



1 AXIOMS OF PROBABILITY

1.1 Definitions

1. For any event A , $0 \leq \Pr(A) \leq 1$.
2. $A \cup B \triangleq A + B$.
3. $A \cap B \triangleq AB$.
4. The null and complete event are $\phi = 0, S = 1$.
5. If $AB = 0$, $\Pr(A + B) = \Pr(A) + \Pr(B)$.
6. $(A + B)' = A'B'$

1.2 Problems

Prove the following:

1.

$$A = AB + AB' \quad (1.2.1.1)$$

2.

$$\Pr(A) = \Pr(AB) + \Pr(AB') \quad (1.2.2.1)$$

3.

$$A + B = B + AB' \quad (1.2.3.1)$$

4.

$$\Pr(A + B) = \Pr(A) + \Pr(B) - \Pr(AB) \quad (1.2.4.1)$$

2 DISTRIBUTION OF THE SUM OF RANDOM VARIABLES

2.1 Definitions

1. The mean of X is defined as

$$E(X) = \sum_k k p_X(k) \quad (2.1.1.1)$$

2. The Z transform of X is defined as

$$M_X(z) = E(z^{-X}) = \sum_{k=-\infty}^{\infty} z^{-k} p_X(k) \quad (2.1.2.1)$$

3. There is a one to one relationship between the pmf and its Z transform.
4. If X_1 and X_2 are independent,

$$E[f(X_1)g(X_2)] = E[f(X_1)]E[g(X_2)] \quad (2.1.4.1)$$

5. For a Bernoulli random variable X , the pmf is

$$p_X(n) = \begin{cases} p & k = 1 \\ 1 - p & k = 0 \\ 0 & \text{otherwise} \end{cases} \quad (2.1.5.1)$$

6. X_i are said to be i.i.d (independent and identically distributed) if they are independent and have the same pmf.

2.2 Problem

1. Find the Z-transform for X , given that X is a Bernoulli random variable with parameter p .
2. If X_1 and X_2 are independent, and

$$Y = X_1 + X_2, \quad (2.2.2.1)$$

show that

$$M_Y(z) = M_{X_1}(z)M_{X_2}(z) \quad (2.2.2.2)$$

3. Find the Z-transform of Y , given that X_i are i.i.d Bernoulli random variables with parameter p .
4. Find the pmf of Y .
5. Find the pmf of

$$Y = \sum_{i=1}^N X_i, \quad (2.2.5.1)$$

where X_i are i.i.d.

3 MOMENTS AND VARIANCE

3.1 Definitions

1. The variance of X is defined as:

$$Var(X) = E(X - E(X))^2 \quad (3.1.1.1)$$

2. The Z transform of X is defined as

$$M_X(z) = E(z^{-X}) = \sum_{k=-\infty}^{\infty} z^{-k} p_X(k) \quad (3.1.2.1)$$

3. Let X be a random variable with pmf.

$$p_X(k) = \begin{cases} 1/6 & 1 \leq k \leq 6 \\ 0 & \text{otherwise} \end{cases} \quad (3.1.3.1)$$

X is said to be Discrete Uniform Random Variable

4. The n^{th} moment of X is defined as:

$$E(X^n) = \sum_{k=-\infty}^{\infty} k^n p_X(k) \quad (3.1.4.1)$$

3.2 Problems

1. Show that $Var(X) = E(X^2) - [E(X)]^2$
2. Find $M_X(z)$
3. Show that $E(X) = \frac{d}{dz} M_X(z^{-1})|_{z=1}$
4. Find $E(X^2)$
5. Find $Var(X)$.

4 CONVOLUTION

4.1 Definitions

1. The Z transform of X is defined as

$$M_X(z) = E(z^{-X}) = \sum_{k=-\infty}^{\infty} z^{-k} p_X(k) \quad (4.1.1.1)$$

2. Let X be a random variable with pmf.

$$p_X(k) = \begin{cases} 1/6 & 1 \leq k \leq 6 \\ 0 & \text{otherwise} \end{cases} \quad (4.1.2.1)$$

