

Probability Theory

1st. Test

1st. Semester — 2010/11

Duration: 1h30m

2010/11/13 — 9AM, Room P8

- Please justify your answers.
- This test has two pages and four groups. The total of points is 20.0.

Group I — Warm up and probability spaces

3.0 points

During World War II, random walk was used to model the distance that an escaped prisoner of war would travel in a given time.

A military expert assumes that the position of a prisoner at time n ($n = 1, 2, \dots$), X_n , is governed by an asymmetric random walk — starting at “0” (the prison camp) and with probability of an northward (resp. southward) “step” (of one km in one time unit) equal to p (resp. $1 - p$), where $p \in (0, \frac{1}{2})$.

Let A_k be the event that the prisoner ever reaches point k ($k = 0, 1, 2, \dots$) north of the prison camp.

(a) Argue that $A_k \supseteq A_{k+1}$, $k = 0, 1, 2, \dots$, i.e., $A_k \downarrow$. (1.0)

- **Process**

Asymmetric random walk

- **R.v.**

Y_n = size of the n^{th} step

$Y_n \stackrel{i.i.d.}{\sim} Y$, $n \in \mathbb{N}$

$$P(Y = y) = \begin{cases} p, & y = 1 \text{ (step of 1 km north of the prison camp)} \\ 1 - p, & y = -1 \text{ (step of 1 km south of the prison camp)} \\ 0, & \text{otherwise} \end{cases}$$

$X_n = \sum_{i=1}^n Y_i$ = position of the prisoner after n steps ($n \in \mathbb{N}$)

- **Initial condition**

$X_0 = 0$ (prison camp)

- **Event**

A_k = prisoner ever reaches point k north of the prison camp, $k \in \mathbb{N}_0$

- **To argue**

$A_k \downarrow$, i.e., $\{A_k, k \in \mathbb{N}_0\}$ is a non-increasing sequence of events

- **Arguing that $A_k \downarrow$**

Note that:

- (i) $A_0 = \Omega$ because the prisoner starts at the prison camp — $X_0 = 0$ — and therefore $A_0 \supseteq A_1$ is obviously true;

- (ii) exploiting the “continuity” of the RW lead us to conclude that to reach position $k + 1$ the prisoner has to pass through position k , thus the occurrence of A_{k+1} implies the one of A_k , i.e., $A_k \supseteq A_{k+1}$ is true for $k \in \mathbb{N}$.

(b) Derive $P(A_r)$, for $r = 1, 2, \dots$, by using the fact that $A_k \downarrow$, $P(A_0) = 1$ and $P(A_{k+1}|A_k) = \frac{p}{1-p}$, $k = 0, 1, 2, \dots$ (2.0)

- **Requested probability**

$$\begin{aligned} P(A_r) &\stackrel{A_k \downarrow}{=} P(\cap_{k=0}^r A_k) \\ &\stackrel{\text{mult. rule}}{=} P(A_0) \times P(A_1 | A_0) \times P(A_2 | A_0 \cap A_1) \times \dots \\ &\quad \times P(A_r | A_0 \cap A_1 \cap \dots \cap A_{r-1}) \\ &\stackrel{A_k \downarrow}{=} P(A_0) \times P(A_1 | A_0) \times P(A_2 | A_1) \times \dots \times P(A_r | A_{r-1}) \\ &= 1 \times \prod_{k=0}^{r-1} P(A_{k+1}|A_k) \\ &\stackrel{P(A_{k+1}|A_k)=\frac{p}{1-p}}{=} \prod_{k=0}^{r-1} \left(\frac{p}{1-p} \right) \\ &= \left(\frac{p}{1-p} \right)^r, \quad r \in \mathbb{N}. \end{aligned}$$

Group II — Probability spaces and random variables

5.5 points

1. Consider the sample space $\Omega = \{A, B, C\}$, and let $\mathcal{F} = \{\emptyset, \{A\}, \{B, C\}, \Omega\}$ and $\mathcal{H} = \{\emptyset, \{A, B\}, \{C\}, \Omega\}$ be two σ – algebras on Ω .

