Homework 2 solution

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1. Let A be a Bernoulli random variable that shows that passenger is terrorist or not

Let B be another Bernoulli random variable that depicts the scanner result

$$P(B \mid A = 1) = 0.95$$
 for  $B = 1$  i.e.  $P(B = 1 \mid A = 1) = 0.95$   
0.05 for  $B = 0$  i.e.  $P(B = 0 \mid A = 1) = 0.05$ 

Similarly

$$P(B = 1 | A = 0) = 0.05$$

$$P(B = 0 | A = 0) = 0.95$$

So we want probability that passenger is terrorist given that scanner result is terrorist

so by Baye's rule

$$P(A = 1 \mid B = 1) = \frac{P(B = 1 \mid A = 1) * P(A = 1)}{(P(B = 1 \mid A = 1) * P(A = 1)) + (P(B = 1 \mid A = 0) * P(A = 0))}$$

$$= \frac{0.95 * 0.01}{(0.95 * 0.01) + (0.05 * 0.99)}$$

$$= \frac{19}{118}$$

$$= 0.161$$

A coin is tossed twice. So sample space will be S = {HH, HT, TH, TT}
For each H, Red ball(R) is placed or else white ball(W)
so various possibility of balls in the box will be {RR, RW,WR,WW}
each with a probability of ¼

$$P(RR) = P(RW) = P(WR) = P(WW) = \frac{1}{4}$$

The event of picking 3 red balls from the above possibilities are:

By Baye's rule, probability of picking from box of red balls P(RR | RRR) =

$$P(RRR \mid RR)*P(RR)$$

 $\overline{(P(RRR \mid RR)*P(RR))+(P(RRR \mid RW)*P(RW))+(P(RRR \mid WR)*P(WR))+(P(RRR \mid WW)*P(WW))}$ 

$$=\frac{1/4}{\frac{1}{4} + \left(\frac{1}{4} * \frac{1}{8}\right) + \left(\frac{1}{4} * \frac{1}{8}\right) + 0}$$

$$=\frac{4}{5}=0.8$$

We stop the coin toss as soon as we get Heads,
 So expected number of heads E(H) = 1

let X be number of coin flips required until head is received. E(X) be first flip, E(X|T) is number of remaining coin flips given tail is received in first flip and E(X|H) is number of remaining coin flips given head is received in first flip

$$E(X) = \frac{1}{2} * (1 + E(X|T)) + \frac{1}{2} * (1 + E(X|H))$$

E(X|H) = 0, as no need to flip after head is received in first flip and E(X|T) = E(X) as no progress was made towards getting head

$$E(X) = \frac{1}{2} * (1 + E(X)) + \frac{1}{2} * (1 + 0)$$
  
 $E(X) = 2$ 

Expected total flips = 2 and expected number of heads is 1 so the expected number of tails is 1.

4. 
$$P(x) = \frac{1}{b-a}$$
  $a \le x \le b$ 

0 elsewhere

Mean: E[X] = 
$$\int_{-\infty}^{+\infty} x p(x) dx$$
$$= \int_{a}^{b} \frac{x}{b-a} dx$$
$$= \frac{1}{b-a} \left[ \frac{x^2}{2} \right]_{a}^{b}$$
$$= \frac{1}{b-a} \left[ \frac{b^2}{2} - \frac{a^2}{2} \right]$$
$$= \frac{a+b}{2}$$

$$E[X^{2}] = \int_{-\infty}^{+\infty} x^{2} p(x) dx$$

$$= \int_{a}^{b} \frac{x^{2}}{b-a} dx$$

$$= \frac{1}{b-a} \left[ \frac{x^{3}}{3} \right]_{a}^{b}$$

$$= \frac{1}{b-a} \left[ \frac{b^{3}}{3} - \frac{a^{3}}{3} \right]$$

$$= \frac{a^{2} + ab + b^{2}}{3}$$

Variance : Var[X] = E[X<sup>2</sup>] – E[X]<sup>2</sup>  $= \frac{a^2 + ab + b^2}{3} - (\frac{a+b}{2})^2$   $= \frac{a^2 - 2ab + b^2}{12}$   $= \frac{(a-b)^2}{12}$ 

## 5. To prove:

$$E[X] = E_y [E_{x|y} [X]]$$

$$Var[X] = E_y [Var_{x|y} [X]] + Var_y [E_{x|y} [X]]$$

# 1<sup>st</sup> Proof:

# **RHS**

$$E_y[E_{x|y}[X]] = \int [\int x p(x|y) dx] p(y) dy$$

=
$$\iint x p(x|y) p(y) dx dy$$

=
$$\iint x p(x,y) dx dy$$

$$=\int x \int p(x,y) dy dx$$

$$= \int x p(x) dx$$

$$= E[X]$$

#### 2<sup>nd</sup> Proof

### <u>RHS</u>

$$\mathsf{E}_{\mathsf{y}}\left[\mathsf{Var}_{\mathsf{x}|\mathsf{y}}\left[\mathsf{X}\right]\right] + \mathsf{Var}_{\mathsf{y}}\left[\mathsf{E}_{\mathsf{x}|\mathsf{y}}\left[\mathsf{X}\right]\right]$$

$$=\mathsf{E}_{\mathsf{y}}\left[\mathsf{E}_{\mathsf{x}|\mathsf{y}}\left[\mathsf{X}^{2}\right]\right]\,-\mathsf{E}_{\mathsf{y}}\left[\left(\mathsf{E}_{\mathsf{x}|\mathsf{y}}\left[\mathsf{X}\right]\right)^{2}\right]+\mathsf{E}_{\mathsf{y}}\left[\left(\mathsf{E}_{\mathsf{x}|\mathsf{y}}\left[\mathsf{X}\right]\right)^{2}\right]-\mathsf{E}_{\mathsf{y}}\left(\left[\mathsf{E}_{\mathsf{x}|\mathsf{y}}\left[\mathsf{X}\right]\right)^{2}\right]$$

= 
$$E_y [E_{x|y} [X^2]] - E_y ([E_{x|y} [X])^2$$

= 
$$E_y [E_{x|y} [X^2]] - Ex^2 [X]$$
 -----using first tower property

= 
$$E_X[X^2]$$
 -  $E_{X}^2[X]$  -----using first tower property

6. 
$$P(\left|\frac{1}{n}\sum_{i=1}^{n}X_{i}-\sum[X_{i}]\right|>\varepsilon)$$

= P(
$$|\sum_{i=1}^{n} X_i - n\sum[X_i]| > n\varepsilon$$
)

we take 
$$\mbox{ Y} = \sum_{i=1}^n \mbox{ Xi and E[Y]} = \sum_{i=1}^n \mbox{ Xi } = \mbox{nE[X]}$$

= 
$$p(|Y - E[Y]| > n\varepsilon)$$

From Chebyshev inequality, we can write above eqn as

= p(|Y - E[Y]| > 
$$n\varepsilon$$
) <  $\frac{Var[Y]}{n^2\varepsilon^2}$  =  $\frac{Var[X]}{n\varepsilon^2}$  which tends to 0