X is said to be Discrete Uniform Random Variable

3. Convolution of two sequences using Toeplitz matrices

$$\mathbf{y} = \mathbf{x} \otimes \mathbf{h} \quad (4.1.3.1)$$

$$\mathbf{y} = \begin{pmatrix} h_1 & 0 & \cdot & \cdot & \cdot & 0 \\ h_2 & h_1 & \cdot & \cdot & \cdot & 0 \\ h_3 & h_2 & h_3 & \cdot & \cdot & 0 \\ h_{m-1} & \cdot & \cdot & \cdot & h_2 & h_1 \\ h_m & h_{m-1} & \cdot & \cdot & \cdot & h_2 \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \cdot & \cdot & \cdot & h_m \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \cdot \\ \cdot \\ \cdot \\ \cdot \\ \cdot \\ x_n \end{pmatrix} \quad (4.1.3.2)$$

4.2 Problems

1. If $\mathbf{x} = \mathbf{h} = \frac{1}{2} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$, find \mathbf{y} .
2. Find $p_{X_1}(k) \otimes p_{X_2}(k)$ using toeplitz matrices.
3. Find $M_Y(z)$, such that $Y = X_1 + X_2$
4. Find $p_Y(k)$

5 Z-TRANSFORM APPLICATIONS

5.1 Definitions

1.

$$u(n) = \begin{cases} 1 & n \geq 0 \\ 0 & n < 0 \end{cases} \quad (5.1.1.1)$$

2. The Z transform of X is defined as

$$M_X(z) = E(z^{-X}) = \sum_{k=-\infty}^{\infty} z^{-k} p_X(k) \quad (5.1.2.1)$$

5.2 Problems

1. If

$$p_Y(n) \xleftrightarrow{Z} M_Y(z), \quad (5.2.1.1)$$

show that

$$p_Y(n-k) \xleftrightarrow{Z} M_Y(z)z^{-k}, \quad (5.2.1.2)$$

2. Show that

$$u(n) \xleftrightarrow{z} \frac{1}{(1 - z^{-1})}, \quad |z| > 1 \quad (5.2.2.1)$$

3. Show that

$$nu(n) \xleftrightarrow{z} \frac{z^{-1}}{(1 - z^{-1})^2}, \quad |z| > 1 \quad (5.2.3.1)$$

4. Let

$$M_Y(z) = \left\{ \frac{z^{-1}(1 - z^{-6})}{6(1 - z^{-1})} \right\}^2, \quad |z| > 1 \quad (5.2.4.1)$$

Show that

$$p_Y(n) = \frac{(n-1)u(n-1) - 2(n-7)u(n-7) + (n-13)u(n-13)}{36} \quad (5.2.4.2)$$

6 MARKOV CHAIN

6.1 Definitions

1. Fig. 6.1.1.1 shows a Markov chain with 5 states. Transition from one state to another happens over time. s_0 and s_4 are absorbing states.

$$p + q = 1 \quad (6.1.1.1)$$

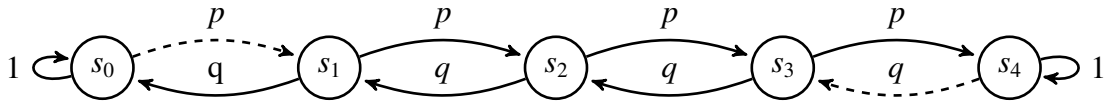


Fig. 6.1.1.1

2. At time instant n ,

$$\mathbf{p}^{(n)} = \begin{pmatrix} P_0^{(n)} \\ P_1^{(n)} \\ P_2^{(n)} \\ P_3^{(n)} \\ P_4^{(n)} \end{pmatrix} \quad (6.1.2.1)$$

where $P_i^{(n)}$ are defined to be the *stationary* probabilities.

3. P_{ij} is defined as the *transition* probability of going to state i from state j .

4. For a matrix \mathbf{A} , let

$$\mathbf{A}\mathbf{x} = \lambda\mathbf{x}. \quad (6.1.4.1)$$

Then λ is a scalar defined to be the eigenvalue of \mathbf{A} and \mathbf{x} is the corresponding eigenvector.

