

Probability Theory

1st. TEST
Duration: 1h30m

1st. Semester — 2010/11
2011/01/31 — 3PM, Room P2

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

Group I — Warm up

3.0 points

Suppose that in an election, candidate A receives a votes and candidate B receives b votes where $a \in \mathbb{N}$, $b \in \mathbb{N}_0$, and $a > b$.

Assuming a random ordering of the votes, what is the probability that A is always ahead of B in the vote count?

This is an historically famous problem known as the “Ballot Problem”, that was solved by Joseph Louis Bertrand in 1887.

(a) Show that the ballot problem is intimately related to asymmetric random walks. (1.0)

- **Process**

$$\{X_n, n = 0, 1, \dots, a + b\}$$

$X_0 = 0$ (candidate A is zero votes ahead of B before the counting of the votes starts)

X_n = difference of votes between the candidates A and B at the count of the n^{th} vote

- **Relating the “Ballot Problem” with a random walk**

Y_i = size of the i^{th} step

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

$$P(Y = y) = \begin{cases} \frac{a}{a+b}, & y = 1 \text{ (if candidate } A \text{ gets the vote we just counted)} \\ \frac{b}{a+b}, & y = -1 \text{ (if candidate } B \text{ gets the vote we just counted)} \\ 0, & \text{otherwise} \end{cases}$$

$$X_n = \sum_{i=1}^n Y_i, n \in \mathbb{N}$$

- **Conclusion**

Since $a \in \mathbb{N}$, $b \in \mathbb{N}_0$ and $a > b$, $\{X_n, n = 0, 1, \dots, a + b\}$ is indeed an asymmetrical random walk.

(b) Let $f(a, b)$ denote the probability that A is always ahead of B in the vote count. Condition on the candidate that receives the last vote to prove that (2.0)

$$f(a, b) = \frac{a}{a+b} \times f(a-1, b) + \frac{b}{a+b} \times f(a, b-1),$$

with $f(1, 0) = 1$.

Hint: Use the relationship you established in (a).

- **Requested identity**

The probability that A is always ahead of B in the vote count is equal to:

- (i) $f(a-1, b) = P[X_n > 0, n = 1, \dots, (a-1) + b]$, when candidate A receives $a-1$ votes and candidate B receives b votes;
- (ii) $f(a, b-1) = P[X_n > 0, n = 1, \dots, a + (b-1)]$, when candidate A receives a votes and candidate B receives $b-1$ votes.

If an additional and final vote is counted, it is either for candidate A with probability $\frac{a}{a+b}$, or for the candidate B with probability $\frac{b}{a+b}$. So the probability of candidate A being ahead throughout the count to the penultimate vote counted and also after the final vote is:

$$\begin{aligned} f(a, b) &= P(X_n > 0, n = 1, \dots, a + b) \\ &= f(a-1, b) \times \frac{a}{a+b} + f(a, b-1) \times \frac{b}{a+b}. \end{aligned}$$

QED

For more details on the “Ballot Problem”, the reader is referred to http://en.wikipedia.org/wiki/Bertrand's_ballot_theorem.¹

Group II — Probability spaces and random variables

5.5 points

1. Consider the sample space Ω , and let \mathcal{F} and \mathcal{H} be two σ – algebras on Ω . (1.5)

Show that $\mathcal{F} \cap \mathcal{H}$, the collection of subsets of Ω lying both in \mathcal{F} and \mathcal{H} , is also a σ – algebra.

- **σ – algebras on Ω**

$$\mathcal{F}, \mathcal{H}$$

- **To prove**

$\mathcal{F} \cap \mathcal{H}$ is also 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) $A_1, A_2, \dots \in \mathcal{A} \Rightarrow \bigcup_{i=1}^{+\infty} A_i \in \mathcal{A}$.

Hence, we have to prove that all these 3 postulates are true for $\mathcal{A} = \mathcal{F} \cap \mathcal{H}$.

