\end{vmatrix}=E[Y]$$

(d)
$$E[X^2] = \int_0^1 x^2 f_X(x) dx = \int_0^1 x^2 \cdot 6x(1-x) dx$$

$$=\left(6.\frac{x^{4}}{4}-6\frac{x^{5}}{5}\right)\Big|_{0}^{1}$$

$$= 6\left[\frac{1}{4} - \frac{1}{5}\right] = 6 \cdot \frac{5 - 4}{20} = \frac{3}{10} = E[X^{2}]$$

$$Vor(X) = E[X^2] - (E[X])^2 = \frac{3}{10} - (\frac{1}{2})^2 = \frac{3}{10} - \frac{1}{4}$$

$$Vor(X) = \frac{2}{40} = \frac{1}{20}$$

(e)
$$E[Y^2] = \int_0^1 y^2 f_Y(y) dy = \int_0^1 y^2 2y dy$$

 $= 2 \cdot \frac{y^4}{4} \Big|_0^1 = \Big[\frac{1}{2} = E[Y^2]\Big]$
 $Var(Y) = E[Y^2] - (E[Y])^2 = \frac{1}{2} - (\frac{2}{3})^2 = \frac{1}{2} - \frac{4}{9}$
 $\Big[Var(Y) = \frac{1}{18}\Big]$

PROVE:
$$P_{X,Y}(i,j) = \frac{P_{X|Y}(i|j)}{P_{Y|X}(j|k)}$$

START WITH

$$\sum_{k} \left[\frac{P_{X|Y}(k|i)}{P_{Y|X}(i|k)} \right] = \sum_{k} \frac{P_{X|Y}(k|i) \cdot P_{Y}(i)}{P_{Y|X}(i|k) \cdot P_{Y}(i)}$$

$$= \frac{1}{P_{Y}(i)} \cdot \left[\sum_{k} \frac{P_{X,Y}(k|i) \cdot P_{Y}(i)}{P_{Y}|X(i|i)} \right]$$

$$= \frac{1}{P_{Y}(i)} \cdot \left[\sum_{k} \frac{P_{X,Y}(k,j)}{P_{Y}|X(i|i)} \right] = \frac{1}{P_{Y}(i)} \cdot \left[\sum_{k} P_{X}(k) \right]$$

$$= \frac{1}{P_{Y}(i)} \cdot \left[\sum_{k} \frac{P_{X,Y}(k,j)}{P_{Y}|X(i|k)} \right] = \frac{1}{P_{Y}(i)} \cdot \left[\sum_{k} P_{X}(k) \right]$$

So
$$\left| \sum_{k} \frac{P_{X|Y}(k|i)}{P_{Y|X}(i|k)} \right| = \frac{1}{P_{Y}(i)}$$
 (*)

THEREFORE

$$\frac{P_{X|Y}(i|i)}{P_{X|Y}(k|i)} = \frac{P_{X|Y}(i|i)}{P_{Y}(i)} = \frac{P_{X|Y}(i|i)}{P_{Y}(i)}$$

