

Examiner: Xiangfeng Yang (Tel: 013 28 57 88).

Things allowed (Hjälpmedel): a calculator.

Scores rating (Betygsgränser): 8-11 points giving rate 3; 11.5-14.5 points giving rate 4; 15-18 points giving rate 5.

## 1 (3 points)

A fair die is tossed and the upper side is observed. The sample space here can be denoted as  $S = \{1, 2, 3, 4, 5, 6\}$ . Let us consider the following three events:

$$A = \{1, 3, 5\}, \quad B = \{2, 4, 6\}, \quad C = \{1, 6\}.$$

(1.1) (1p) Are the two events  $A$  and  $B$  independent? Why?

(1.2) (1p) Are the two events  $A$  and  $C$  independent? Why?

(1.3) (1p) Find the conditional probability  $P(B|C)$ .

*Solution.* (1.1) The two events  $A$  and  $B$  are NOT independent because  $P(A \cap B) \neq P(A) \cdot P(B)$  :

$$P(A \cap B) = P(\emptyset) = 0, \quad P(A) = \frac{1}{2}, \quad P(B) = \frac{1}{2}.$$

(1.2) The two events  $A$  and  $C$  are independent because  $P(A \cap C) = P(A) \cdot P(C)$  :

$$P(A \cap C) = P(\{1\}) = \frac{1}{6}, \quad P(A) = \frac{1}{2}, \quad P(C) = \frac{1}{3}.$$

(1.3) The conditional probability is computed as:

$$P(B|C) = \frac{P(B \cap C)}{P(C)} = \frac{P(\{6\})}{P(C)} = \frac{\frac{1}{6}}{\frac{1}{3}} = \frac{1}{2} = 0.5.$$

□

## 2 (3 points)

On 22 March 2024, ICA Maxi Linköping sells (royal gala) apples. Assume that on this day the distribution  $X$  of weights (in gram) of all such apples in ICA Maxi Linköping is normal  $X \sim N(150, 16)$ . That is, a randomly selected such apple from ICA Maxi Linköping on 22 March 2024 has a weight according to the distribution  $X \sim N(150, 16)$ .

(2.1) (1p) If one buys one such apple, and let  $X_1$  denote the weight. Find the probability that the weight of this apple is more than 162 grams.

(2.2) (2p) If one buys 9 such apples, and let  $X_1, \dots, X_9$  denote the weights. Find the probability that the total weight of these 9 apples is less than 1380 grams.

*Solution.* (2.1)

$$P(X_1 > 162) = P\left(\frac{X_1 - \mu}{\sigma} > \frac{162 - 150}{4}\right) = P(N(0, 1) > 3) = 1 - P(N(0, 1) \leq 3) = 1 - \Phi(3) = 1 - 0.9987 = 0.0013.$$

(2.2)

$$\begin{aligned} P(X_1 + \dots + X_9 < 1380) &= P\left(\frac{X_1 + \dots + X_9}{9} < 1380/9\right) = P(\bar{X} < 1380/9) \\ &= P\left(\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} < \frac{1380/9 - 150}{4/\sqrt{9}}\right) = P(N(0, 1) < 2.5) = \Phi(2.5) = 0.9938. \end{aligned}$$

□

### 3 (3 points)

Let  $(X, Y)$  be a two dimensional random variable with the following joint probability mass function (pmf)  $p(x, y)$ :

$X \setminus Y$	-1	0	1
0	0.25	0	0.25
1	0	0.5	0

The table tells that  $X$  can take values 0 and 1, and  $Y$  can take values -1, 0 and 1.

(3.1) (1p) Find the probability  $P(X + Y \leq 0.8)$ .

(3.2) (1p) Find the marginal pmf  $p_X(x)$  of  $X$ , and the marginal pmf  $p_Y(y)$  of  $Y$ . Are  $X$  and  $Y$  independent? Why?

(3.3) (1p) Find the mean  $\mu = E(X)$  and the variance  $\sigma^2 = V(X)$  of  $X$ .

*Solution.* (3.1)

$$P(X + Y \leq 0.8) = p(0, -1) + p(0, 0) + p(1, -1) = 0.25 + 0 + 0 = 0.25.$$

(3.2) The marginal pmfs are as follows:

$X$	0	1
$p_X(x)$	0.5	0.5

$Y$	-1	0	1
$p_Y(y)$	0.25	0.5	0.25

$X$  and  $Y$  are NOT independent because  $p(0, 0) \neq p_X(0) \cdot p_Y(0)$ .