Note that $\mathcal{F} \cap \mathcal{H}$ is the collection of subsets of Ω lying both in \mathcal{F} and \mathcal{H} . Consequently:

- (i) $\Omega \in \mathcal{F}, \mathcal{H} \Rightarrow \Omega \in \mathcal{F} \cap \mathcal{H}$;
- (ii) $A \in \mathcal{F}, \mathcal{H} \Rightarrow A^c \in \mathcal{F}, \mathcal{H} \Rightarrow A^c \in \mathcal{F} \cap \mathcal{H}$;

¹The result $f(a, b) = \frac{a-b}{a+b}$ was first published by W.A. Whitworth in 1878, but is named after Joseph Louis François Bertrand who rediscovered it in 1887.

(iii) $A_1, A_2, \dots \in \mathcal{F}, \mathcal{H} \Rightarrow \cup_{i=1}^{+\infty} A_i \in \mathcal{F}, \mathcal{H} \Rightarrow \cup_{i=1}^{+\infty} A_i \in \mathcal{F} \cap \mathcal{H}$.

I.e., $\mathcal{A} = \mathcal{F} \cap \mathcal{H}$ is also a σ -algebra on Ω .

QED

2. Let \mathcal{A} be an algebra of subsets of Ω and μ be a **finitely** additive probability measure on \mathcal{A} . (2.0)
(This requires that $\mu(\Omega) = 1$.) Moreover, consider $\{A_n, n = 1, 2, \dots\}$ a sequence of subsets of Ω such that $A_n \in \mathcal{A}, n = 1, 2, \dots$

Prove that if $A_n \downarrow \emptyset$ then $\mu(A_n) \downarrow 0$.

Hint: Consider $A_0 = \Omega$ and use the Cauchy's convergence criterion.²

• **Setting**

(Ω, \mathcal{A}) is a measurable space

μ is finitely additive probability measure on \mathcal{A} , $\mu : \mathcal{A} \rightarrow [0, 1]$, $\mu(\Omega) = 1$

• **To prove**

$A_n \downarrow \emptyset \Rightarrow \mu(A_n) \downarrow 0$ (monotone continuity for finitely additive probability measures!)

• **Proof**

Firstly, μ is finitely additive probability measure on \mathcal{A} essentially means that

$$\mu(\cup_{i=1}^n A_i) = \sum_{i=1}^n \mu(A_i),$$

for any A_1, \dots, A_n pairwise disjoint events in \mathcal{A} .

Secondly, recall that $A_n \downarrow \emptyset$ means that (thinking complements!):

(i) A_n^c is an increasing sequence of events, i.e., $\emptyset = A_0^c \subseteq A_1^c \subseteq A_2^c \subseteq \dots$

(ii) $\Omega = \lim A_n^c = \limsup A_n^c = \liminf A_n^c$.

Thirdly, Prop. 1.54, equation (1.41), suggests this probability measure is monotone (prove it!), thus, (i) implies that

$$\mu(A_n^c) \uparrow .$$

Moreover, as a result of (i) and (ii)

$$\begin{aligned} \Omega &\stackrel{\text{Prop. 1.27}}{=} \cup_{i=1}^{+\infty} A_i^c \\ \mu(\Omega) &= 1 \\ &= \mu(\cup_{i=1}^{+\infty} A_i^c) . \end{aligned}$$

Now, applying the “disjointification” technique, the finite additivity of μ and conveniently recognising a telescopic sum with $\mu(A_0^c) = \mu(\emptyset) = 0$ yields:

$$\begin{aligned} \mu(\cup_{i=1}^{+\infty} A_i^c) &= \mu([\cup_{i=1}^n (A_i^c \setminus A_{i-1}^c)] \cup [\cup_{i=n+1}^{+\infty} (A_i^c \setminus A_{i-1}^c)]) \\ &= \mu[\cup_{i=1}^n (A_i^c \setminus A_{i-1}^c)] + \mu[\cup_{i=n+1}^{+\infty} (A_i^c \setminus A_{i-1}^c)] \\ &= \sum_{i=1}^n [\mu(A_i^c) - \mu(A_{i-1}^c)] + \mu[\cup_{i=n+1}^{+\infty} (A_i^c \setminus A_{i-1}^c)] \quad (\text{because } A_{i-1}^c \subseteq A_i^c) \\ &= \mu(A_n^c) + \mu[\cup_{i=n+1}^{+\infty} (A_i^c \setminus A_{i-1}^c)] . \end{aligned}$$

²The series $\sum_{i=0}^{+\infty} a_i$ is convergent iff, for any $m \in \mathbb{N}_0$, $\sum_{i=n}^{n+m} a_i \rightarrow 0$ when $n \rightarrow +\infty$.