$$= \frac{P_{X|Y}(i|i)}{P_{Y}(i|i)} = \frac{P_{X|Y}(i|i)}{P_{Y}(i)}$$

= Px, y (i, 5) proof done

2) a)
$$m_{\overline{x}}(t) = E[e^{tX}]$$

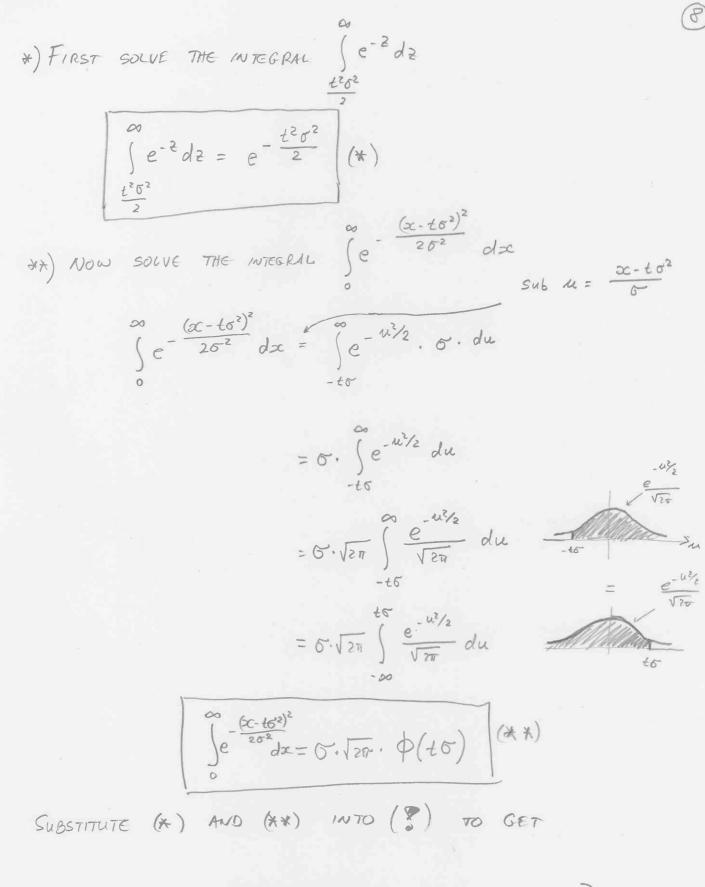
$$= \int_{e^{-tX}}^{\infty} f_{\overline{x}}(x) dx$$

$$= \int_{0}^{\infty} e^{-tx} \frac{x}{8^{2}} e^{-\frac{x^{2}}{25^{2}}} u(x) dx$$

$$= \int \frac{(x-t\sigma^2+t\sigma^2)}{\sigma^2} e^{+\frac{t^2\sigma^2}{2}} e^{-\frac{(x-t\sigma^2)^2}{2\sigma^2}} dx$$

$$= e^{\frac{t^2 \delta^2}{2}} \int_{0}^{\infty} \frac{(x-t\delta^2)}{\delta^2} e^{-\frac{(x-t\delta^2)^2}{2\delta^2}} dx + \int_{0}^{\infty} t \cdot e^{-\frac{(x-t\delta^2)^2}{2\delta^2}} dx$$

$$= e^{+t^2\sigma_{2}^{2}} \left[\int_{0}^{2\pi} e^{-2} dz + t \cdot \int_{0}^{2\pi} e^{-\frac{(x-t\sigma^{2})^{2}}{2\sigma^{2}}} dx \right]$$



 $u_{X}(t) = e^{+\frac{t^{2}\sigma^{2}}{2}} \left[e^{-\frac{t^{2}\sigma^{2}}{2}} + t \cdot \sigma \cdot \sqrt{2\pi} \, \phi(t\sigma) \right]$ $\left[u_{X}(t) = 1 + t \cdot \sigma \cdot \sqrt{2\pi} \cdot e^{\frac{t^{2}\sigma^{2}}{2}} \cdot \phi(t\sigma) \right]$

b)
$$u_{X} = \frac{\partial u_{X}(t)}{\partial t}\Big|_{t=0}$$

$$= \sigma \cdot \sqrt{2\pi} \cdot \frac{\partial \left[t \cdot e^{t^{2}\sigma_{X}^{2}} \cdot \phi(t\sigma)\right]}{\partial t}\Big|_{t=0}$$

$$= \sigma \cdot \sqrt{2\pi} \left[e^{t^{2}\sigma_{X}^{2}} \cdot \phi(t\sigma) \cdot \frac{\partial \left[t\right]}{\partial t} + t \cdot \phi(t\sigma) \frac{\partial \left(e^{t^{2}\sigma_{X}^{2}} \cdot t\right)}{\partial t} + t e^{t^{2}\sigma_{X}^{2}} \cdot \frac{\partial \phi(\sigma t)}{\partial t}\Big|_{t=0}$$

$$= \sigma \cdot \sqrt{2\pi} \left[\phi(0) + 0 + 0\right]$$

$$= \sigma \cdot \sqrt{2\pi} \cdot \frac{1}{2} = \frac{\sigma \cdot \sqrt{2\pi}}{2} \cdot \frac{1}{2} = \frac{\sigma \cdot \sqrt{2\pi}}$$