(3.3)

$$\mu = E(X) = 0 \cdot 0.5 + 1 \cdot 0.5 = 0.5,$$

$$\sigma^2 = V(X) = E(X^2) - \mu^2 = (0^2 \cdot 0.5 + 1^2 \cdot 0.5) - 0.5^2 = 0.5 - 0.25 = 0.25.$$

□

### 4 (3 points)

(4.1) (1p) A population  $X$  is continuous with a probability density function (pdf) given as

$$f(x) = \frac{1}{\theta}, \quad \text{for } 0 < x < \theta,$$

where  $\theta > 0$  is an unknown parameter. A sample  $\{x_1, x_2, \dots, x_n\}$  is taken from this population. Use the method of moments to find a point estimate  $\hat{\theta}_{MM}$  of  $\theta$ .

(4.2) (2p) Another population  $Y$  is continuous with a probability density function (pdf) given as

$$f(y) = \sqrt{\frac{2}{\pi}} \cdot \frac{y^2}{\lambda^3} \cdot e^{-\frac{y^2}{2\lambda^2}}, \quad \text{for } y > 0,$$

where  $\lambda > 0$  is an unknown parameter. A sample  $\{y_1, y_2, \dots, y_n\}$  is taken from this population. Use the maximum-likelihood method to find a point estimate  $\hat{\lambda}_{ML}$  of  $\lambda$ .

*Solution.* (4.1) For MM, we use the first equation  $E(X) = \bar{x}$ , where the population mean  $E(X)$  can be computed as

$$E(X) = \int_0^\theta x \cdot \frac{1}{\theta} dx = \frac{1}{\theta} \cdot \frac{1}{2} \theta^2 = \frac{\theta}{2}.$$

Then the first equation  $E(X) = \bar{x}$  directly gives  $\frac{\theta}{2} = \bar{x}$ , which implies  $\hat{\theta}_{MM} = 2\bar{x}$ .

(4.2) The likelihood function is

$$\begin{aligned} L(\lambda) &= f(y_1) \cdot f(y_2) \cdot \dots \cdot f(y_n) = \left[ \sqrt{\frac{2}{\pi}} \cdot \frac{y_1^2}{\lambda^3} \cdot e^{-\frac{y_1^2}{2\lambda^2}} \right] \cdot \dots \cdot \left[ \sqrt{\frac{2}{\pi}} \cdot \frac{y_n^2}{\lambda^3} \cdot e^{-\frac{y_n^2}{2\lambda^2}} \right] \\ &= \left( \sqrt{\frac{2}{\pi}} \right)^n \cdot \frac{y_1^2 \cdot \dots \cdot y_n^2}{\lambda^{3n}} \cdot e^{-\frac{y_1^2 + \dots + y_n^2}{2\lambda^2}}, \end{aligned}$$

and the log likelihood function is

$$\ln L(\lambda) = \ln \left( \sqrt{\frac{2}{\pi}} \right)^n + \ln(y_1^2 \dots y_n^2) - 3n \ln \lambda - \frac{y_1^2 + \dots + y_n^2}{2\lambda^2}.$$

Now taking derivative and setting it as zero give

$$0 = \ln' L(\lambda) = -\frac{3n}{\lambda} + \frac{y_1^2 + \dots + y_n^2}{\lambda^3} \implies \hat{\lambda}_{ML} = \sqrt{\frac{y_1^2 + \dots + y_n^2}{3n}}.$$

□

## 5 (3 points)

One wants to compare the sizes of perch ('Abborre' in Swedish) in Roxen and Glan. Suppose that the distribution  $X$  of sizes of perch in Roxen is normal  $X \sim N(\mu_1, \sigma^2)$ , and the distribution  $Y$  of sizes of perch in Glan is also normal  $Y \sim N(\mu_2, \sigma^2)$ . We further assume that  $X$  and  $Y$  are independent. A sample of perch from Roxen contains 8 perch with sample mean 16.5 cm, and sample standard deviation 2.8 cm. Another sample of perch from Glan contains 10 perch with sample mean 14.2 cm, and sample standard deviation 2.2 cm. Do the samples provide any evidence that the perch in Roxen are bigger than the perch in Glan? Answer this question by constructing a 95% one-sided confidence interval  $I_{\mu_1 - \mu_2}$  of  $\mu_1 - \mu_2$  in the form  $I_{\mu_1 - \mu_2} = (a, +\infty)$ . (Hint: do NOT use any language of Hypotheses Testing).

*Solution.* This is the Case 1.3'. The two samples are:

$$n_1 = 8, \bar{x} = 16.5, s_x = 2.8; \quad n_2 = 10, \bar{y} = 14.2, s_y = 2.2.$$

The one-sided CI is:

$$\begin{aligned} I_{\mu_1 - \mu_2} &= ((\bar{x} - \bar{y}) - t_\alpha(n_1 + n_2 - 2) \cdot s \cdot \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}, +\infty) \\ &= ((16.5 - 14.2) - t_{0.05}(8 + 10 - 2) \cdot s \cdot \sqrt{\frac{1}{8} + \frac{1}{10}}, +\infty) \\ &= (2.3 - 1.75 \cdot 2.48 \cdot 0.4743, +\infty) = (0.242, +\infty), \end{aligned}$$

where  $s^2 = \frac{(n_1-1)s_x^2 + (n_2-1)s_y^2}{n_1+n_2-2} = 6.1525$ . As  $0.242 > 0$ , it follows that  $\mu_1 - \mu_2 > 0$ , that is  $\mu_1 > \mu_2$ . Yes, the samples do provide evidence that the perch in Roxen are bigger than the perch in Glan. □

