

Stat 2911 Lecture Notes

Class 8 , 2017

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Coupon collector problem (Rice 4.1), expectation of a product of ind RVs, L^2 is a vector subspace of L^1 corresponding to RVs with finite variance

Fubini's Theorem

- (a) A positive series can be summed up in any order.
- (b) An absolutely convergent series can be summed up in any order.

Let $g: \mathbb{R}^n \rightarrow \mathbb{R}$ and let $Z = g(\underline{x})$ where \underline{x} is an n -dimensional random vector.

Claim:

$$(i) Z \in L' \text{ iff } \sum_{x \in \mathbb{R}^n} |g(x)| P_{\underline{x}}(x) < \infty$$

$$(ii) \text{ If } Z \in L' \text{ then } E(Z) = \sum_{\substack{x \in \mathbb{R}^n: \\ P_{\underline{x}}(x) > 0}} g(x) P_{\underline{x}}(x)$$

(or $Z \geq 0$, or $Z \leq 0$)

Cor. If \underline{x} & y are jointly distributed RVs

with $\underline{x}, y \in L'$ and $\alpha, \beta \in \mathbb{R}$ then

$$(i) \alpha \underline{x} + \beta y \in L' \quad (L' \text{ is a vector space})$$

$$(ii) E(\alpha \underline{x} + \beta y) = \alpha E(\underline{x}) + \beta E(y)$$

(expectation is linear regardless of dependence)

Note. If we take $y = 1$ (constant RV) then

$$E(\alpha \underline{x} + \beta) = \alpha E(\underline{x}) + \beta$$

By induction, if $\underline{X}_1, \dots, \underline{X}_n \in \mathcal{L}'(\Omega)$ then for $\alpha_i \in \mathbb{R}$,

$$\sum_1^n \alpha_i \underline{X}_i \in \mathcal{L}' \text{ and } E\left(\sum_1^n \alpha_i \underline{X}_i\right) = \sum_1^n \alpha_i E(\underline{X}_i)$$

Examples

1) $\underline{X}_i \sim \text{Bernoulli}(p)$ RVs (not necessarily ind.)

Let $S_n = \sum_1^n \underline{X}_i$

Then,

$$\begin{aligned} E(S_n) &= E\left(\sum_1^n \underline{X}_i\right) \\ &= \sum_1^n E(\underline{X}_i) \\ &= \sum_1^n p \\ &= np \end{aligned}$$

In particular this shows that if $\gamma \sim \text{Binom}(n, p)$ then $E(\gamma) = np$.
 (The \underline{X}_i are ind. in this case)

2) Coupon Collector Problem

n types of coupons.

At each trial a single randomly selected coupon is collected.

What is the expected number of trials for completing a collection with all n types?

Let S'_n = number of trials for collecting all n types

X_1 = # of trials for accumulating 1 type

X_2 = # of additional trials for owning 2 types

X_3 = " " " 3 "

⋮

X_n = " " " n "

$$S'_n = \sum_{i=1}^n X_i \Rightarrow E(S'_n) = \sum_{i=1}^n E(X_i)$$

$$X_1 = 1 \Rightarrow E(X_1) = 1$$

$$X_2 \sim \text{geometric}\left(\frac{n-1}{n}\right) \Rightarrow E(X_2) = \frac{n}{n-1}$$

$$X_3 \sim \text{geometric}\left(\frac{n-2}{n}\right) \Rightarrow E(X_3) = \frac{n}{n-2}$$

$$\vdots \vdots \vdots \\ X_i \sim \text{geometric}\left(\frac{n-(i-1)}{n}\right) \Rightarrow E(X_i) = \frac{n}{n-i+1}$$

$$\Rightarrow E(S'_n) = \sum_{i=1}^n \frac{n}{n-i+1} = n\left(\frac{1}{n} + \frac{1}{n-1} + \dots + \frac{1}{2} + 1\right)$$

$\approx n \log n$ Aside: Are X_i independent RVs? Does it matter?
Much simpler than finding the pmf of S'_n

$E(\bar{X}+Y) = E(\bar{X}) + E(Y)$ but is $E(\bar{XY}) = E(\bar{X})E(Y)$?