6.2 Problems

1. Let

$$P_0^{(n+1)} = P_0^{(n)} \quad (6.2.1.1)$$

$$P_1^{(n+1)} = pP_0^{(n)} + qP_2^{(n)} \quad (6.2.1.2)$$

$$P_2^{(n+1)} = pP_1^{(n)} + qP_3^{(n)} \quad (6.2.1.3)$$

$$P_3^{(n+1)} = pP_2^{(n)} + qP_4^{(n)} \quad (6.2.1.4)$$

$$P_4^{(n+1)} = P_4^{(n)} \quad (6.2.1.5)$$

Find the matrix \mathbf{P} such that $\mathbf{p}^{(n+1)} = \mathbf{P}\mathbf{p}^{(n)}$

2. Show that 1 is an eigen value of \mathbf{P} .

3. Show that

$$P_2^{(n+1)} = pP_1^{(n)} + qP_3^{(n)} \quad (6.2.3.1)$$

4. If $P_0 = 1, P_N = 0$ and

$$P_i = pP_{i-1} + qP_{i+1} \quad (6.2.4.1)$$

show that

$$P_i = \frac{\left(\frac{p}{q}\right)^i - \left(\frac{p}{q}\right)^N}{1 - \left(\frac{p}{q}\right)^N}, 0 \leq i \leq N \quad (6.2.4.2)$$

7 GAUSSIAN DISTRIBUTION

7.1 Definitions

1. The CDF of X is defined as,

$$F_X(x) = \Pr(X \leq x) \quad (7.1.1.1)$$

2. The PDF of X is defined as,

$$p_X(x) = \frac{d}{dx}F_X(x) \quad (7.1.2.1)$$

3. Let $X \sim \mathcal{N}(0, 1)$. Then the Q function is defined as,

$$Q(x) = \Pr(X > x), \quad x \geq 0 \quad (7.1.3.1)$$

7.2 Problems

1. Find

$$\Pr(|X - \mu| \leq k\sigma) \quad (7.2.1.1)$$

in terms of Q function.

2. Find

$$\Pr(X \leq x, |X - \mu| \leq k\sigma) \quad (7.2.2.1)$$

in terms of $F_X(x)$

3. Find

$$F_X(x | |X - \mu| \leq k\sigma) \quad (7.2.3.1)$$

4. Find

$$p_X(x | |X - \mu| \leq k\sigma) \quad (7.2.4.1)$$

8 BIVARIATE GAUSSIAN

8.1 Definitions

1. Let

$$\mathbf{X} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \boldsymbol{\mu} = E(\mathbf{X}) \quad (8.1.1.1)$$

$$\boldsymbol{\Sigma}_{\mathbf{x}} = E[(\mathbf{x} - \boldsymbol{\mu})(\mathbf{x} - \boldsymbol{\mu})^\top] \quad (8.1.1.2)$$

Then $\boldsymbol{\Sigma}_{\mathbf{x}}$ is defined to be the *covariance* matrix of \mathbf{x} .

2. For $\mathbf{x} \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma}_{\mathbf{x}})$,

$$f_{\mathbf{X}}(\mathbf{x}) = \frac{1}{2\pi \sqrt{|\boldsymbol{\Sigma}_{\mathbf{x}}|}} \exp -\frac{1}{2} \left((\mathbf{x} - \boldsymbol{\mu})^\top \boldsymbol{\Sigma}_{\mathbf{x}}^{-1} (\mathbf{x} - \boldsymbol{\mu}) \right) \quad (8.1.2.1)$$

3. The *correlation coefficient* is defined as

$$\rho = \frac{E[(x_1 - \mu_1)(x_2 - \mu_2)]}{\sigma_1 \sigma_2} \quad (8.1.3.1)$$

where μ_i, σ_i^2 are the mean and variance of x_i .

8.2 Problems

1. Show that

$$E[(\mathbf{x} - \boldsymbol{\mu})(\mathbf{x} - \boldsymbol{\mu})^\top] = \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix} \quad (8.2.1.1)$$

2. Prove that $\boldsymbol{\Sigma}_{\mathbf{x}}$ is a diagonal matrix when x_1 and x_2 are independent.
3. Let

$$z_1 = x_1 + x_2 \quad (8.2.3.1)$$

$$z_2 = x_1 - x_2 \quad (8.2.3.2)$$

Find \mathbf{P} such that

$$\mathbf{z} = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \mathbf{P}\mathbf{x} \quad (8.2.3.3)$$

4. Show that

$$\boldsymbol{\Sigma}_{\mathbf{z}} = \mathbf{P}\boldsymbol{\Sigma}_{\mathbf{x}}\mathbf{P}^\top \quad (8.2.4.1)$$

5. Check the independence of z_1 and z_2 given that $\sigma_1 = \sigma_2$.
6. Show that columns of \mathbf{P} are eigenvectors of $\boldsymbol{\Sigma}_{\mathbf{z}}$.
7. Show that the eigenvectors of $\boldsymbol{\Sigma}_{\mathbf{z}}$ are orthogonal to each other.
8. Summarize your conclusion in one line.