Finally, using the fact that

$$\lim_{n \rightarrow +\infty} \{\mu(A_n^c) + \mu[\cup_{i=n+1}^{+\infty} (A_i^c \setminus A_{i-1}^c)]\} = 1$$

and the Cauchy's convergence criterion, we can state that

$$\lim_{n \rightarrow +\infty} \mu[\cup_{i=n+1}^{+\infty} (A_i^c \setminus A_{i-1}^c)] = 0.$$

and also that

$$\begin{aligned} \lim_{n \rightarrow +\infty} \mu(A_n^c) &= \mu\left(\lim_{n \rightarrow +\infty} A_n^c\right) \\ \lim_{n \rightarrow +\infty} \mu(A_n) &= \mu\left(\lim_{n \rightarrow +\infty} A_n\right) \end{aligned}$$

Thus, $\mu(A_n) \downarrow$ and $\lim_{n \rightarrow +\infty} \mu(A_n) = \mu(\emptyset)$, that is,

$$\mu(A_n) \downarrow \mu(\emptyset) = 0.$$

QED

3. Three brothers and their partners decide to have children until each couple has two female children. (2.0)

Determine the probability that the total number of children born to the three couples does not exceed 8, by admitting that:

- $p = P(\text{birth of a female}) = 0.5$ at any time and for every couple;
- the couples procreate independently;
- the gender of the next child is independent of the gender(s) of the previous child(ren);
- there is no possibility of twins, triplets, etc.

• **R.v.**

X_i = number of children until couple i has two two female children, $i = 1, 2, 3$

• **Distribution**

$X_i \stackrel{i.i.d.}{\sim} \text{NegativeBinomial}(r, p)$, $i = 1, 2, 3$

• **Parameters**

$r = 2$

$p = P(\text{birth of a female}) = 0.5$

• **New r.v.**

$Y = \sum_{i=1}^3 X_i$ = total number of children born to the 3 couples until all have 2 female children

• **Distribution of Y**

According to Prop. 3.55,

$$Y \sim \text{NegativeBinomial}(r^*, p^*),$$

where

$$\begin{aligned} r^* &= r + r + r \\ &= 6 \\ p^* &= p \\ &= 0.5. \end{aligned}$$

• **Requested probability**

$$\begin{aligned} P(Y \leq 8) &= F_{\text{NegativeBinomial}(6,0.5)}(8) \\ &\stackrel{\text{form}}{=} 1 - F_{\text{Binomial}(8,0.5)}(6-1) \\ &\stackrel{\text{table}}{=} 1 - 0.8555 \\ &= 0.1445. \end{aligned}$$

Group III — Random variables

5.0 points

Let X be the mass/weight (in g) of an industrial article and admit that $X \sim \text{Exponential}(\lambda)$, where λ is such that $E(X) = 75$.

The only available scale automatically reduces to 100g any mass/weight larger than 100g. Let Y be the mass/weight shown by this scale.

(a) Define Y in terms of X and show that Y is a Borel measurable function, therefore also a r.v. (1.5)

• **R.v.**

$X = \text{mass/weight (in g) of an industrial article}$

• **Distribution**

$X \sim \text{Exponential}(\lambda)$

• **Parameter**

$$\begin{aligned} \lambda &: E(X) = 75 \\ &\lambda = 75^{-1} \end{aligned}$$

• **Transform**

$$\begin{aligned} Y &= \text{mass/weight shown by the scale} \\ &= \begin{cases} X, & \text{if } 0 \leq X \leq 100 \\ 100, & \text{if } X > 100 \end{cases} \\ &= \min\{X, 100\} \\ &= g(X) \end{aligned}$$

• **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**

$Y = g(X) = \min\{X, 100\}$ 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

$$\mathbb{R}_X = \mathbb{R}_0^+,$$

thus, the range of Y is given by

$$\begin{aligned} \mathbb{R}_Y &= g(\mathbb{R}_X) \\ &= [0, 100]. \end{aligned}$$