Show that $\mathcal{F} \cup \mathcal{H}$, the collection of subsets of Ω lying in either \mathcal{F} or \mathcal{H} , is not a σ – algebra. (1.5)

- **Sample space**

$\Omega = \{A, B, C\}$

- **σ – algebras on Ω**

$\mathcal{F} = \{\emptyset, \{A\}, \{B, C\}, \Omega\}$

$\mathcal{H} = \{\emptyset, \{A, B\}, \{C\}, \Omega\}$

- **To prove**

$\mathcal{F} \cup \mathcal{H}$ is not a σ – algebra

- **Proof**

We ought to mention that, according to Def. 1.38, a minimal set of postulates for a non-empty class of subsets \mathcal{A} of Ω to be a σ – algebra on Ω is:

- (i) $\Omega \in \mathcal{A}$;
 (ii) $A \in \mathcal{A} \Rightarrow A^c \in \mathcal{A}$;
 (iii) $A_1, A_2, \dots \in \mathcal{A} \Rightarrow \cup_{i=1}^{+\infty} A_i \in \mathcal{A}$.

Hence, we have to prove that at least one of these 3 postulates are not true for $\mathcal{A} = \mathcal{F} \cup \mathcal{H}$.

Firstly, note that

$$\mathcal{A} = \mathcal{F} \cup \mathcal{H} = \{\emptyset, \{A\}, \{C\}, \{A, B\}, \{B, C\}, \Omega\}.$$

Thus, $\Omega \in \mathcal{A} = \mathcal{F} \cup \mathcal{H}$.

Secondly,

D	$D^c = \Omega \setminus D$	$D^c \in \mathcal{A} = \mathcal{F} \cup \mathcal{H} ?$
\emptyset	Ω	TRUE
$\{A\}$	$\{B, C\}$	TRUE
$\{A, B\}$	$\{C\}$	TRUE
$\{B, C\}$	$\{A\}$	TRUE
$\{C\}$	$\{A, B\}$	TRUE
Ω	\emptyset	TRUE

Thirdly, the following disjoint events

$$\begin{aligned} D_1 &= \{A\} \\ D_2 &= \{C\} \\ D_3 &= D_4 = \dots = \emptyset \end{aligned}$$

yield

$$\begin{aligned} \bigcup_{i=1}^{+\infty} D_i &= \{A, C\} \\ &\notin \mathcal{A} = \mathcal{F} \cup \mathcal{H}. \end{aligned}$$

As a consequence, $\mathcal{A} = \mathcal{F} \cup \mathcal{H}$ is not a σ -algebra.

2. Suppose that P_1 and P_2 are probability functions on (Ω, \mathcal{F}) and that $\alpha \in [0, 1]$. Prove that (2.0) the set function

$$P(A) = \alpha \times P_1(A) + (1 - \alpha) \times P_2(A), \quad A \in \mathcal{F},$$

is also a probability function on (Ω, \mathcal{F}) .

• **Setting**

Consider two p.f. on (Ω, \mathcal{F}) :

$$P_j : \Omega \rightarrow [0, 1], \quad j = 1, 2.$$

• **To prove**

$P(A) = \alpha \times P_1(A) + (1 - \alpha) \times P_2(A)$, $A \in \mathcal{F}$, where $\alpha \in [0, 1]$, is also a p.f. on (Ω, \mathcal{F}) .

• **Proof**

According to Def. 1.48, the two p.f. defined of (Ω, \mathcal{F}) , P_j , $j = 1, 2$, satisfy

- (i) Axiom 1 — $P_j(A) \geq 0$, $\forall A \in \mathcal{F}$

- (ii) Axiom 2 — $P_j(\Omega) = 1$

- (iii) Axiom 3 (countable additivity) — Whenever A_1, A_2, \dots are (pairwise) disjoint events in \mathcal{F} , $P_j(\bigcup_{i=1}^{+\infty} A_i) = \sum_{i=1}^{+\infty} P_j(A_i)$,

for $j = 1, 2$.

As a result, the set function P verifies:

$$\begin{aligned} P(A) &= \alpha \times P_1(A) + (1 - \alpha) \times P_2(A) \\ &\geq 0, \text{ for } A \in \mathcal{F} \text{ (by (i))}; \end{aligned}$$

$$\begin{aligned} P(\Omega) &= \alpha \times P_1(\Omega) + (1 - \alpha) \times P_2(\Omega) \\ &= \alpha \times 1 + (1 - \alpha) \times 1 \text{ (by (ii))} \\ &= 1; \end{aligned}$$

$$\begin{aligned} P\left(\bigcup_{i=1}^{+\infty} A_i\right) &= \alpha \times P_1\left(\bigcup_{i=1}^{+\infty} A_i\right) + (1 - \alpha) \times P_2\left(\bigcup_{i=1}^{+\infty} A_i\right) \\ &= \alpha \times \sum_{i=1}^{+\infty} P_1(A_i) + (1 - \alpha) \times \sum_{i=1}^{+\infty} P_2(A_i) \text{ (by (iii))} \\ &= \sum_{i=1}^{+\infty} [\alpha \times P_1(A_i) + (1 - \alpha) \times P_2(A_i)] \\ &= \sum_{i=1}^{+\infty} P(A_i), \text{ for any } A_1, A_2, \dots \text{ (pairwise) disjoint events in } \mathcal{F}. \end{aligned}$$

That is, P is indeed a p.f. on (Ω, \mathcal{F}) .

