



BRAINWARE UNIVERSITY

Term End Examination 2020 - 21

Programme – Master of Science in Mathematics

Course Name – Measure and Probability Theory

Course Code - MSCME302

Semester / Year - Semester III

Time allotted : 75 Minutes

Full Marks : 60

[The figure in the margin indicates full marks. Candidates are required to give their answers in their own words as far as practicable.]

Group-A

(Multiple Choice Type Question)

1 x 60=60

1. *(Answer any Sixty)*

(i) <input type="image" src="/apttest/fck_image/q18.png" width="687" height="51" />

a) <input type="image" src="/apttest/fck_image/q18,1.png" width="110" height="31" />

b)

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$$= \sum P(A_k)$$

c)

d) Other

$$\geq \sum P(A_k)$$

(ii) <input type="image" src="/apttest/fck_image/q3(1).png" width="502" height="62" />

a)

b)

infinitely many

all but finitely many

c)

d)

finitely many

all

(iii)

The upper limit, $\limsup X_n$ of an increasing sequence $\{X_n\}$ of sets equals to

a) <input type="image" src="/apttest/fck_image/q5,1.png" width="58" height="50" />
c)

Does not exists

b) <input type="image" src="/apttest/fck_image/q5,2.png" width="63" height="48" />
d) Other

(iv)

Let μ is defined on an algebra of sets. Then μ is countably additive if for any finite (or countably infinite) system of (disjoint) sets X_n , we have

a) <input type="image" src="/apttest/fck_image/q19,2(1).png" width="173" height="31" />

c)

b)

$$\mu\left(\sum X_n\right) = \sum \mu(X_n)$$

d) Other

$$\mu\left(\sum X_n\right) \leq \sum \mu(X_n)$$

(v)

An extended realvalued, non-negative, countably additive function μ defined on a ring \mathbf{R} of subsets of a set X with the value at empty set zero is called a

a) measurable function
c) probability measure

b)
measure
d) None of these

(vi)

A countably additive set function is

a) b)

additive but not continuous

continuous but not additive

c)

d)

both continuous and additive

neither continuous nor additive

(vii)

A measure μ on an algebra \mathbf{R} of subsets of a set X is called totally finite measure if

a)

$\mu(X)=1$

c)

$\mu(X)$ is finite

b)

$\mu(X)=0$

d)

$\mu(X)$ is infinity

(viii)

a)

b)

$$\bigcap_{n=1}^{\infty} X_n$$

c)

d)

(ix)

A measure is said to be complete if every subset of a measurable set of is also measurable.

- a)
finite measure
c)
zero measure

- b)
infinite measure
d)
None of these

(x) <input type="image" src="/apttest/fck_image/q27.jpg" width="451" height="22" />

- a) <input type="image" src="/apttest/fck_image/q27,a.jpg" width="82" height="26" />
c) <input type="image" src="/apttest/fck_image/q27,c.jpg" width="197" height="25" />

- b) <input type="image" src="/apttest/fck_image/q27,b.jpg" width="78" height="26" />
d) Other

(xi)

Which of the following set is not a Borel set of \mathbf{R} of the set of all real numbers?

- a)
open set
c) half open interval
- b)
closed set
d) Other

(xii) <input type="image" src="/apttest/fck_image/q30.jpg" width="517" height="45" />

- a)
finite subset of X only
c)
measurable subset E of X
- b)
all subset of X
d)
None of these

(xiii)

a)

b)

$$\mu(S) \leq \mu(T)$$

$$\mu(S) \geq \mu(T)$$

c)

d)

$$\mu(S) = \mu(T)$$

Other

(xiv)

If $a \leq f(x) \leq b$ in E , then

a)

b)

$$a\mu(E) \leq F(E) \leq b\mu(E)$$

$$a\mu(E) \geq F(E) \geq b\mu(E)$$

c)

d)

$$a\mu(E) = F(E) = b\mu(E)$$

$$a\mu(E) = F(E) = b\mu(E)$$

(xv)

Three coins are tossed at random. Then the probability of the event of at least one head is

a) $3/8$

b) $7/8$

c) $2/9$

d) $5/9$

(xvi)

If F is a nondecreasing, right-continuous function satisfying....., then there exists on some probability space a random variable X for which $F(x) = P[X \leq x]$.

a)

b)

$$\lim_{x \rightarrow \infty} F(X) = 1$$

$$\lim_{x \rightarrow -\infty} F(X) = 0$$

$$\lim_{x \rightarrow \infty} F(X) = 1$$

c)

d) None of these

$$\lim_{x \rightarrow -\infty} F(X) = 0$$

(xvii)

Borel-Cantelli lemma is about

a)

b)