Thirdly, $g^{-1}((-\infty, y]) = Y^{-1}((-\infty, y]) = \{\omega \in \Omega : Y(\omega) \leq y\}$ equals:

– for $y < 0$,

$$\begin{aligned} \{Y \leq y\} &= \emptyset \\ &\in \mathcal{F}; \end{aligned}$$

– for $0 \leq y \leq 100$,

$$\begin{aligned} \{Y \leq y\} &= \{X \leq y\} \\ &= X^{-1}((-\infty, y]) \\ &\in \mathcal{F}. \end{aligned}$$

– for $y > 100$,

$$\begin{aligned} \{Y \leq y\} &= \Omega \\ &\in \mathcal{F}. \end{aligned}$$

Consequently, we can conclude that

$$\{Y \leq y\} \in \mathcal{F}, \forall y \in \mathbb{R},$$

hence, Y is a Borel measurable function and, thus, a r.v.

QED

(b) Derive and draw the graph of the d.f. of Y and calculate $P(\{Y = 100\})$.

• **D.f. of X**

$$F_X(x) = \begin{cases} 0, & x < 0 \\ 1 - e^{-\frac{x}{75}}, & x \geq 0 \end{cases}$$

• **D.f. of Y**

$$\begin{aligned} F_Y(y) &= P(Y \leq y) \\ &= P(\min\{X, 100\} \leq y) \\ &= \begin{cases} P(\emptyset) = 0, & y < 0 \\ P(X \leq y) = F_X(y) = 1 - e^{-\frac{y}{75}}, & 0 \leq y < 100 \\ P(\Omega) = 1, & y \geq 100 \end{cases} \end{aligned}$$

• **Graph of $F_Y(y)$**

(It has a discontinuity at $y = 100$, etc.)

• **Requested probability**

$F_Y(y)$ has a discontinuity point at $y = 100$, thus,

$$\begin{aligned} P(Y = 100) &= F_Y(100) - F_Y(100^-) \\ &= 1 - (1 - e^{-\frac{100}{75}}) \\ &= e^{-\frac{4}{3}}. \end{aligned}$$

This probability coincides with $P(X > 100)$.

(c) Describe a method to generate (pseudo-)random numbers from the distribution of Y .

• **Quantile function of X**

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

$$\begin{aligned} F_X(x) &= u \\ F_X^{-1}(u) &= x \\ &= -75 \ln(1 - u). \end{aligned}$$

• **Quantile transformation**

By Prop. 2.140, if

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

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

$$-75 \ln(1 - U) \stackrel{d}{=} X.$$

(2.0)

Consequently, if we want to generate (pseudo-)random numbers from the distribution of $Y = \min\{X, 100\}$ then we have to:

- (1) generate u from a Uniform(0, 1) distribution;
- (2) assign $x = -75 \ln(1 - u)$ and $y = \min\{x, 100\}$.

Group IV — Independence and random variables

6.5 points

1. Some form of prophylaxis is said to be 90% effective at prevention during one year's treatment. (1.5)

If the degrees of effectiveness in different years are independent, show that the treatment is more likely to fail than not to fail within 7 years.

• **R.v.**

$X_i =$ indicator of the effectiveness of treatment i , $i = 1, \dots, 7$

$X_i \stackrel{i.i.d.}{\sim} X$, $i = 1, \dots, 7$

$X \sim \text{Bernoulli}(p = 0.9)$

• **Requested probability**

$$\begin{aligned} P(\text{all 7 treatments do not fail}) &= P(X_i = 1, i = 1, \dots, 7) \\ &\stackrel{X_i \stackrel{i.i.d.}{\sim} X}{=} 0.9^7 \\ &\simeq 0.478297 \\ &< \frac{1}{2}. \end{aligned}$$

Thus the treatment is more likely to fail than not to fail within 7 years. QED

2. Each front tire on a particular type of vehicle is supposed to be filled to a pressure of 2.6 psi.³

Suppose that the actual air pressure in each front tire is a r.v. — X (resp. Y) for the right (resp. left) tire — and that the joint p.d.f. of (X, Y) equals

$$f_{X,Y}(x, y) = \begin{cases} \frac{4}{25}xy, & 2 \leq x \leq 3, 2 \leq y \leq 3 \\ 0, & \text{otherwise.} \end{cases}$$