## 6 (3 points)

A factory produces plenty of items. The CEO of this factory claims that the defective rate  $p$  of the items is less than 1%. Now a group of statisticians is employed to check whether or not the claim of the CEO is correct, based on hypotheses testing (HT). To this end, a sample of 2000 items is selected and there are 15 defective items among them. (Hint: \* Please keep as many decimals as possible during your calculation.

\* Pay attention to percentages, for example 1%=0.01, 0.75%=0.0075, 0.25%=0.0025, etc).

- (6.1) (1p) Write down appropriate hypotheses  $H_0$  and  $H_a$ .
- (6.2) (1p) With a significance level 5%, is  $H_0$  rejected? Why?
- (6.3) (1p) Find the power  $h(0.8\%)$  of the test in (6.2).

*Solution.* (6.1)

$$H_0 : p = 1\%, \quad H_a : p < 1\%.$$

(6.2) The observed value of the test statistic and the rejection region are as follows:

$$\begin{aligned} TS &= \frac{\hat{p} - p_0}{\sqrt{p_0(1-p_0)/n}} = \frac{15/2000 - 1\%}{\sqrt{1\%(1-1\%)/2000}} = -1.1237, \\ C &= (-\infty, -z_\alpha) = (-\infty, -1.64). \end{aligned}$$

Since  $TS \notin C$ , we do NOT reject  $H_0$  (which mean that the sample does not provide any evidence that  $p < 1\%$ ).

(6.3)

$$\begin{aligned}h(0.8\%) &= P(\text{reject } H_0, \text{ when } p = 0.8\%) = P(TS \in C, \text{ when } p = 0.8\%) \\&= P\left(\frac{\hat{P} - p_0}{\sqrt{p_0(1-p_0)/n}} < -1.64, \text{ when } p = 0.8\%\right) \\&= P(\hat{P} < p_0 - 1.64 \cdot \sqrt{p_0(1-p_0)/n}, \text{ when } p = 0.8\%) \\&= P(\hat{P} < 0.00635, \text{ when } p = 0.8\%) \\&= P\left(\frac{\hat{P} - p}{\sqrt{p(1-p)/n}} < \frac{0.00635 - p}{\sqrt{p(1-p)/n}}, \text{ when } p = 0.8\%\right) \\&= P(N(0,1) < \frac{0.00635 - 0.8\%}{\sqrt{0.8\%(1-0.8\%)/2000}}) \\&= P(N(0,1) < -0.83) = \Phi(-0.83) = 1 - \Phi(0.83) = 1 - 0.7967 = 0.2033.\end{aligned}$$

□

## 1. Basic probability

(1.1) Conditional probability  $P(A|B) = \frac{P(A \cap B)}{P(B)}$ .

(1.2) Total probability  $P(B) = \sum_{i=1}^k P(B|A_i)P(A_i)$  where  $\{A_i\}$  are disjoint and  $\cup_{i=1}^k A_i = S$ .

(1.3) Bayes' Theorem  $P(A_i|B) = \frac{P(B|A_i)P(A_i)}{\sum_{k=1}^k P(B|A_k)P(A_k)}$  where  $\{A_i\}$  are in (1.2).

## 2. Random variables (r.v.s)

(2.1) Discrete r.v.  $X$  has a pmf  $p(x) = P(X = x)$  satisfying  $p(x) \geq 0$  and  $\sum p(x_i) = 1$ ,

$$\begin{array}{c|cccc} X & x_1 & x_2 & \cdots & x_n & \cdots \\ \hline p(x) & p(x_1) & p(x_2) & \cdots & p(x_n) & \cdots \end{array}$$

Expectation (or *Expected value* or *mean*)  $\mu_X = E(X) = \sum x_i p(x_i)$ ;

Variance  $\sigma_X^2 = V(X) = E(X - \mu_X)^2 = E(X^2) - \mu_X^2 = \sum x_i^2 p(x_i) - (\sum x_i p(x_i))^2$ .

(2.2) Continuous r.v.  $X$  has a pdf  $f(x)$  satisfying  $f(x) \geq 0$  and  $\int_{-\infty}^{\infty} f(x) dx = 1$ ,

$$P(a < X < b) = \int_a^b f(x) dx.$$

Expectation (or *Expected value* or *mean*)  $\mu_X = E(X) = \int_{-\infty}^{\infty} x f(x) dx$ ;

Variance  $\sigma_X^2 = V(X) = E(X - \mu_X)^2 = E(X^2) - \mu_X^2 = \int_{-\infty}^{\infty} x^2 f(x) dx - (\int_{-\infty}^{\infty} x f(x) dx)^2$ .

