

1b 30, c, v

Math 505a 2013 Spring Qualifying Exam

1. a) Let X and Y be square integrable random variables such that

$$E(X|Y) = Y \quad \text{and} \quad E(Y|X) = X. \quad (1)$$

Show that

$$P(X = Y) = 1. \quad (2)$$

- b) Prove that (1) implies (2) under the weakened assumption that X and Y are integrable.

2. Suppose k balls are tossed into n boxes, with all n^k possibilities equally likely. Let D be the number of boxes that contain exactly 2 balls.

- a) Compute $p := P(\text{exactly 2 balls land in box 1})$.
- b) In terms of p , give an exact expression for the mean ED .
- c) Compute $r := P(\text{exactly 2 balls land in box 1 and exactly 2 balls land in box 2})$.
- d) Give an exact expression for the second moment ED^2 in terms of p and r .
- e) Compute the variance of D .

3. a) Suppose $g(u) := Eu^S$ is the probability generating function of a non-negative integer valued random variable S satisfying $P(S > 0) > 0$. Let T be distributed as S , conditional on the event $S > 0$. Express $h(u) := Eu^T$, the probability generating function of T , in terms of $g(u)$.

In parts b) and c) below, N is a nonnegative integer valued random variable with probability generating function $f(u) := Eu^N$, and S is the number of heads in N tosses of a $p \in (0, 1)$ coin, with all coin tosses having probability p of coming up heads, independently of each other and of N .

$\mathbb{P}\{S > 0\} = 1 - \mathbb{P}\{S = 0\}$

$$E[u^T] = E[u^T | S > 0] P\{S > 0\} + E[u^T | S = 0] P\{S = 0\}$$

$$= g(u) P\{S > 0\} + \frac{E[u^T \mathbb{1}_{\{S=0\}}]}{P\{S=0\}} P\{S=0\}$$

$$= g(u) P\{S > 0\} + E[u^T \mathbb{1}_{\{S=0\}}]$$

*T integer valued?
nonnegative?*

b) Write the probability generating function $g(u) := Eu^S$ of S in a simple form.

c) Now combine parts a) and b): what is the probability generating function h of the number T of heads, in N tosses of a p -coin, conditional on getting at least one head, when N has probability generating function f ?

Parts d,e) can be worked on even if you are stumped by a,b,c).

d) Suppose someone claims that for $\alpha \in (0, 1)$, the function

$$f(u) := 1 - (1 - u)^\alpha$$

is a probability generating function of a nonnegative, non constant integer valued random variable N . What properties of f must you check? Is the hypothesis $\alpha > 0$ used? What happens in the cases $\alpha = 0$, $\alpha = 1$ and $\alpha > 1$?

e) Combine parts a)-d), that is suppose $\alpha \in (0, 1)$, N has the generating function $f(u) := 1 - (1 - u)^\alpha$, and T is the number of heads in N tosses of a p -coin, conditional on getting at least one head. Do N and T have the same distribution?

$$P(N=k) = \frac{f^{(k)}(0)}{k!}$$

$\Rightarrow$

$$P(N=0) = 0$$

$$P(N=1) = \alpha$$

$$P(N=k) = \frac{1}{k!} \prod_{i=0}^{k-1} (\alpha - i), \quad k \geq 1$$

- " $\alpha > 0$ " is required since $P(N=1) = \alpha$
- If $\alpha = 0$ $f(u)$ is not a prob. g. func.
- If $\alpha = 1$ $P(N=1) = 1$
- If $\alpha > 1$ $f(u)$ is not a prob. g. func. since $P(N=1) = \alpha$

1. a) $\rightarrow \underline{\text{Var}(X|Y)=0}$:

$$\begin{aligned}\text{Var}(X) &= \mathbb{E}[\text{Var}(X|Y)] + \text{Var}(\mathbb{E}[X|Y]) \\ &= \mathbb{E}[\text{Var}(X|Y)] + \text{Var}(Y) \\ &= \mathbb{E}[\text{Var}(X|Y)] + \mathbb{E}[\text{Var}(Y|X)] + \text{Var}(\mathbb{E}[Y|X]) \\ &= \mathbb{E}[\text{Var}(X|Y) + \text{Var}(Y|X)] + \text{Var}(X)\end{aligned}$$

Since $\text{Var}(X) < \infty$ (because of assumption) we get $\mathbb{E}[\underbrace{\text{Var}(X|Y) + \text{Var}(Y|X)}_{\geq 0}] = 0$
 $\Rightarrow \text{Var}(X|Y) + \text{Var}(Y|X) \geq 0 \Rightarrow \text{Var}(X|Y) = 0$ and $\text{Var}(Y|X) = 0$.