- (a) Prove that X and Y are independent r.v. (1.0)

• **Random vector (X, Y)**

$X =$ air pressure in the left front tire

$Y =$ air pressure in the right front tire

• **Joint p.d.f.**

$$f_{X,Y}(x, y) = \begin{cases} \frac{4}{25}xy, & 2 \leq x \leq 3, 2 \leq y \leq 3 \\ 0, & \text{otherwise} \end{cases}$$

³Non-SI measures of pressure such as pounds per square inch (psi) are used primarily in the USA. 1 psi approximately equals 6894.757 Pa, where pascal (Pa) is the SI unit of pressure (http://en.wikipedia.org/wiki/Pound-force_per_square_inch).

• **Marginal p.d.f.**

$$f_X(x) = \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dy$$

$$= \begin{cases} \int_2^3 \frac{4}{25} xy dy = \frac{2x}{25} (y^2|_2^3) = \frac{2x}{5}, & 2 \leq x \leq 3 \\ 0, & \text{otherwise} \end{cases}$$

$$f_Y(y) = \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dx$$

$$= \begin{cases} \int_2^3 \frac{4}{25} xy dx = \frac{2y}{25} (x^2|_2^3) = \frac{2y}{5}, & 2 \leq y \leq 3 \\ 0, & \text{otherwise} \end{cases}$$

• **Checking if X and Y are independent r.v.**

Since

$$f_X(x) \times f_Y(y) = \begin{cases} \frac{4}{25} xy, & 2 \leq x \leq 3, 2 \leq y \leq 3 \\ 0, & \text{otherwise} \end{cases}$$

$$\equiv f_{X,Y}(x,y),$$

we can conclude that $X \perp\!\!\!\perp Y$, by Th. 3.38.

• **Requested probability**

$$P(|Y - X| \leq 0.2) = P(-0.2 \leq Y - X \leq 0.2)$$

$$= P(X - 0.2 \leq Y \leq X + 0.2)$$

$$= \int \int_{\{(x,y): x-0.2 \leq y \leq x+0.2\}} f_{X,Y}(x,y) dy dx$$

$$= \int_2^{2.2} \int_2^{x+0.2} \frac{4xy}{25} dy dx$$

$$+ \int_{2.2}^{2.8} \int_{x-0.2}^{x+0.2} \frac{4xy}{25} dy dx$$

$$+ \int_{2.8}^3 \int_{x-0.2}^3 \frac{4xy}{25} dy dx$$

$$= \dots$$

$$\stackrel{\text{Mathematica}}{=} 0.363755.$$

(b) *Derive the joint d.f. of (X, Y) .* (1.0)

• **Joint d.f. of (X, Y)**

From the independence criterion in Th. 3.30, we can add that

$$F_{X,Y}(x,y) = F_X(x) \times F_Y(y), (x,y) \in \mathbb{R}^2,$$

where:

$$F_X(x) = \int_{-\infty}^x f_X(t) dt$$

$$= \begin{cases} 0, & x < 2 \\ \int_2^x \frac{2t}{5} dt = \frac{x^2-4}{5}, & 2 \leq x \leq 3 \\ 1, & x > 3; \end{cases}$$

$$F_Y(y) = \begin{cases} 0, & y < 2 \\ \frac{y^2-4}{5}, & 2 \leq y \leq 3 \\ 1, & y > 3. \end{cases}$$

Hence,

$$F_{X,Y}(x,y) = \begin{cases} 0, & x < 2 \text{ or } y < 2 \\ \frac{x^2-4}{5} \times \frac{y^2-4}{5}, & 2 \leq x \leq 3 \text{ and } 2 \leq y \leq 3 \\ \frac{x^2-4}{5}, & 2 \leq x \leq 3 \text{ and } y > 3 \\ \frac{y^2-4}{5}, & 2 \leq y \leq 3 \text{ and } x > 3 \\ 1, & x > 3 \text{ and } y > 3. \end{cases}$$

(c) *Determine the probability that the absolute value of the difference in air pressure between the two front tires is at most 0.2 psi.* (3.0)