(2.3) Cumulative distribution function (cdf) of a r.v.  $X$  is  $F(x) = P(X \leq x)$ .

(2.4)  $X$  and  $Y$  are r.v.s,  $a, b$  and  $c$  are scalars, then

$$E(aX + bY + c) = aE(X) + bE(Y) + c,$$

$$V(aX + bY + c) = a^2V(X) + b^2V(Y) + 2ab \operatorname{cov}(X, Y),$$

$$E(g(X, Y)) = \begin{cases} \sum_{i,j} g(x_i, y_j) \cdot p(x_i, y_j), & \text{for discrete } (X, Y), \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) \cdot f(x, y) dx dy, & \text{for continuous } (X, Y). \end{cases}$$

(2.5) • Discrete r.v.  $(X, Y)$  has a joint pmf  $p(x, y)$  satisfying  $p(x, y) \geq 0$  and  $\sum_{x_i} \sum_{y_j} p(x_i, y_j) = 1$ .

The *marginal pmf* of  $X$  is  $p_X(x) = \sum_y p(x, y)$ ;

The *marginal pmf* of  $Y$  is  $p_Y(y) = \sum_x p(x, y)$ ;

$X$  and  $Y$  are *independent* if  $p(x, y) = p_X(x) \cdot p_Y(y)$ .

• Continuous r.v.  $(X, Y)$  has a joint pdf  $p(x, y)$  satisfying  $f(x, y) \geq 0$  and  $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$ .

The *marginal pdf* of  $X$  is  $f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy$ ;

The *marginal pdf* of  $Y$  is  $f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx$ ;

$X$  and  $Y$  are *independent* if  $f(x, y) = f_X(x) \cdot f_Y(y)$ .

## 3. Several special r.v.s

(3.1)  $X \sim \operatorname{Bin}(n, p)$  has a pmf  $p(x) = P(X = x) = \binom{n}{x} \cdot p^x \cdot (1-p)^{n-x}$ ,  $x = 0, 1, 2, \dots, n$ .

$$E(X) = n \cdot p, \quad V(X) = n \cdot p \cdot (1-p).$$

(3.2)  $X \sim \operatorname{Po}(\lambda)$  has a pmf  $p(x) = P(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}$ ,  $x = 0, 1, 2, \dots$   
 $E(X) = \lambda, \quad V(X) = \lambda$ .

(3.3)  $X \sim \operatorname{Hypergeometric}$  has a pmf  $p(x) = P(X = x) = \frac{\binom{M}{x} \binom{N-M}{n-x}}{\binom{N}{n}}$ .

(3.4)  $X \sim \operatorname{Exp}(\lambda)$  has a pdf

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \geq 0, \\ 0, & \text{otherwise.} \end{cases}$$

(3.5)  $E(X) = \frac{1}{\lambda}, \quad V(X) = (\frac{1}{\lambda})^2$ .

$X \sim N(\mu, \sigma^2)$  has a pdf

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \quad -\infty < x < \infty.$$

(3.6)  $E(X) = \mu, \quad V(X) = \sigma^2$ .

$X \sim U(a, b)$  has a pdf

$$f(x) = \begin{cases} \frac{1}{b-a}, & a < x < b, \\ 0, & \text{otherwise.} \end{cases}$$

$$E(X) = \frac{a+b}{2}, \quad V(X) = \frac{(b-a)^2}{12}.$$

## 4. Central Limit Theorem (CLT)

Suppose that a population has mean  $= \mu$  and variance  $= \sigma^2$ . A random sample  $\{X_1, X_2, \dots, X_n\}$  from this population is given. Then for large  $n \geq 30$ ,

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0, 1). \tag{1}$$

• If the population is normal, then (1) holds for any  $n$ .

• Note that  $\mu = E(\bar{X})$  and  $(\sigma/\sqrt{n})^2 = V(\bar{X})$ .

## 5. Several notations in statistics

(5.1) Sample mean:  $\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n} = \sum \frac{X_i}{n}; \quad \bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n} = \sum \frac{x_i}{n}$ .

(5.2) Sample variance:

$$S^2 = \frac{\sum (X_i - \bar{X})^2}{n-1} = \frac{1}{n-1} \left( \sum X_i^2 - \frac{(\sum X_i)^2}{n} \right); \quad s^2 = \frac{\sum (x_i - \bar{x})^2}{n-1} = \frac{1}{n-1} \left( \sum x_i^2 - \frac{(\sum x_i)^2}{n} \right).$$

• Capital letters  $\bar{X}$  and  $S^2$  refer to the objects based on random sample (therefore they are in general r.v.s), while small letters  $\bar{x}$  and  $s^2$  are the objects based on observations (so they are scalars).

(5.3) A point estimator of  $\theta$  obtained by Methods of Moments is denoted as  $\hat{\theta}_{MM}$ .