Sums of events

Sequences of events

c)

d) Other

Independence of events

(xviii)

A sequence of random variables $\{X_n\}$ is convergent a. e. to X implies

a)

b)

$$\lim P\{|X_n - X| \geq 1/2\} = 0$$

$$\lim P\{|X_n - X| \leq 1/2\} = 0$$

c)

d)

$$\lim P\{|X_n - X| > 1/2\} = 0$$

$$\lim P\{|X_n - X| < 1/2\} = 0$$

(xix)

A sequence of random variables $\{X_n\}$ is convergent a. e. to X . Then for every $\varepsilon > 0$

a)

b)

$$\lim_{m \rightarrow \infty} P\{|X_n - X_{n'}| < \varepsilon, n' > n \geq m\} = 1$$

$$\lim_{m \rightarrow \infty} P\{|X_n - X_{n'}| < \varepsilon, n' > n \geq m\} = 0$$

c) d)

$$\lim_{m \rightarrow \infty} P\{|X_n - X_{n'}| \geq \varepsilon, n' > n \geq m\} = 1$$

$$\lim_{m \rightarrow \infty} P\{|X_n - X_{n'}| > \varepsilon, n' > n \geq m\} = 0$$

(xx)

$X_n \rightarrow 0$ in probability

a) b)

$$\text{if } E\left(\frac{|X_n|}{1+|X_n|}\right) \rightarrow 0 \quad \text{if and only if } E\left(\frac{|X_n|}{1+|X_n|}\right) \rightarrow 0$$

c) d) Other

$$\text{only if } E\left(\frac{|X_n|}{1+|X_n|}\right) \rightarrow 0$$

(xxi)

Convergence in L^p implies that in

a) b)
 $L^r, r < p$
 c) d)

L^r , r is an integer

None of these

(xxii)

If $X_n \rightarrow X$ in L^p and $Y_n \rightarrow Y$ in $L^{\frac{p}{p-1}}$ then $X_n Y_n \rightarrow XY$ in

- a) L^p b) L^1

- c) $L^{\frac{p}{p-1}}$ d) Does not converge

(xxiii)

For each events E_n , we have

- a) $P(\limsup E_n) = \lim P\left(\bigcup_{n=m}^{\infty} E_n\right)$ b) $P(\limsup E_n) = 1 - \lim P\left(\bigcup_{n=m}^{\infty} E_n\right)$

- c) $P(\limsup E_n) = \lim P\left(\bigcap_{n=m}^{\infty} E_n\right)$ d) Other

(xxiv)

A sequence of random variables $\{X_n\}$ is convergent a. e. to 0 if for every $\varepsilon > 0$

a)

$$P\{|X_n| < \varepsilon, i.o.\} = 0$$

b)

$$P\{|X_n| > \varepsilon, i.o.\} = 0$$

c)

d) Other

$$P\{|X_n| < \varepsilon, i.o.\} = 1$$

(xxv)

The value of $\{w : \lim_{n \rightarrow \infty} X_n(w) = 0\}$ equals to

a)

$$\bigcup_{m=1}^{\infty} \bigcap_{n=m}^{\infty} \{|X_n| > \frac{1}{m}\}$$

b)

$$\bigcap_{m=1}^{\infty} \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} \{|X_n| \leq \frac{1}{m}\}$$

c)

d)

$$\bigcup_{m=1}^{\infty} \bigcap_{n=m}^{\infty} \{|X_n| > \frac{1}{n}\}$$

$$\bigcap_{m=1}^{\infty} \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} \{|X_n| > \frac{1}{k}\}$$

(xxvi)

If the events $\{E_n\}$ are independent, then

a)

$$P(E_n, i.o.) = 1$$

b)

$$P(E_n, i.o.) = 0$$

c)

d) Other

$$P(E_n, i.o.) = \infty$$

(xxvii)

If the events $\{E_n\}$ are pair-wise independent, then the value of $P(\limsup E_n) = 1$ if

a)

$$\sum_n P(E_n) = \infty$$

b)

$$\sum_n P(E_n) < \infty$$

c)

d) Other

$$\sum_n P(E_n) = 0$$

(xxviii)

A sequence of random variables $\{x_n\}$ converges in probability to a random variable x .