$\rightarrow \underline{\text{Var}(X) = \text{Var}(Y)}$:

$$0 = \text{Var}(X|Y) = \mathbb{E}[X^2|Y] - (\mathbb{E}[X|Y])^2 = \mathbb{E}[X^2|Y] - Y^2$$

So, we have $\mathbb{E}[X^2|Y] = Y^2$ and taking expectation of both sides, we obtain $\mathbb{E}[X^2] = \mathbb{E}[Y^2]$. This also gives us $\text{Var}(X) = \text{Var}(Y)$ since $\mathbb{E}[X] = \mathbb{E}[Y]$ by assumption ($\mathbb{E}[\mathbb{E}[X|Y]] = \mathbb{E}[Y]$).

$\rightarrow \underline{\mathbb{E}[XY] = \mathbb{E}[X^2]}$:

$$\mathbb{E}[XY] = \mathbb{E}[\mathbb{E}[XY|X]] = \mathbb{E}[X\mathbb{E}[Y|X]] = \mathbb{E}[X^2]$$

$\rightarrow \underline{P(X=Y)=1}$.

$$\begin{aligned}\text{Var}(X-Y) &= \text{Var}(X) + \text{Var}(Y) - 2\text{Cov}(X, Y) \\ &= 2\text{Var}(X) - 2\mathbb{E}[XY] + 2\mathbb{E}[X]\mathbb{E}[Y] \\ &= 2\mathbb{E}[X^2] - 2(\mathbb{E}[X])^2 - 2\mathbb{E}[X^2] + 2(\mathbb{E}[X])^2 \\ &= 0\end{aligned}$$

So, $P(X-Y=c)=1$ i.e. $X=Y+c$ a.e. Since $\mathbb{E}[X]=\mathbb{E}[Y]$, we must have $c=0$.
 Thus, $P(X=Y)=1$.

2. $D = \#$ of boxes that contain exactly 2 balls

a) $X = \#$ of balls in box 1, and $P(\text{any ball lands in box 1}) = \frac{1}{n}$. So,
 $X \sim \text{binomial}(k, \frac{1}{n})$ and $p = P(X=2) = \binom{k}{2} \left(\frac{1}{n}\right)^2 \left(\frac{n-1}{n}\right)^{k-2} = \binom{k}{2} \cdot \frac{(n-1)^{k-2}}{n^k}$.

b) $D = \sum_{i=1}^n 1_{D_i}$ where D_i is the event that i^{th} box contains exactly 2 balls. Then, $E[D] = \sum_{i=1}^n P(D_i)$ where $P(D_i) = \binom{k}{2} \frac{(n-1)^{k-2}}{n^k}$ by part (a).

So,

$$E[D] = \sum_{i=1}^n \binom{k}{2} \frac{(n-1)^{k-2}}{n^k} = \sum_{i=1}^n p = np$$

c) Firstly, let $X = \#$ of balls land in first box. So, $X \sim \text{Binomial}(k, \frac{1}{n})$, then letting $Y = \#$ of balls land in second box, we see that $Y|X \sim \text{Binomial}(k-X, \frac{1}{n-1})$. So,

$$\begin{aligned} P(X=2, Y=2) &= P(Y=2|X=2) P(X=2) \\ &= \binom{k-2}{2} \frac{1}{(n-1)^2} \left(\frac{n-2}{n-1}\right)^{k-4} \binom{k}{2} \frac{1}{n^2} \left(\frac{n-1}{n}\right)^{k-2} \\ &= \binom{k}{2} \binom{k-2}{2} \frac{(n-2)^{k-4}}{n^k} \quad \left(\begin{array}{l} \text{*choose 2 then choose 2*} \\ \frac{\text{all possible in } n-1}{\text{all possible in } n} \end{array} \right) \end{aligned}$$

$$\text{So, } r = \binom{k}{2} \binom{k-2}{2} \frac{(n-2)^{k-4}}{n^k}$$

$$\begin{aligned} d) E[D^2] &= \sum_{i=1}^n E[1_{D_i}^2] + \sum_{1 \leq i < j \leq n} E[1_{D_i} 1_{D_j}] \\ &= \sum_{i=1}^n P(D_i) + \sum_{1 \leq i < j \leq n} P(D_i \cap D_j) \\ &= \sum_{i=1}^n p + \sum_{1 \leq i < j \leq n} r \\ &= np + \frac{(n-1)n}{2} r \end{aligned}$$

$$e) \text{Var}(D) = E[D^2] - E[D]^2 = np + \frac{(n-1)n}{2} r - n^2 p^2 = np(1-np) + \frac{(n-1)n}{2} r$$

$$\begin{aligned} 3. a) E[u^T] &= E[u^T | S > 0] P\{S > 0\} + E[u^T | S = 0] P\{S = 0\} \\ &= g(u) P\{S > 0\} + \frac{E[u^T 1_{\{S=0\}}]}{P\{S=0\}} P\{S=0\} \\ &= g(u) P\{S > 0\} + E[u^T 1_{\{S=0\}}]. \end{aligned}$$

b) We have $S \sim \text{Binomial}(N, p)$. Then,

$$\begin{aligned} \mathbb{E}[u^S | N=n] &= \sum_{k=0}^n u^k P(S=k | N=n) \\ &= \sum_{k=0}^n u^k \binom{n}{k} p^k (1-p)^{n-k} \\ &= \sum_{k=0}^n \binom{n}{k} (pu)^k (1-p)^{n-k} \\ &= (1-p+pu)^n \end{aligned}$$