(5.4) A point estimator of  $\theta$  obtained by Maximum Likelihood method is denoted as  $\hat{\theta}_{ML}$ .

## 6. Confidence Interval (CI)

In this course, three types of confidence intervals are studied depending on the unknown population parameter(s): CI-1 (confidence intervals for population mean(s)), CI-2 (confidence intervals for population variance(s)), and CI-3 (confidence intervals for population proportion(s)).

**CI-1: (1 - α) CI of a population mean μ**

**case 1.1 (any n)** If population  $X \sim N(\mu, \sigma^2)$  and  $\sigma^2$  is known, then  $\frac{\bar{X}-\mu}{\sigma/\sqrt{n}} \sim N(0, 1)$  and

$$I_\mu = (\bar{x} - z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}, \bar{x} + z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}) := \bar{x} \mp z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}.$$

**case 1.2 (n ≥ 30)** For any population X, it holds that  $\frac{\bar{X}-\mu}{\sigma/\sqrt{n}} \sim N(0, 1)$  and

$$I_\mu = \bar{x} \mp z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}} \text{ or } I_\mu = \bar{x} \mp z_{\alpha/2} \cdot \frac{\hat{\sigma}}{\sqrt{n}}.$$

**case 1.3 (any n)** If population  $X \sim N(\mu, \sigma^2)$  and  $\sigma^2$  is unknown, then  $\frac{\bar{X}-\mu}{S/\sqrt{n}} \sim T(n-1)$  and

$$I_\mu = \bar{x} \mp t_{\alpha/2}(n-1) \cdot \frac{s}{\sqrt{n}}.$$

**CI-1': (1 - α) CI of the difference of two population means  $\mu_X - \mu_Y$**

**case 1.1' (any  $n_1, n_2$ )** If independent populations  $X \sim N(\mu_X, \sigma_X^2)$ ,  $Y \sim N(\mu_Y, \sigma_Y^2)$ , and  $\sigma_X^2, \sigma_Y^2$  are known, then

$$\frac{(\bar{X}-\bar{Y})-(\mu_X-\mu_Y)}{\sqrt{\frac{\sigma_X^2}{n_1} + \frac{\sigma_Y^2}{n_2}}} \sim N(0, 1), \text{ and } I_{\mu_X-\mu_Y} = (\bar{x}-\bar{y}) \mp z_{\alpha/2} \cdot \sqrt{\frac{\sigma_X^2}{n_1} + \frac{\sigma_Y^2}{n_2}}.$$

**case 1.2' ( $n_1, n_2 \geq 30$ )** For any independent populations X and Y, it holds that

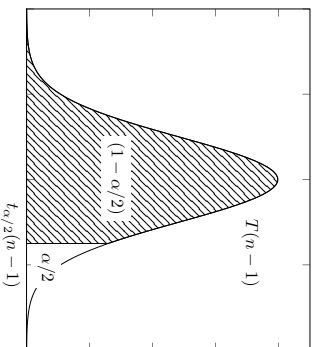
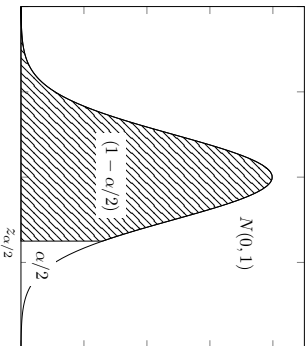
$$\frac{(\bar{X}-\bar{Y})-(\mu_X-\mu_Y)}{\sqrt{\frac{\sigma_X^2}{n_1} + \frac{\sigma_Y^2}{n_2}}} \sim N(0, 1) \text{ and}$$

$$I_{\mu_X-\mu_Y} = (\bar{x}-\bar{y}) \mp z_{\alpha/2} \cdot \sqrt{\frac{\sigma_X^2}{n_1} + \frac{\sigma_Y^2}{n_2}} \text{ or } I_{\mu_X-\mu_Y} = (\bar{x}-\bar{y}) \mp z_{\alpha/2} \cdot \sqrt{\frac{\hat{\sigma}_X^2}{n_1} + \frac{\hat{\sigma}_Y^2}{n_2}}.$$

**case 1.3' (any  $n_1, n_2$ )** If independent populations  $X \sim N(\mu_X, \sigma_X^2)$ ,  $Y \sim N(\mu_Y, \sigma_Y^2)$ , where  $\sigma_X^2, \sigma_Y^2$  are unknown but  $\sigma_X^2 = \sigma_Y^2$ , then

$$\frac{(\bar{X}-\bar{Y})-(\mu_X-\mu_Y)}{S\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \sim T(n_1+n_2-2), \text{ where } S^2 = \frac{(n_1-1)S_X^2 + (n_2-1)S_Y^2}{n_1+n_2-2}, \text{ and}$$

$$I_{\mu_X-\mu_Y} = (\bar{x}-\bar{y}) \mp t_{\alpha/2}(n_1+n_2-2) \cdot s\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}.$$