Example. Let $\bar{X} = Y \sim \text{Bernoulli}(p)$, then

$$E(\bar{XY}) = E(\bar{X}^2) = p \neq p^2 = E(\bar{X})E(Y)$$

However, if \bar{X}, Y are ind $\text{Bernoulli}(p)$, then

$$E(\bar{XY}) = 1 \cdot p^2 = E(\bar{X})E(Y)$$

Is this the general rule?

Moreover, do we even know that $\bar{X} \cdot Y \in L'$?

Generally, $\bar{X}, Y \in L' \not\Rightarrow \bar{X} \cdot Y \in L'$ (later)

Claim. Suppose $\bar{X}, Y \in L'$ are ind then $\bar{X}Y \in L'$

and

$$E(\bar{XY}) = E(\bar{X})E(Y)$$

Proof. Let $g: \mathbb{R}^2 \rightarrow \mathbb{R}$ be $g(x, y) = xy$ and define $Z = g(\bar{X}, Y) = \bar{X}Y$.

$$\begin{aligned} \sum_{x,y} |g(x,y)| p_{xy}(x,y) &= \sum_{x,y} |xy| p_x(x) p_y(y) \\ &\stackrel{?}{=} \sum_x |x| p_x(x) \left(\sum_y |y| p_y(y) \right) \stackrel{?}{<} \infty \end{aligned}$$

$$\stackrel{?}{\Rightarrow} Z = \bar{X}Y \in L' \text{ and } E(Z) \stackrel{?}{=} E(\bar{X})E(Y)$$

Repeat the above argument w/o the $| \cdot |$ and use Fubini (ii)

Recall:

$$\begin{aligned} L' &= \left\{ \underline{X} : \sum_i |x_k| p_x(x_k) < \infty \right\} & [\underline{X} \in L' \Leftrightarrow E|\underline{X}| < \infty] \\ L^2 &= \left\{ \underline{X} : \sum_i x_k^2 p_x(x_k) < \infty \right\} & \begin{aligned} [\underline{X} \in L^2 \Leftrightarrow E\underline{X}^2 < \infty] \\ \Leftrightarrow \underline{X} \in L' \end{aligned} \end{aligned}$$

What is the relation between L' and L^2 ?

Let \underline{X} be a RV with pmf $p_x(n) = C/n^3 \cdot 1_{n \in \mathbb{N}}$, where $C = 1/\sum_1^\infty n^{-3}$.

Is $\underline{X} \in L'$?

$$\sum_1^\infty n p_x(n) = \sum_1^\infty C/n^2 < \infty \Rightarrow \underline{X} \in L'$$

Is $\underline{X} \in L^2$?

$$\sum_1^\infty n^2 p_x(n) = \sum_1^\infty C/n > \infty \Rightarrow \underline{X} \notin L^2$$

An example of $\underline{X} \in L' \setminus L^2$ or $\underline{X} \in L' \not\Rightarrow \underline{X} \in L^2$.

What about the converse?

Claim. $\underline{X} \in L^2 \Rightarrow \underline{X} \in L'$

Proof. $|x| < (1+x^2)^{1/2} \quad \forall x \in \mathbb{R}$ (exercise)

$$\begin{aligned} \Rightarrow \sum_k |x_k| p_x(x_k) &< \sum_k (1+x_k^2)^{1/2} p_x(x_k) \\ &= \sum_k p_x(x_k) + \sum_k x_k^2 p_x(x_k) < \infty \end{aligned}$$

Recall: $\underline{X} \in L^1$ iff $E(\underline{X})$ exists and is finite, we now have

Cor. $\underline{X} \in L^2$ iff $V(\underline{X})$ exists and is finite

Proof. $\underline{X} \in L^2 \Rightarrow \underline{X} \in L^1 \Rightarrow E(\underline{X}) \in \mathbb{R}$
 $\Rightarrow V(\underline{X}) = E(\underline{X}^2) - E^2(\underline{X}) \in \mathbb{R}$

Conversely, we showed that, if $V(\underline{X})$ exists and is finite then $E(\underline{X}^2) < \infty$.