Then $\lim_{n \rightarrow \infty} p(|x_n - x| > 1/2)$ equals to

a) 0

b) 1/2

c) 1

d) Other

(xxix)

If $x_n \rightarrow x$ in probability then

a)

$$F_n(x_n) \rightarrow F(x)$$

b)

$$F_n(x) \rightarrow F(x)$$

c) <input type="image"

d) Other

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(xxx)

If x_n has mean \bar{x}_n and standard deviation σ_n then $x_n - \bar{x}_n$ converges to

- a) 0 if $\sigma_n \rightarrow 0$ b) 1 if $\sigma_n \rightarrow 0$
c) 0 d) 1

(xxxi)

Let $y_k, k = 1, 2, \dots$ be independent random variables with mean $\bar{y}_k, k = 1, 2, \dots$ and standard deviation $\sigma_k, k = 1, 2, \dots$ respectively. Let $x_n = \frac{1}{n} \sum_{k=1}^n y_k$, $m_n = \frac{1}{n} \sum_{k=1}^n \bar{y}_k$. Then $x_n - m_n \rightarrow 0$ in probability if

- a) $\sum_{k=1}^n \sigma_k^2 = O(n)$ b) $\sum_{k=1}^n \sigma_k^2 = O(n^2)$
c) $\sum_{k=1}^n \sigma_k^2 \rightarrow 0$ d) Other

(xxxii)

Let θ be a irrational number then $\theta, 2\theta, 3\theta, \dots$ is

- a) Uniformly distributed modulo 1 b) Uniformly distributed module 0
c) d) Other

Not uniformly distributed

(xxxiii)

A K – L function is..... in every finite interval.

a) b)

unbounded bounded

c) d) Other

continuous

(xxxiv)

Let $\psi(t)$ be a K-L function with representation (a, G). Then G is.....in R

a) b)

unbounded bounded

c) d) None of these

continuous

(xxxv)

Let $\psi(t)$ be a K-L function with representation (a, G). Then G is bounded only in

- | | |
|---|---|
| a) the set of all positive real numbers | b) the set of all negative real numbers |
| c) the set of all real numbers | d) on the finite interval |

(xxxvi)

Let $\psi_n(t)$ has K-L representation (α_n, G_n) and $\psi_n(t) \rightarrow \psi(t), \forall t$. Then $\psi(t)$ have K-L representation (α, G) and

a)

$$\alpha_n \rightarrow \alpha$$

c) All the above

b)

$$G_n \rightarrow G$$

d) None of the above

(xxxvii)

If $X_n \Rightarrow X$ and X_n are uniformly integrable, then X is

a)

always integrable

b)

never uniformly integrable

c)

d) Other

may not be integrable

(xxxviii)

$\mu_n(A) \rightarrow \mu(A), \forall A$, where A is a μ -continuity set if and only if
 $\int f d\mu_n \rightarrow \int f d\mu$, where f

a)

bounded real functions

b)

continuous real functions

c)

d) None of these

bounded and continuous real functions

(xxxix)

If $X_n \Rightarrow X$ and $X_n - Y_n \Rightarrow 0$. Then

a)

b)

$$X_n Y_n \Rightarrow X$$

$$X_n Y_n \Rightarrow 2X$$

c)

d) Other

$$Y_n \Rightarrow X$$

(xl)

For every sequence $\{F_n\}$ of distribution functions there exists a subsequence $\{F_{n_k}\}$ and a non-decreasing, function F such that $\lim F_{n_k}(x) = F(x)$ at continuity points x of F .

a)

b)

non-increasing

non-decreasing

c)

d)

strictly-decreasing

strictly- increasing

(xli)

The variance of the random variable is

a)

b)

$$\int_X |x - \bar{x}| d\mu$$

c)

$$\int_X |x - \bar{x}|^2 d\mu$$

d) Other

$$\sqrt{\int_X |x - \bar{x}| d\mu}$$

(xlvi)

Let X and Y be two random variables such that $Y = a + bX$ where a and b are two constants. Then $Var(Y)$ is

a)

$$b^2 Var(X)$$

c)

$$a + b^2 Var(X)$$

b)

$$a^2 Var(X)$$

d)

$$a^2 + b^2 Var(X)$$

(xlvi)

Let $\alpha(x)$ be a non-negative function of the random variable x and $k > 0$ then

a)

$$\frac{1}{k} E(\alpha(x)) \leq P(\alpha(x) \geq k)$$

c)

$$E(\alpha(x)) \leq P(\alpha(x) \geq k)$$

b)

$$\frac{1}{k} E(\alpha(x)) \geq P(\alpha(x) \geq k)$$

d)

$$\frac{1}{k} E(\alpha(x)) \leq P(\alpha(x) \leq k)$$

(xliv)

Let $\alpha(x)$ be a non-negative function of the random variable x then

a)

$$E(\alpha(x)) \leq P(\alpha(x) \geq 1)$$

b)

$$E(\alpha(x)) \geq P(\alpha(x) \geq 1)$$

c)

d) Other

$$E(\alpha(x)) = P(\alpha(x) \geq 1)$$

(xlv)

The mean of a Poisson distribution with parameter μ is

a)

$$\mu$$

b)

$$\mu^2$$

c)

d) Other

$$2\mu$$

(xlvi)