**CI-2: (1 - α) CI of population variance(s)  $\sigma^2$**

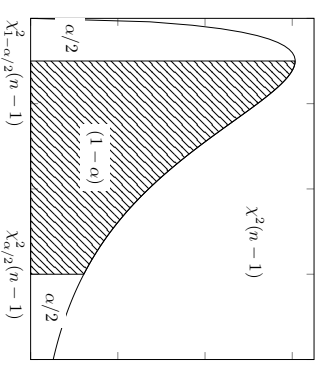
• If a population  $X \sim N(\mu, \sigma^2)$  and  $\sigma^2$  is unknown, then  $\frac{(n-1)S^2}{\sigma^2} \sim \chi^2(n-1)$ , and

$$I_{\sigma^2} = \left( \frac{(n-1)s^2}{\chi_{\alpha/2}^2(n-1)}, \frac{(n-1)s^2}{\chi_{1-\alpha/2}^2(n-1)} \right).$$

• If two independent populations  $X \sim N(\mu_X, \sigma^2)$  and  $Y \sim N(\mu_Y, \sigma^2)$ , and  $\sigma^2$  is unknown, then  $\frac{(n_1+n_2-2)S^2}{\sigma^2} \sim \chi^2(n_1+n_2-2)$ , and

$$I_{\sigma^2} = \left( \frac{(n_1+n_2-2)s^2}{\chi_{\alpha/2}^2(n_1+n_2-2)}, \frac{(n_1+n_2-2)s^2}{\chi_{1-\alpha/2}^2(n_1+n_2-2)} \right),$$

where  $S^2 = \frac{(n_1-1)S_X^2 + (n_2-1)S_Y^2}{n_1+n_2-2}$ .



**CI-3: (1 - α) CI of population proportion(s)**

• If a (large) population has an unknown proportion p, then  $\frac{\hat{p}-p}{\sqrt{p(1-p)/n}} \sim N(0, 1)$  if  $n\hat{p} \geq 10$  and  $n(1-\hat{p}) \geq 10$  with  $\hat{p} = x/n$ , and  $I_p = \hat{p} \mp z_{\alpha/2} \cdot \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$ .

• If two independent (large) populations have unknown proportions  $p_1$  and  $p_2$ , then

$$\frac{(\hat{p}_1-\hat{p}_2)-(p_1-p_2)}{\sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}} \sim N(0, 1)$$

if  $n_i\hat{p}_i \geq 10$  and  $n_i(1-\hat{p}_i) \geq 10$  for  $i = 1, 2$ , and  $I_{p_1-p_2} = (\hat{p}_1-\hat{p}_2) \mp z_{\alpha/2} \cdot \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$ .

**7. Hypothesis Test (HT)**

	$H_0$ is true	$H_0$ is false and $\theta = \theta_1$
reject $H_0$	(type I error or significance level) $\alpha$	(power) $h(\theta_1)$
don't reject $H_0$	$1 - \alpha$	(type II error) $\beta(\theta_1) = 1 - h(\theta_1)$

reject  $H_0 \Leftrightarrow TS \in C \Leftrightarrow p\text{-value} < \alpha$

**$\chi^2$  tests for populations (non-parametric tests)**

Suppose that for a random sample of a population X, the n elements of it are classified into k disjoint groups  $A_i, 1 \leq i \leq k$ . For each group  $A_i, 1 \leq i \leq k$ , suppose that there are  $N_i, 1 \leq i \leq k$  elements inside. Let  $p_i = P(A_i)$  assuming a given distribution of X. Note that  $p_1 + p_2 + \dots + p_k = 1$  and  $N_1 + N_2 + \dots + N_k = n$ . One wants to test the hypotheses

$$H_0 : P(A_i) = p_i, \quad 1 \leq i \leq k, \quad H_a : P(A_i) \neq p_i \text{ for some } 1 \leq i \leq k.$$

If  $n$  is large in the sense that  $np_i \geq 5$  for all  $1 \leq i \leq k$ , then the test statistic is

$$\sum_{i=1}^k \frac{(N_i - np_i)^2}{np_i} \approx \chi^2(k-1).$$

Therefore the observation of the test statistic is

$$TS = \sum_{i=1}^k \frac{(n_i - np_i)^2}{np_i}, \text{ where } n_i \text{ is the observation of } N_i, 1 \leq i \leq k.$$

For the critical region  $C$ , one can take (note that if  $H_0$  is true, then  $TS$  should be close to zero)

$$C = (\chi^2_{\alpha}(k-1), \infty).$$

The conclusion would be  $TS \in C \iff H_0$  is rejected.

## 8. Linear and logistic regression

**(Multiple) linear regression:**  $Y = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k + \varepsilon$ ,  $\varepsilon \sim N(0, \sigma^2)$ .