9 TRANSFORMATION OF RANDOM VARIABLES

9.1 Definitions

1. The pdf of $X \sim \mathcal{N}(\mu, \sigma^2)$ is defined as

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right] \quad -\infty < x < \infty \quad (9.1.1.1)$$

2. The Jacobian matrix transforming R, Θ to X_1, X_2 is defined as

$$\mathbf{J} = \begin{pmatrix} \frac{\partial X_1}{\partial R} & \frac{\partial X_1}{\partial \Theta} \\ \frac{\partial X_2}{\partial R} & \frac{\partial X_2}{\partial \Theta} \end{pmatrix} \quad (9.1.2.1)$$

- 3.

$$p_{R,\Theta}(r, \theta) = p_{X_1, X_2}(x_1, x_2) |\mathbf{J}| \quad (9.1.3.1)$$

4. The marginal distribution

$$p_R(r) = \int_0^{2\pi} p_{R,\Theta}(r, \theta) d\theta \quad (9.1.4.1)$$

5. The Laplace transform of $p_Y(y)$ is given by

$$M_Y(s) = E(e^{-sY}) \quad (9.1.5.1)$$

6. The unit step function is defined as

$$u(y) = \begin{cases} 1 & y \geq 0 \\ 0 & \text{otherwise} \end{cases} \quad (9.1.6.1)$$

9.2 Problems

1. Let $X_1, X_2 \in N \sim (0, 1)$ be i.i.d. Find $p_{X_1, X_2}(x_1, x_2)$.

2. Let

$$X_1 = R \cos \Theta \quad (9.2.2.1)$$

$$X_2 = R \sin \Theta \quad (9.2.2.2)$$

Find $p_{R,\theta}(r, \theta)$.

3. Find $p_R(r)$.

4. Find $p_\Theta(\theta)$.

5. Find the distribution of

$$Y = X_1^2 + X_2^2 \quad (9.2.5.1)$$

6. Find the Laplace transform of $e^{-y}u(y)$

7. Find the Laplace transform of $p_{X_1^2}(x_1)$.

8. Find the Laplace transform of $p_Y(y)$ using (9.2.5.1).

9. Find the distribution of Y .

10 ORDER STATISTICS

10.1 Definitions

1. The pdf of an *exponential* distribution is given by

$$p_X(x) = e^{-x}u(x) \quad (10.1.1.1)$$

where $u(\cdot)$ is the unit step function.

10.2 Problems

1. Find $F_X(x)$.
2. Let X and Y be iid exponential. Find $F_{XY}(z, z)$
3. Show that

$$\Pr(X \leq z, X > Y) = \int_{x=-\infty}^z \int_{y=-\infty}^x p_{X,Y}(x, y) dx dy \quad (10.2.3.1)$$

$$= \frac{e^{-2z}}{2} - e^{-z} \quad (10.2.3.2)$$

4. Find

$$\Pr(Y \leq z, X \leq Y) \quad (10.2.4.1)$$

5. Let

$$Z = \max(X, Y) \quad (10.2.5.1)$$

Show that

$$F_Z(z) = \Pr(X \leq z, X > Y) + \Pr(Y \leq z, X \leq Y) \quad (10.2.5.2)$$

6. Find the pdf of Z .
7. Find the pdf of $W = \min(X, Y)$.
8. Find $F_W(vZ)$, where v is a constant.
9. Find $E[F_W(vZ)]$.
10. Find the pdf of $V = \frac{W}{Z}$.

11 SOME DISTRIBUTIONS

11.1 Definitions

1. If $Y = f(Z)$ be monotonic,

$$p_Y(y) dy = p_Z(z) dz \quad (11.1.1.1)$$

2. The pdf of the $\chi^2(k)$ distribution is given by

$$p_X(x) = \frac{x^{\frac{k}{2}-1} e^{-\frac{x}{2}}}{2^{\frac{k}{2}} \Gamma\left(\frac{k}{2}\right)} u(x) \quad (11.1.2.1)$$