(c)
$$\frac{\partial w_{x}(t)}{\partial t} = 5\sqrt{2\pi} e^{\frac{t^{2}\sigma^{2}}{2}} \cdot \phi(t\sigma) + t^{2}\sigma^{2}e^{\frac{t^{2}\sigma^{2}}{2}} \phi(t\sigma) + te^{\frac{t^{2}\sigma^{2}}{2}} \cdot \sigma \cdot \frac{e^{-\frac{t^{2}\sigma^{2}}{2}}}{\sqrt{2\pi}}$$

$$= 6.\sqrt{2\pi} e^{\frac{t^{2}\sigma^{2}}{2}} \left[1 + t^{2}\sigma^{2}\right] \phi(t\sigma) + t\sigma^{2}$$

$$\frac{\partial w_{X}^{2}(t)}{\partial t^{2}} = 6.\sqrt{20} \cdot \left[t \cdot \sigma^{2} e^{\frac{t^{2}}{2}(4+t^{2}\sigma^{2})} \phi(t\sigma) + e^{\frac{t^{2}}{2}\sigma^{2}} \cdot 2t\sigma^{2} \phi(t\sigma) + e^{\frac{t^{2}}{2}\sigma^{2}} \cdot 2t\sigma^{2} \phi(t\sigma) + \sigma^{2}\right]$$

$$\frac{\partial w_X^2(t)}{\partial t^2}\Big|_{t=0} = 5.\sqrt{2\pi} \cdot \left[0+0+\frac{\partial \phi(t\sigma)}{\partial t}\Big|_{t=0}\right] + \sigma^2$$

$$V_{or}(X) = E[X^{2}] - (E[X])^{2} = 26^{2} - (5 \cdot \sqrt{\frac{1}{2}})^{2}$$

$$V_{or}(X) = (2 - \frac{\pi}{2}) \cdot 5^{2}$$

 $3|S_{Y} = [1,e], F_{Y}(y) = P(\exp(X) \le y) = P(X \le \log(y)) = \log(y) \text{ when } 1 \le y \le e.$

Then $f_Y(y) = (1/y) (u(y-1) - u(y-e))$. E(Y) = 1-e, $E(Y^2) = .5 (e^2 - 1)$, Var(Y) = .2420.

Using matlab, to generate pdf dt=.001; t=dt:dt:5; pdfy=(1<t & t<exp(1))./t;

CDF = cumsum(pdfy)*dt;

To generate random samples $y = \exp(rand(1,100000))$;

Sample mean is 1.7198 and sample variance is .2427

Sample pdf looks similar to real pdf, but curve is not as smooth. Sample CDF looks very similar to real CDF.

4) $S_Z = [0,1], F_Z(z) = P(\sin(\pi X/2) \le z) = P(X \le (2/\pi) \arcsin(z))$ when $0 \le z \le 1$.

Then
$$f_Z(z) = (2/\pi) (1-z^2)^{-.5} (u(z) - u(z-1))$$
. $E(Z) = 2/\pi$, $E(Z^2) = .5$, $Var(Z) = .0947$.

Matlab simulations similar to part a) except use different function. Sample mean is .6370 and sample variance is .0952.

Sample pdf looks similar to real pdf except near Z=1 as here Z goes to ∞. Sample CDF looks very similar to real CDF.

5) $S_W = [0,1]$, $F_W(w) = P((2X-1)2 \le w) = P(.5(1-w.5) \le X \le .5(1+w.5) = w.5$ when

 $0 \le w \le 1$. The $f_W(w) = (1/2) w^{-.5} u(w) - u(w-1)$. E(W) = 1/3, $E(W^2) = 1/5$, Var(W) = 4/45.

Matlab simulations similar to part a) except use different function. Sample mean is .3318 and sample variance is .0879. Sample pdf looks similar to real pdf except when W=0 as here W goes to ∞. Sample CDF looks very similar to real CDF.