- $Y$  : response variable (which is normal r.v.),  $\{x_1, \dots, x_k\}$  : predictors (which are scalars).
- sample:  $\{(x_{11}, \dots, x_{1k}; y_1), (x_{21}, \dots, x_{2k}; y_2), \dots, (x_{n1}, \dots, x_{nk}; y_n)\}$ .
- how to estimate  $\beta_j \approx \hat{\beta}_j$  : least square method, that is, to minimize  $\sum_{i=1}^n (y_i - \hat{y}_i)^2$ , where the estimated (multiple) linear regression line  $\hat{y}$  is

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x_1 + \dots + \hat{\beta}_k x_k.$$

- $\frac{\hat{\beta}_j - \beta_j}{se(\hat{\beta}_j)} \sim T(n-k-1)$ , this helps determine whether or not the real  $\beta_j = 0$ ?
- $\sigma^2 \approx \frac{SSE}{n-k-1}$ , this gives an estimation of the size of the error.
- $R^2 = \frac{SSR}{SSY}$ , this gives how well the model is (if  $R^2 \approx 1$ , then the model fits the sample very well).
- How to test  $\beta_1 = \dots = \beta_k = 0$  ? Use the random variable  $\frac{SSR/k}{SSE/(n-k-1)} \sim F(k, n-k-1)$ .

**Logistic regression:** Let  $Y$  can only take 0 or 1 with  $P(Y=1) = p$  and  $P(Y=0) = 1-p$ .

$$E(Y) = p(x_1, \dots, x_k) = \frac{e^{\beta_0 + \beta_1 x_1 + \dots + \beta_k x_k}}{1 + e^{\beta_0 + \beta_1 x_1 + \dots + \beta_k x_k}}.$$

- $Y$  : response variable (which is Bernoulli r.v.  $P(Y=1) = p$  and  $P(Y=0) = 1-p$ , so  $E(Y) = p$ ),  $\{x_1, \dots, x_k\}$  : predictors (which are scalars).
- sample:  $\{(x_{11}, \dots, x_{1k}; y_1), (x_{21}, \dots, x_{2k}; y_2), \dots, (x_{n1}, \dots, x_{nk}; y_n)\}$ .
- how to estimate  $\beta_j \approx \hat{\beta}_j$  : maximal likelihood method (maximize  $\prod_{i=1}^n p(x_{i1}, \dots, x_{ik})^{y_i} (1 - p(x_{i1}, \dots, x_{ik}))^{1-y_i}$ ).
- $\frac{\hat{\beta}_j - \beta_j}{se(\hat{\beta}_j)} \approx N(0, 1)$  for large  $n \geq 30$ , this helps determine whether or not the real  $\beta_j = 0$ ?
- Classification of a new object  $Y(x_1, \dots, x_k)$  as 1 or 0 according

$$Y(x_1, \dots, x_k) = \begin{cases} 1, & \text{if } \hat{p}(x_1, \dots, x_k) \geq 0.5, \\ 0, & \text{if } \hat{p}(x_1, \dots, x_k) < 0.5, \end{cases}$$

where the estimated logit function  $\hat{p}(x_1, \dots, x_k)$  is

$$\hat{p}(x_1, \dots, x_k) = \frac{e^{\hat{\beta}_0 + \hat{\beta}_1 x_1 + \dots + \hat{\beta}_k x_k}}{1 + e^{\hat{\beta}_0 + \hat{\beta}_1 x_1 + \dots + \hat{\beta}_k x_k}}.$$

## 9. Tables

(9.1) Table for  $N(0, 1)$  standard normal random variable  $\Phi(x) = P(N(0, 1) \leq x)$ ,  $x \geq 0$ .  
There is an important relation  $\Phi(-x) = 1 - \Phi(x)$ ,  $x \geq 0$ .

x	0	1	2	3	4	5	6	7	8	9
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9564	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0.9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986
3.0	0.9987	0.9987	0.9987	0.9988	0.9988	0.9989	0.9989	0.9989	0.9990	0.9990
3.1	0.9990	0.9991	0.9991	0.9991	0.9992	0.9992	0.9992	0.9993	0.9993	0.9993
3.2	0.9993	0.9993	0.9994	0.9994	0.9994	0.9994	0.9994	0.9995	0.9995	0.9995
3.3	0.9995	0.9995	0.9995	0.9996	0.9996	0.9996	0.9996	0.9996	0.9997	0.9997
3.4	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997	0.9998	0.9998	0.9998
3.5	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998
3.6	0.9998	0.9998	0.9998	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999
3.7	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999
3.8	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999
3.9	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
4.0	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000

(9.2) Table for  $T(f)$  random variable  $F(x) = P(T(f) \leq x)$ ,  
where  $f$  is a parameter called 'degrees of freedom'.