3. The Beta function is defined as

$$B(x, y) = \frac{\Gamma(x)\Gamma(y)}{\Gamma(x+y)} \quad (11.1.3.1)$$

11.2 Problems

1. Let $X_1 \sim \chi^2(m)$ and $X_2 \sim \chi^2(n)$ be independent. For

$$Z = \frac{X_1/m}{X_2/n}, \quad (11.2.1.1)$$

show that

$$F_Z(z) = E \left[F_{X_1} \left(\frac{mzX_2}{n} \right) \right] \quad (11.2.1.2)$$

2. Show that

$$p_Z(z) = E \left[\frac{mX_2}{n} p_{X_1} \left(\frac{mzX_2}{n} \right) \right] \quad (11.2.2.1)$$

3. Show that

$$p_Z(z) = \frac{\left(\frac{m}{n}\right)^{m/2}}{B\left(\frac{m}{2}, \frac{n}{2}\right)} x^{\frac{m}{2}-1} \left(1 + \frac{m}{n}z\right)^{-\frac{m+n}{2}} u(z) \quad (11.2.3.1)$$

Z has an F distribution with (m, n) degrees of freedom.

4. Show that $Y = \frac{1}{Z}$ is monotonic.
 5. Show that Y also has an F distribution with (n, m) degrees of freedom.
 6. Find the pdf of $\frac{mZ}{mZ+n}$.

12 INEQUALITIES

12.1 Definitions

1. The mean of Y is defined as

$$E(Y) = \sum_k k p_Y(k) \quad (12.1.1.1)$$

2. The MGF of X is defined as

$$M_X(s) = E(e^{-sX}) \quad (12.1.2.1)$$

3. Let $X \sim \mathcal{N}(0, 1)$. Then the Q function is defined as,

$$Q(x) = \Pr(X > x), \quad x \geq 0 \quad (12.1.3.1)$$

12.2 Problems

1. Show that

$$E(Y) \geq \sum_{k=m}^{\infty} k p_Y(k), \quad m > 0 \quad (12.2.1.1)$$

2. Show that

$$\Pr(Y > m) \leq \frac{E(Y)}{m}, \quad m > 0 \quad (12.2.2.1)$$

3. Using (12.2.2.1), show that Show that

$$\Pr([Y - E(Y)]^2 > b^2) \leq \frac{\text{var}(Y)}{b^2}, \quad b > 0 \quad (12.2.3.1)$$

4. Show that

$$\Pr(|Y - E(Y)| > b) \leq \frac{\text{var}(Y)}{b^2}, \quad b > 0 \quad (12.2.4.1)$$

5. Show that

$$\Pr(X > a) = \Pr(e^{-sX} > e^{-sa}), \quad s < 0 \quad (12.2.5.1)$$

6. Using (12.2.2.1), show that

$$\Pr(X > a) \leq e^{as} M_X(s), \quad s < 0 \quad (12.2.6.1)$$

7. Show that the MGF of X is

$$M_X(s) = e^{\frac{1}{2}s^2} \quad (12.2.7.1)$$

8. Using (12.2.6.1) show that

$$Q(x) \leq e^{-\frac{x^2}{2}} \quad (12.2.8.1)$$

13 JENSEN'S INEQUALITY

13.1 Definitions

1. A function g is said to be convex if

$$g(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda g(x_1) + (1 - \lambda)g(x_2) \quad (13.1.1.1)$$

2. g is convex if and only if

$$g''(x) \geq 0 \quad (13.1.2.1)$$

3. The *information* associated with an event E is defined as

$$I(E) = -\log_2(p(E)) \quad (13.1.3.1)$$

4. The *entropy* of a random variable X is given by

$$H(X) = -E[\log_2(p(X))] \quad (13.1.4.1)$$

5. For a *bernoulli* random variable $Y \in \{0, 1\}$, the p.m.f is given by

$$p_Y(k) = \begin{cases} p & k = 0 \\ 1 - p & k = 1 \end{cases} \quad (13.1.5.1)$$

13.2 Problems

1. Show that $\log\left(\frac{1}{x}\right)$ is convex.
2. Show that

$$H(Y) = -p \log_2 p - (1 - p) \log_2 (1 - p) \quad (13.2.2.1)$$

and find the maximum value of $H(Y)$.