Suppose X and y are jointly distributed RVs with finite variance.

Want $V(X+y)$ but is $\underline{X+y} \in L^2$?

The answer is positive but to see that we first need to verify that $\underline{X+y} \in L^1$.

Aside. Recall we said $\underline{X}, \underline{y} \in L^1 \nrightarrow \underline{X+y} \in L^1$
 Can you think of an example now?

Take $\underline{X} = \underline{y}$ the same RV with $p_x = \frac{1}{n^3} \mathbf{1}_{n \in \mathbb{N}}$,
 We saw that $\underline{X} \in L^1$ but $\underline{X} \notin L^2$, so

$$\underline{X+y} = \underline{X}^2 \notin L^1$$

Claim. If $\underline{x}, \underline{y} \in L^2$ then $\underline{x}\underline{y} \in L'$

Proof. For $x, y \in \mathbb{R}$: $0 \leq (|x| - |y|)^2 = x^2 - 2|x||y| + y^2$

$$\Rightarrow |xy| \leq x^2/2 + y^2/2 = \frac{x^2 + y^2}{2}$$

Again, using $Z = g(\underline{x}, \underline{y}) = \underline{x} \cdot \underline{y}$

geometric mean of x^2 and y^2 is \leq arithmetic mean

$$\Rightarrow \sum_{x,y} |xy| P_{xy}(x,y) \leq \sum_{x,y} \left(x^2/2 + y^2/2 \right) P_{xy}(x,y)$$

$$= \sum_x x^2/2 \sum_y P_{xy}(x,y) + \sum_y y^2/2 \sum_x P_{xy}(x,y)$$

$$= \frac{1}{2} \sum_x x^2 P_x(x) + \frac{1}{2} \sum_y y^2 P_y(y) < \infty$$

Q
⇒ $\underline{x}\underline{y} \in L'$

□

Cor. If $\underline{x}, \underline{y} \in L^2$ then $\forall \alpha, \beta \in \mathbb{R}: \alpha\underline{x} + \beta\underline{y} \in L^2$

L^2 is a vector space (subspace of L')

Proof. $(\alpha\underline{x} + \beta\underline{y})^2 = \alpha^2\underline{x}^2 + 2\alpha\beta\underline{x}\underline{y} + \beta^2\underline{y}^2$

$\underline{x}^2, \underline{y}^2, \underline{x}\underline{y} \in L'$ and the RHS is a linear comb. of these three L' -RVs, therefore it is in L' itself.

$$\Rightarrow \alpha\underline{x} + \beta\underline{y} \in L^2$$

We will soon find $V(\bar{X}+Y)$ but first consider the following special case.

Claim. If $V(\underline{X}) = \sigma^2 < \infty$, then $\forall \alpha, \beta \in \mathbb{R}$

$$V(\alpha \underline{X} + \beta) = \alpha^2 \sigma^2 \quad (\text{variance is shift-invariant and it scales quadratically})$$

↑ scale ↑ shift

Proof.

Let $\mu = E(\underline{X}) \in \mathbb{R}$ then

$$E(\alpha \underline{X} + \beta) = \alpha \mu + \beta$$

$$\begin{aligned} \Rightarrow V(\alpha \underline{X} + \beta) &= E[(\alpha \underline{X} + \beta) - (\alpha \mu + \beta)]^2 \\ &= E[\alpha(\underline{X} - \mu)]^2 \\ &= \alpha^2 E(\underline{X} - \mu)^2 \\ &= \alpha^2 V(\underline{X}) \end{aligned}$$

□