So, $\mathbb{E}[u^S | N] = (1-p+pu)^N$. Then,

$$\mathbb{E}[u^S] = \mathbb{E}[\mathbb{E}[u^S | N]] = \mathbb{E}[(1-p+pu)^N] = f(1-p+pu).$$

c) $T = \# \text{ of heads in } N \text{ tosses of a } p\text{-coin, conditional on getting at least one head}$

$$m(u) = \mathbb{E}[u^T | T > 0] = \frac{\mathbb{E}[u^T \mathbb{1}_{\{T > 0\}}]}{P\{T > 0\}} \quad \text{where } T \sim \text{Binomial}(N, p).$$

$$\begin{aligned} \rightarrow P\{T > 0\} &= \sum_{n=1}^{\infty} P\{T > 0 | N=n\} P\{N=n\} = \sum_{n=1}^{\infty} [1 - (1-p)^n] P\{N=n\} \\ &= \underbrace{\sum_{n=1}^{\infty} P\{N=n\}}_{\text{since } = 1 < \infty} - \underbrace{\sum_{n=1}^{\infty} (1-p)^n P\{N=n\}}_{< \infty} \\ &= 1 - \mathbb{E}[(1-p)^N] = 1 - f(1-p). \end{aligned}$$

$$\begin{aligned} \rightarrow \mathbb{E}[u^T \mathbb{1}_{\{T > 0\}}] &= \sum_{k=0}^{\infty} u^k \mathbb{1}_{\{T > 0\}} P\{T=k\} = \sum_{k=1}^{\infty} u^k P\{T=k\} = \sum_{k=0}^{\infty} u^k P\{T=k\} - P\{T=0\} \\ &= h(u) - P\{T=0\} = h(u) - 1 + P\{T > 0\} = h(u) - 1 + 1 - f(1-p) \\ &= h(u) - f(1-p) \end{aligned}$$

$$\text{Thus, } m(u) = \frac{h(u) - f(1-p)}{1 - f(1-p)}.$$

d) To have $f(u) = 1 - (1-u)^\alpha$ as a probability generating function of N , we must have

- i) $f(1) = 1$ since $f(u) = \mathbb{E}[u^N] = \sum_{k=0}^{\infty} u^k P\{N=k\}$ implies $f(1) = 1$. For the given function $f(u)$, $f(1) = 1 - (1-1)^\alpha = 1$, if $\alpha > 0$, this condition is satisfied. Note that if $\alpha = 0$, then $f(u) \equiv 1$ which is not a probability generating function. So, let $\alpha > 0$.
- ii) $P\{N=k\} = \frac{f^{(k)}(0)}{k!}$, i.e. f must be infinitely many times differentiable and $0 < \frac{f^{(k)}(0)}{k!} < 1$ and $\sum_{k=0}^{\infty} \frac{f^{(k)}(0)}{k!} = 1$.

The given function f is infinitely many times differentiable. Then, consider

$$f'(u) = \alpha(1-u)^{\alpha-1}$$

$$\rightarrow P\{N=1\} = f'(0) = \alpha$$

$$f''(u) = -\alpha(\alpha-1)(1-u)^{\alpha-2}$$

$$\rightarrow P\{N=2\} = \frac{\alpha(1-\alpha)}{2}$$

$$f'''(u) = \alpha(\alpha-1)(\alpha-2)(1-u)^{\alpha-3}$$

$$\rightarrow P\{N=3\} = \frac{\alpha(\alpha-1)(\alpha-2)}{6}$$

$$\vdots$$

$$f^{(k)}(u) = (-1)^{k-1} \prod_{i=1}^{k-1} (\alpha-i) (1-u)^{\alpha-k}$$

$$\rightarrow P\{N=k\} = \frac{(-1)^{k-1}}{k!} \prod_{i=1}^{k-1} (\alpha-i)$$

Since $P\{N=1\}$ we must have $\alpha \leq 1$

If $\alpha=1$, $P\{N=0\}=0$ and $P\{N=1\}=1$

So, we must have $0 < \alpha \leq 1$ to have $f(u)$ as a probability generating function

If $\alpha > 1$, $f(u)$ cannot be a probability generating function.