(e)
$$P(DEN) = 0.25 \times 0.75 \times 0.25$$

= 0.046875

Problem 2.

Let E_i be the event that you pick iPhone i at the *i*th step. We need to find $P(\bigcup_{i=1}^n E_i)$. By the principle of inclusion-exclusion we have

$$P(\bigcup_{i=1}^{n} E_i) = \sum_{i=1}^{n} \sum_{i} P(E_i) - \sum_{i < j} P(E_i \cap E_j) + \sum_{i < j < k} P(E_i \cap E_j \cap E_k) - \dots + (-1)^{n+1} P(E_i \cap E_j \cap E_k)$$

Note that $P(E_i) = 1/n$ for all i. One way to see this is by using the full sample space: there are n! possible orderings of the iPhones, all equally likely, and (n-1)! of these are favorable to E_i .

Similarly

$$P(E_i \cap E_j) = \frac{(n-2)!}{n!} = \frac{1}{n(n-1)}$$

$$P(E_i \cap E_j \cap E_k) = \frac{(n-3)!}{n!} = \frac{1}{n(n-1)(n-2)}$$

and so on.

In the inclusion-exclusion formula, there are n terms involving one event, $\binom{n}{2}$ terms involving two events, $\binom{n}{3}$ terms involving three events, and so forth. By the symmetry of the problem, all n terms of the form $P(E_i)$ are equal, all $\binom{n}{2}$ terms of the form $P(E_i \cap E_j)$ are equal. Therefore one has

$$P(\bigcup_{i=1}^{n} E_i) = \frac{n}{n} - \frac{\binom{n}{2}}{n(n-1)} + \frac{\binom{n}{3}}{n(n-1)(n-2)} - \dots + (-1)^{n+1} \frac{1}{n!}$$
$$= 1 - \frac{1}{2!} + \frac{1}{3!} - \dots + (-1)^{n+1} \frac{1}{n!}$$

problem 3

(a)

Let A and R be the variables for "the owner have an above-average lifespan" and "the ring is the One Ring", respectively. Then,

 $P(A|R) = 0.95, P(\bar{A}|R) = 0.05, P(\bar{A}|\bar{R}) = 0.75 \text{ and } P(A|\bar{R}) = 0.25.$ Then,

$$\begin{split} P(R|A) &= \frac{P(A|R)P(R)}{P(A)} \\ &= \frac{P(A|R)P(R)}{P(A|R)P(R) + P(A|\bar{R})P(\bar{R})} \\ &= \frac{0.95 \cdot 10^{-4}}{0.95 \cdot 10^{-4} + 0.25 \cdot (1 - 10^{-4})} \\ &= \frac{0.95}{0.95 + 0.25 * 9999} \\ &= 0.00038 \end{split}$$

(b)

Now Let W be the variable for "writing will apear on it". Thus P(W|R) = 0.9, $P(\bar{W}|R) = 0.1$, $P(\bar{W}|\bar{R}) = 0.95$ and $P(W|\bar{R}) = 0.05$. Since A and W are conditionally independent tests, we have

$$P(R|WA) = \frac{P(WA|R)P(R)}{P(WA)} = \frac{P(WA|R)P(R)}{P(WA|R)P(R) + P(WA|\bar{R})P(\bar{R})}$$

Since A and W are conditionally independent, therefore,

$$P(R|WA) = \frac{P(W|R)P(A|R)P(R)}{P(WA|R)P(R) + P(WA|\bar{R})P(\bar{R})} = \frac{P(W|R)P(A|R)P(R)}{P(W|R)P(A|R)P(R) + P(W|\bar{R})P(A|\bar{R})P(\bar{R})}$$

Thus,

$$P(R|WA) = \frac{0.9 \cdot 0.95 \cdot 10^{-4}}{0.9 \cdot 0.95 \cdot 10^{-4} + 0.05 \cdot 0.25 \cdot (1 - 10^{-4})} \approx 0.006794$$

Way 1

First introduce a random variable I_x , where $I_x = \begin{cases} 1, & if \ X > x \\ 0, & otherwise \end{cases}$. Now assume $X = \sum_{x=0}^{\infty} I_x$. Then $E[X] = E[\sum_{x=0}^{\infty} I_x] = \sum_{x=0}^{\infty} E[I_x] = \sum_{x=0}^{\infty} P(X > x)$ QED.

Way 2

$$\sum_{x=0}^{\infty} Pr[X > x] = \sum_{x=0}^{\infty} \sum_{y=x+1}^{\infty} Pr[X = y]$$

$$= \sum_{x=0}^{\infty} \sum_{y=x+1}^{\infty} P_x(y)$$

$$= \sum_{y=1}^{\infty} \sum_{x=0}^{y-1} Pr[X = y]$$

$$= \sum_{y=1}^{\infty} P_x(y) \sum_{x=0}^{y-1} 1$$

$$= \sum_{y=1}^{\infty} P_x(y)y$$

$$= E[X]$$

(a)

$$\begin{split} E[I_E] &= 1.Pr(I_E = 1) + 0.Pr(I_E = 0) \\ &= Pr(I_E = 1) \\ &= Pr(E) \end{split}$$

(b)

$$\begin{split} Var(I_E) &= E[I_E^2] - E[I_E]^2 \\ &E[I_E^2] = 1^2.Pr(I_E = 1) + 0^2.Pr(I_E = 0) \\ &= pr(I_E = 1) \\ &= Pr(E) \\ Var(I_E) &= pr(E) - pr(E)^2 \\ &= Pr(E)(1 - Pr(E) \end{split}$$