If (x_1, x_2, \dots, x_n) is a real random vector, then

a)

$$\begin{aligned} E(x_1 + x_2 + \dots + x_n) \\ = E(x_1, x_2, \dots, x_n) \end{aligned}$$

b)

$$\begin{aligned} E(x_1 + x_2 + \dots + x_n) \\ = E(x_1) + E(x_2) + \dots + E(x_n) \end{aligned}$$

c)

d)

$$E(x_1 + x_2 + \dots + x_n) \\ \leq E(x_1, x_2, \dots, x_n)$$

$$E(x_1 + x_2 + \dots + x_n) \\ \leq E(x_1) + E(x_2) + \dots + E(x_n)$$

(xlvii)

Let $\alpha(x_1, x_2)$ be a function of two independent variables x_1, x_2 and $X = X_1 \times X_2$. Then

a)

$$\int_X \alpha(x_1, x_2) d\mu = \iint_X \alpha(x_1, x_2) d\mu^2$$

b)

$$\int_X \alpha(x_1, x_2) d\mu = \iint_X \alpha(x_1, x_2) d\mu_1 d\mu_2$$

c)

d) Other

$$\int_X \alpha(x_1, x_2) d\mu = 0$$

(xlviii) The sum of independent variables with binomial distributions

- | | |
|---------------------------------------|------------------------------------|
| a) may not be a binomial distribution | b) must be a binomial distribution |
| c) never a binomial distribution | d) None of these |

(xlix) The sum of independent variables of normal distributions is

- | | |
|-------------------------------------|----------------------------------|
| a) may not be a normal distribution | b) must be a normal distribution |
| c) never a normal distribution | d) None of these |

(li) The sum of two independent variables has Poisson distributions then

- | | |
|--|---|
| a) At-least one has Poisson distribution | b) Both variables have Poisson distribution |
| c) At-most one variable has Poisson distribution | d) Both variables may not have Poisson distribution |

(li)

Let x_1, x_2 be two independent variables with their distribution function F_1, F_2 respectively then their sum has the distribution function F given by the formula

a)

b)

$$F(x) = F_1(x) + F_2(x)$$

$$F(x) = F_1(x)F_2(x)$$

c)

d) Other

$$F(x) = F_1^2(x)F_2(x)$$

(lii)

The characteristic function $\varphi(t)$ of a distribution function $F(x)$ satisfies

a)

b)

$$\varphi(t) = 1$$

$$|\varphi(t)| = 1$$

c)

d)

$$|\varphi(t)| \leq 1$$

$$|\varphi(t)| \geq 1$$

(liii)

The characteristic function $\varphi(t)$ of a distribution function $F(x)$ satisfies

a)

b)

$$\varphi(0) = 1$$

$$\varphi(t) = 1$$

c)

d)

$$\varphi(t) = 0$$

$$|\varphi(t)| = 1$$

(liv)

Let A be a support of a distribution m. Then $m(A)$ is equal to

- | | |
|-------------------------------|----------|
| a) 0 | b) 1 |
| c) Non conclusion can be done | d) Other |

(lv)

The sum of independent variables with Poisson distributions c_1, c_2 is Poisson and has parameter

- | | |
|-----------------|-----------------|
| a) | b) |
| $c_1 \cdot c_2$ | $(c_1 + c_2)^2$ |
| c) | d) |
| $c_1 + c_2$ | None of these |

(lvi)

Let A and B are two m-measurable sets and A is a subset of B. If $m(B)=0$, then $m(A)$

- | | |
|----------------|----------|
| a) 1 | b) 0 |
| c) less than 0 | d) Other |

(lvii) The mean of a Poisson distribution with parameter 10 is

- | | |
|--------|---------|
| a) 10 | b) -10 |
| c) 100 | d) -100 |

(lviii)

Which of the following is a ring of subsets of $X=\{1,2,3\}$?

- | | |
|-------------------------------------|---|
| a) $\{\emptyset, X, \{1\}, \{2\}\}$ | b) $\{\emptyset, X, \{1\}, \{2\}, \{1, 2\}\}$ |
| c) $\{\emptyset, X\}$ | d) Other |

(lix)

Let X be a finite set. Then the class of all finite subsets of X is not a/an

- a) ring
- b) algebra
- c) algebra but a ring
- d) Other

(lx) <div>An extended real valued set function m defined on a class E of sets is additive if for any two disjoint sets A and B belong to the class, we have:</div>

- a) $m(A \cup B) < m(A) + m(B)$
- b) $m(A \cup B) > m(A) + m(B)$
- c) $m(A \cup B) = m(A) + m(B)$
- d) Other