3. Let $X \in \{x_1, x_2\}$ and

$$\lambda = q = p_X(X = x_1) \quad (13.2.3.1)$$

in (13.1.1.1). Show that

$$E[g(X)] \geq g[E(X)]. \quad (13.2.3.2)$$

4. Using (13.2.3.2), show that

$$H(Y) \leq 1 \quad (13.2.4.1)$$

14 MAXIMUM LIKELIHOOD CONDITION

14.1 Definitions

1. The pmf for a Binomial Random variable $Y \sim (n, p)$ is given by

$$p_Y(Y = k) = \binom{n}{k} p^k (1 - p)^{n-k}, \quad 0 \leq k \leq n \quad (14.1)$$

2. The pdf of an Exponential Random variable X is given by

$$p_X(x) = ce^{-cx}u(x) \quad (14.1)$$

14.2 Problems

1. Show that the solution of

$$\max_p p_Y(Y = k) \quad (14.2.1.1)$$

is

$$\hat{p} = \frac{k}{n} \quad (14.2.1.2)$$

2. Find

$$\Pr(X > T) \quad (14.2.2.1)$$

where T is a constant.

3. If X represents the lifetime of a bulb, show that the value of c that maximizes the probability of k out of n bulbs working after T hours is

$$c = \frac{1}{T} \log\left(\frac{n}{k}\right) \quad (14.2.3.1)$$

15 MAXIMUM LIKELIHOOD CONDITION

15.1 Definitions

1. The pmf for a Bernoulli r.v. X_i is given by

$$p_{X_i}(k) = \begin{cases} p_0 & k = 1 \\ 1 - p_0 & k = 0 \\ 0 & \text{otherwise} \end{cases} \quad (15.1)$$

2. The pmf for a Binomial Random variable

$$X = \sum_{i=1}^n X_i \quad (15.1)$$

is given by

$$p_X(X = k) = \binom{n}{k} p_0^k (1 - p_0)^{n-k}, \quad 0 \leq k \leq n \quad (15.2)$$

where X_i are i.i.d.

3. Let $Z \sim \mathcal{N}(0, 1)$. Then the Q function is defined as,

$$Q(z) = \Pr(Z > z), \quad z \geq 0 \quad (15.1)$$

4. The MGF of Y is defined as

$$M_Y(s) = E(e^{-sY}) \quad (15.1)$$

5. The MGF of Z is

$$M_Z(s) = \frac{1}{2} e^{\frac{1}{2}s^2} \quad (15.1)$$

6.

$$\lim_{n \rightarrow \infty} \left(1 + \frac{1}{n}\right)^n = e \quad (15.1)$$

15.2 Problems

1. Show that the mean and variance of X_i are $\mu = p_0$ and $\sigma^2 = p(1 - p)$.
2. Show that the mean and variance of

$$Y_i = \frac{X_i - \mu}{\sigma} \quad (15.1)$$

are 0 and 1 respectively.

3. Show that the mean and variance of

$$Y = \frac{1}{\sqrt{n}} \sum_{i=1}^n Y_i \quad (15.1)$$

are 0 and 1 respectively.

4. Show that

$$M_Y(s) = \left[E \left(e^{-\frac{sY_i}{\sqrt{n}}} \right) \right]^n \quad (15.1)$$

5. Show that

$$E \left(e^{-\frac{sY_i}{\sqrt{n}}} \right) = 1 + \frac{1}{2n} + \frac{1}{n} R(s, n) \quad (15.1)$$

where $R(s, n)$ is an infinite series.

6. Show that

$$\lim_{n \rightarrow \infty} Y = Z \quad (15.1)$$

This is known as the *Central Limit Theorem*.

7. Let

$$p = \frac{1}{n} \sum_{i=1}^n X_i \quad (15.1)$$

Show that

$$E(p) = \mu = p_0 \quad (15.2)$$

$$\text{var}(p) = \frac{\sigma^2}{n} = \frac{p_0(1 - p_0)}{n} \quad (15.3)$$

8. Show that

$$\lim_{n \rightarrow \infty} \frac{p - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}} = Z \quad (15.1)$$

9. Find z if

$$\Pr(|Z| < z) = 1 - \alpha \quad (15.1)$$

10. Show that

$$p_0 - \sqrt{\frac{p_0(1-p_0)}{n}} Q^{-1}\left(\frac{\alpha}{2}\right) < p < p_0 + \sqrt{\frac{p_0(1-p_0)}{n}} Q^{-1}\left(\frac{\alpha}{2}\right) \quad (15.1)$$

11. Among 4000 newborns, 2080 are male. Find the $1 - \alpha = 0.99$ confidence interval of the probability that a male child is born.