(c)

$$\begin{split} E[X] &= \sum_{x=1}^{\infty} x (1-p)^{x-1}.p \\ &= p.[1+2.(1-p)+3.(1-p)^2+\ldots] \ Say \ eq(1) \\ &(1-p).E[X] = p.[0+1.(1-p)+2.(1-p)^2+3.(1-p)^3+\ldots] \ Say \ eq(2) \\ let's \ take \ eq(1) \ - \ eq(2) \end{split}$$

$$\begin{split} E[X] - (1-p).E[X] &= p.[1+(1-p)+(1-p)^2+\ldots]\\ p.E[X] &= p.[\frac{1}{1-(1-p)}]\\ E[X] &= \frac{1}{p} \end{split}$$

 $P_X = (1-p)^{x-1}.p$

To find the variance we need to find $E[x^2]$, we will start With E[X] and will find $E[X^2]$

$$E[X] = p \cdot \sum_{x=1}^{\infty} x(1-p)^{x-1}$$

$$(1-p) \cdot E[X] = p \cdot \sum_{x=1}^{\infty} x(1-p)^{x}$$

$$\frac{1-p}{p^{2}} = \sum_{x=1}^{\infty} x(1-p)^{x}$$

$$\frac{d}{dp} (\frac{1-p}{p^{2}}) = \frac{d}{dp} (\sum_{x=1}^{\infty} x(1-p)^{x})$$

$$\frac{d}{dp} (\frac{1-p}{p^{2}}) = \sum_{x=1}^{\infty} \frac{d}{dp} (x(1-p)^{x})$$

$$\frac{d}{dp} (\frac{1-p}{p^{2}}) = -1 \cdot \sum_{x=1}^{\infty} x^{2} (1-p)^{x-1}$$

$$\frac{-2}{p^{3}} + \frac{1}{p^{2}} = -1 \cdot \sum_{x=1}^{\infty} x^{2} (1-p)^{x-1}$$

$$\frac{2-p}{p^{3}} = \sum_{x=1}^{\infty} x^{2} (1-p)^{x-1}$$

$$p \cdot \frac{2-p}{p^{3}} = p \cdot \sum_{x=1}^{\infty} x^{2} (1-p)^{x-1}$$

$$\frac{2-p}{p^{2}} = E[X^{2}]$$

$$Var(X) = E[X^{2}] - E[X]^{2}$$

$$= \frac{2 - p}{p^{2}} - \frac{1}{p^{2}}$$

$$= \frac{1 - p}{p^{2}}$$

(a)

$$\begin{split} \sum_{i=0}^{\infty} P_X(i) &= \sum_{i=0}^{\infty} \frac{e^{-\lambda} \lambda^i}{i!} = e^{-\lambda} \sum_{i=0}^{\infty} \frac{\lambda^i}{i!} \\ &= e^{-\lambda} (1 + \frac{\lambda^1}{1!} + \frac{\lambda^2}{2!} + \frac{\lambda^3}{3!} + \ldots) \\ &= e^{-\lambda} e^{\lambda} \; (\text{From Taylor series expansion of } e^{\lambda}) \\ &= 1 \end{split}$$

(b)

$$\begin{split} E[X] &= \sum_{i=0}^{\infty} i P_X(i) = \sum_{i=0}^{\infty} i \frac{e^{-\lambda} \lambda^i}{i!} \\ &= \sum_{i=1}^{\infty} i \frac{e^{-\lambda} \lambda^i}{i!} = e^{-\lambda} \sum_{i=1}^{\infty} i \frac{\lambda^i}{i!} \\ &= \lambda e^{-\lambda} \sum_{i=1}^{\infty} \frac{\lambda^{i-1}}{(i-1)!} \\ &= \lambda e^{-\lambda} \sum_{i=0}^{\infty} \frac{\lambda^i}{i!} \\ &= \lambda e^{-\lambda} e^{\lambda} = \lambda \end{split}$$

(a)

$$\int_{1}^{+\infty} f_X(x) = \int_{1}^{+\infty} \alpha x^{-\alpha - 1} dx = -x^{-\alpha} |_{1}^{+\infty} = 0 + 1 = 1$$

(b)

$$E[x] = \int_1^{+\infty} x \cdot \alpha \cdot x^{-\alpha - 1} dx = \int_1^{+\infty} \alpha x^{-\alpha} dx = \frac{\alpha}{-\alpha + 1} x^{-\alpha + 1} \Big|_1^{+\infty} = \frac{\alpha}{\alpha - 1}$$

(c)

Similarly, we have

$$E[x^2] = \int_1^{+\infty} \alpha x^{-\alpha+1} dx = \frac{\alpha}{-\alpha+2} x^{-\alpha+2} |_1^{+\infty}$$

Since $1 < \alpha < 2$, then $0 < -\alpha + 2 < 1$. Thus $x^{-\alpha + 2} \to +\infty$, so $E[x^2] = +\infty$ and $Var[x] = +\infty$

Problem 8.

(a) Let F be the CDF y U.

Let Fx: R -> [0,1] be the CDF of X.

For any ZETR, we have

 $F_X(z) = P(X \le z) = P(F(u) \le z)$

 $= P(U \leq F(z)) = f(z)$

($since 0 \le F(z) \le 1$ for all z, $P(U \le F(z)) = F(z)$)

Since $F_X(z) = F(z) + z \in \mathbb{R}$, F is the CDF of X.

(b). Let Z be a random variable such that

Z= F(Y):

Since F(Y) takes values in (011),

 $P(Z \leq z) = P(F(Y) \leq z) = 0$ for all $z \leq 0$

& P(Z≤z) = P(F(Y) ≤z)=1 for all ZZ!

For OZZZI, we have

 $P(Z \le Z) = P(F(Y) \le Z) = P(Y \le F(Z))$

= F(F'(z)) = Z

Thus Z has Uniform (0,1) CDF.