3. A parent particle can split into 0, 1 or 2 particles with probabilities $\frac{1}{4}, \frac{1}{2}, \frac{1}{4}$, respectively. (2.0)

It disappears after splitting. Beginning with one particle (the progenitor) let us denote by X_i the number of particles in generation i ($i = 0, 1, 2, \dots$) and remind the reader that $P(\{X_0 = 1\}) = 1$.

Find $P(\{X_2 > 0\})$.

• **R.v.**

$X_0 = 1$ (we began with one particle, the progenitor)

X_1 = number of particles in which the progenitor splits into

⋮

X_i = number of particles in generation i , $i \in \mathbb{N}$

• **P.f. of X_1**

$$P(X_1 = x) = \begin{cases} \frac{1}{4}, & x = 0, 2 \\ \frac{1}{2}, & x = 1 \\ 0, & \text{otherwise} \end{cases}$$

- **Other r.v.**

$$Y_1^{(0)} = X_1$$

$Y_j^{(i)}$ = number of particles in which particle j splits into in generation i , $i, j \in \mathbb{N}$

$\{Y_j^{(i)}, j \in \mathbb{N}\} \stackrel{i.i.d.}{\sim} Y_1^{(0)}$, for all $i \in \mathbb{N}$

- **Obs.**

$$X_1 = Y_1^{(0)}$$

$$X_2 = \sum_{j=1}^{X_1} Y_j^{(1)}$$

⋮

$$X_{i+1} = \sum_{j=1}^{X_i} Y_j^{(i)}, i \in \mathbb{N}_0$$

- **Requested probability**

$$\begin{aligned} P(X_2 > 0) &= 1 - P(X_2 = 0) \\ &\stackrel{\text{total prob. law}}{=} 1 - \sum_{i=0}^2 P(X_2 = 0 \mid X_1 = i) \times P(X_1 = i) \\ &= 1 - [P(X_2 = 0 \mid X_1 = 0) \times P(X_1 = 0) \\ &\quad + P(X_2 = 0 \mid X_1 = 1) \times P(X_1 = 1) \\ &\quad + P(X_2 = 0 \mid X_1 = 2) \times P(X_1 = 2)] \\ &= 1 - \left[1 \times \frac{1}{4} + \frac{1}{4} \times \frac{1}{2} + \left(\frac{1}{4}\right)^2 \times \frac{1}{4} \right] \\ &= 1 - \frac{16 + 8 + 1}{64} \\ &= \frac{39}{64}. \end{aligned}$$

Group III — Random variables

5.0 points

Consider a cubical fuel tank with capacity equal to 1000 liters. For safety reasons, the tank is kept three-quarters full (750 liters). Moreover, the only tank wall subjected to corrosion is the front one; and if a hole appears in that wall at height X (in meters), all the fuel above the hole leaks out of the tank.

Let Y be the amount of fuel (in liters) that will remain in the tank after a hole appears and assume that $X \sim \text{Uniform}(0, 1)$.

(a) Define Y in terms of X and show that Y is also a r.v. (2.0)

- **R.v.**

X = height (in meters) at which the hole appears in the front wall of the fuel tank

- **Distribution**

$X \sim \text{Uniform}(0, 1)$

- **Transform**

$$\begin{aligned} Y &= \text{amount of fuel (in liters) that will remain in the tank after leakage} \\ &= \begin{cases} 1000X, & \text{if } 0 \leq X \leq 0.75 \\ 750, & \text{if } 0.75 < X \leq 1 \end{cases} \\ &= \min\{1000X, 750\} \\ &= g(X) \end{aligned}$$

- **Important**

Let:

- X be a real r.v.;
- (Ω, \mathcal{F}) and $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ be two measurable spaces.

Then, by Def. 2.13, $X : \Omega \rightarrow \mathbb{R}$ and is such that

$$X^{-1}(B) = \{\omega \in \Omega : X(\omega) \in B\} \in \mathcal{F}, \forall B \in \mathcal{B}(\mathbb{R}), \quad (1)$$

namely,

$$X^{-1}((-\infty, x]) = \{\omega \in \Omega : X(\omega) \leq x\} \in \mathcal{F}, \forall x \in \mathbb{R},$$

according to Prop. 2.16.

- **To prove**

$Y = g(X) = \min\{1000X, 750\}$ is a Borel measurable function, thus, also a r.v., by Corollary 2.40.

- **Proof**

Firstly, let us remind the reader that a function $g : \mathbb{R} \rightarrow \mathbb{R}$ is Borel measurable iff

$$g^{-1}(B) \in \mathcal{B}(\mathbb{R}), \forall B \in \mathcal{B}(\mathbb{R}),$$

according to Def. 2.36, with $n = m = 1$. Furthermore, in order that $g : \mathbb{R} \rightarrow \mathbb{R}$ be Borel measurable it suffices that $g^{-1}((-\infty, y]) \in \mathcal{B}(\mathbb{R}), \forall y \in \mathbb{R}$, according to Remark 2.47 (with $n = 1$).

Secondly, note that the range of X is

$$\mathbb{R}_X = [0, 1],$$

therefore the range of Y is given by

$$\begin{aligned} \mathbb{R}_Y &= g(\mathbb{R}_X) \\ &= [0, 750]. \end{aligned}$$

Thirdly, $g^{-1}((-\infty, y])$ equals:

- for $y < 0$,

$$\begin{aligned} \{x \in \mathbb{R} : g(x) \leq y\} &= \emptyset \\ &\in \mathcal{B}(\mathbb{R}); \end{aligned}$$

◦ for $0 \leq y \leq 750$,

$$\begin{aligned} \{x \in \mathbb{R} : g(x) \leq y\} &= \{x \in \mathbb{R} : 1000x \leq y\} \\ &= (-\infty, y/1000] \\ &\in \mathcal{B}(\mathbb{R}); \end{aligned}$$

◦ for $y > 750$,

$$\begin{aligned} \{x \in \mathbb{R} : g(x) \leq y\} &= \Omega \\ &\in \mathcal{B}(\mathbb{R}). \end{aligned}$$

Consequently, we can conclude that g is indeed a Borel measurable function and therefore Y is a r.v., by Corollary 2.40. QED

(b) Derive and draw the graph of the d.f. of Y and calculate $P(\{Y = 750\})$. (3.0)

• **D.f. of X**

$$F_X(x) = \begin{cases} 0, & x < 0 \\ x, & 0 \leq x \leq 1 \\ 1, & x > 1 \end{cases}$$

• **D.f. of Y**

$$\begin{aligned} F_Y(y) &= P(Y \leq y) \\ &= P(\min\{1000X, 750\} \leq y) \\ &= \begin{cases} P(\emptyset) = 0, & y < 0 \\ P(1000X \leq y) = F_X\left(\frac{y}{1000}\right) = \frac{y}{1000}, & 0 \leq y < 750 \\ P(\Omega) = 1, & y \geq 750 \end{cases} \end{aligned}$$

• **Graph of $F_Y(y)$**

(It has a discontinuity at $y = 750$, etc.)

• **Requested probability**

Since $F_Y(y)$ has a discontinuity point at $y = 750$,

$$\begin{aligned} P(Y = 750) &= F_Y(750) - F_Y(750^-) \\ &= 1 - 0.75 \\ &= 0.25. \end{aligned}$$

This probability coincides with $P(0.75 < X \leq 1) = P(\text{hole above the mark of } 0.75\text{m})$.

Group IV — Independence and random variables

6.5 points

1. If A, B, C are independent events, show directly that $A \cup B$ is independent of C . (1.5)

• **Setting**

A, B, C are independent events, i.e.,

$$\begin{aligned} P(A \cap B) &= P(A) \times P(B) \\ P(A \cap C) &= P(A) \times P(C) \\ P(B \cap C) &= P(B) \times P(C) \\ P(A \cap B \cap C) &= P(A) \times P(B) \times P(C) \end{aligned}$$

• **To prove**

$A \cup B \perp\!\!\!\perp C$, that is, $P[(A \cup B) \cap C] = P(A \cup B) \times P(C)$.

• **Proof**

Invoking the independence, namely the *factorisations above*, we get:

$$\begin{aligned} P[(A \cup B) \cap C] &= P[(A \cap C) \cup (B \cap C)] \\ &= P[A \cap C] + P[B \cap C] - P[(A \cap C) \cap (B \cap C)] \\ &= P[A \cap C] + P[B \cap C] - P(A \cap B \cap C) \\ &= P(A) \times P(C) + P(B) \times P(C) - P(A) \times P(B) \times P(C) \\ &= [P(A) + P(B) - P(A) \times P(B)] \times P(C) \\ &= P(A \cup B) \times P(C). \end{aligned}$$

QED

2. The maximum flying time, V (in hours), of a small aircraft is defined by $V = 30XY$, where: the r.v. X represents the amount of fuel (in tons) in the beginning of the flight; and the r.v. Y represents the reciprocal of the fuel consumption rate (in hours per ton of fuel).

Admit that X and Y are independent r.v. such that $X \sim \text{Uniform}(0, 1)$ and $Y \sim \text{Beta}(2, 1)$.

(a) Obtain the survival function of V . (3.0)

• **R.v.**

X = amount of fuel (in tons) in the beginning of the flight
 Y = reciprocal of the fuel consumption rate (in hours per ton of fuel)

• **Distributions**

$X \sim \text{Uniform}(0, 1)$
 $\perp\!\!\!\perp$
 $Y \sim \text{Beta}(2, 1)$

• **Transformation**

V = maximum flying time (in hours)

- **P.d.f. de X**

$$f_X(x) = \begin{cases} 1, & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

- **P.d.f. de Y**

$$f_Y(y) = \begin{cases} \frac{\Gamma(2+1)}{\Gamma(2)\Gamma(1)} y^{2-1} (1-y)^{1-1} = 2y, & 0 < y < 1 \\ 0, & \text{otherwise} \end{cases}$$

- **Survival function of Y**

$$\begin{aligned} \bar{F}_Y(y) &= 1 - F_Y(y) \\ &= \begin{cases} 1, & y \leq 0 \\ 1 - \int_0^y 2t \, dt = 1 - y^2, & 0 < y < 1 \\ 0, & y \geq 1. \end{cases} \end{aligned}$$

- **Range of V**

$$\mathcal{R}_V = [0, 30]$$

- **Survival function of V**

For $0 \leq v \leq 30$,

$$\begin{aligned} \bar{F}_V(v) &= P(V > v) \\ &= P(30XY > v) \\ &= P\left(Y > \frac{v}{30X}\right) \\ &\stackrel{X \perp\!\!\!\perp Y}{=} \int \int_{\{(x,y) \in [0,1]^2: y > \frac{v}{30x}\}} f_X(x) \times f_Y(y) \, dy \, dx \quad (\text{schematics!}) \\ &= \int_{\frac{v}{30}}^1 f_X(x) \times \left[\int_{\frac{v}{30x}}^1 f_Y(y) \, dy \right] \, dx \\ &= \int_{\frac{v}{30}}^1 f_X(x) \times \bar{F}_Y\left(\frac{v}{30x}\right) \, dx \\ &= \int_{\frac{v}{30}}^1 1 \times \left[1 - \left(\frac{v}{30x}\right)^2 \right] \, dx \\ &= \int_{\frac{v}{30}}^1 dx - \left(\frac{v}{30}\right)^2 \times \int_{\frac{v}{30}}^1 x^{-2} \, dx \\ &= \left(1 - \frac{v}{30}\right) + \left(\frac{v}{30}\right)^2 \times \frac{1}{x} \Big|_{\frac{v}{30}}^1 \\ &= \left(1 - \frac{v}{30}\right)^2. \end{aligned}$$

(b) Describe a method to generate (pseudo-)random numbers from the distribution of V . (2.0)

- **Quantile function**

Let $u \in (0, 1)$ then the quantile function is given by:

$$\begin{aligned} F_V(v) &= u \\ 1 - \bar{F}_V(v) &= u \\ 1 - \left(1 - \frac{v}{30}\right)^2 &= u \end{aligned}$$

$$\begin{aligned} F_V^{-1}(u) &= v \\ &= 30(1 - \sqrt{1-u}). \end{aligned}$$

- **Quantile transformation**

By Prop. 2.140, if

$$U \sim \text{Uniform}(0, 1)$$

then $F_V^{-1}(U) \stackrel{d}{=} V$, i.e.,

$$30(1 - \sqrt{1-U}) \stackrel{d}{=} V.$$

Consequently, if we want to generate (pseudo-)random numbers from the distribution of V then we have to:

- (1) generate u from a $\text{Uniform}(0, 1)$ distribution;
- (2) assign $v = 30(1 - \sqrt{1-u})$.