f	0.75	0.90	0.95	0.975	0.99	0.995	0.9975	0.9995
1	1.00	3.08	6.31	12.71	31.82	63.66	127.32	636.62
2	0.82	1.89	2.92	4.30	6.96	9.92	14.09	31.60
3	0.76	1.64	2.35	3.18	4.54	5.84	7.45	12.92
4	0.74	1.53	2.13	2.78	3.75	4.60	5.60	8.61
5	0.73	1.48	2.02	2.57	3.36	4.03	4.77	6.87
6	0.72	1.44	1.94	2.45	3.14	3.71	4.32	5.96
7	0.71	1.41	1.89	2.36	3.00	3.50	4.03	5.41
8	0.71	1.40	1.86	2.31	2.90	3.36	3.83	5.04
9	0.70	1.38	1.83	2.26	2.82	3.25	3.69	4.78
10	0.70	1.37	1.81	2.23	2.76	3.17	3.58	4.59
11	0.70	1.36	1.80	2.20	2.72	3.11	3.50	4.44
12	0.70	1.36	1.78	2.18	2.68	3.05	3.43	4.32
13	0.69	1.35	1.77	2.16	2.65	3.01	3.37	4.22
14	0.69	1.35	1.76	2.14	2.62	2.98	3.33	4.14
15	0.69	1.34	1.75	2.13	2.60	2.95	3.29	4.07
16	0.69	1.34	1.75	2.12	2.58	2.92	3.25	4.01
17	0.69	1.33	1.74	2.11	2.57	2.90	3.22	3.97
18	0.69	1.33	1.73	2.10	2.55	2.88	3.20	3.92
19	0.69	1.33	1.73	2.09	2.54	2.86	3.17	3.88
20	0.69	1.33	1.72	2.09	2.53	2.85	3.15	3.85
21	0.69	1.32	1.72	2.08	2.52	2.83	3.14	3.82
22	0.69	1.32	1.72	2.07	2.51	2.82	3.12	3.79
23	0.69	1.32	1.71	2.07	2.50	2.81	3.10	3.77
24	0.68	1.32	1.71	2.06	2.49	2.80	3.09	3.75
25	0.68	1.32	1.71	2.06	2.49	2.79	3.08	3.73
26	0.68	1.31	1.71	2.06	2.48	2.78	3.07	3.71
27	0.68	1.31	1.70	2.05	2.47	2.77	3.06	3.69
28	0.68	1.31	1.70	2.05	2.47	2.76	3.05	3.67
29	0.68	1.31	1.70	2.05	2.46	2.76	3.04	3.66
30	0.68	1.31	1.70	2.04	2.46	2.75	3.03	3.65
40	0.68	1.30	1.68	2.02	2.42	2.70	2.97	3.55
50	0.68	1.30	1.68	2.01	2.40	2.68	2.94	3.50
60	0.68	1.30	1.67	2.00	2.39	2.66	2.91	3.46
100	0.68	1.29	1.66	1.98	2.36	2.63	2.87	3.39
$\infty$	0.67	1.28	1.65	1.96	2.33	2.58	2.81	3.29



(9.3) Table for  $\chi^2(f)$  random variable  $F(x) = P(\chi^2(f) \leq x)$ , where  $f$  is a parameter.

$f$	0.0005	0.001	0.005	0.01	0.025	0.05	0.10	0.20	0.30	0.40	0.50
1	0.00	0.00	0.00	0.00	0.00	0.00	0.02	0.06	0.15	0.27	0.45
2	0.00	0.00	0.01	0.02	0.05	0.10	0.21	0.45	0.71	1.02	1.39
3	0.02	0.02	0.07	0.11	0.22	0.35	0.58	1.01	1.42	1.87	2.37
4	0.06	0.09	0.21	0.30	0.48	0.71	1.06	1.65	2.19	2.75	3.36
5	0.16	0.21	0.41	0.55	0.83	1.15	1.61	2.34	3.00	3.66	4.35
6	0.30	0.38	0.68	0.87	1.24	1.64	2.20	3.07	3.83	4.57	5.35
7	0.48	0.60	0.99	1.24	1.69	2.17	2.73	3.62	4.67	5.49	6.35
8	0.71	0.86	1.34	1.65	2.18	2.73	3.49	4.59	5.53	6.42	7.34
9	0.97	1.15	1.73	2.09	2.70	3.33	4.17	5.38	6.39	7.36	8.34
10	1.26	1.48	2.16	2.56	3.25	3.94	4.87	6.18	7.27	8.30	9.34
11	1.59	1.83	2.60	3.05	3.82	4.57	5.58	6.99	8.15	9.24	10.34
12	1.93	2.21	3.07	3.57	4.40	5.23	6.30	7.81	9.03	10.18	11.34
13	2.31	2.62	3.57	4.11	5.01	5.89	7.04	8.63	9.93	11.13	12.34
14	2.70	3.04	4.07	4.66	5.63	6.57	7.79	9.47	10.82	12.08	13.34
15	3.11	3.48	4.60	5.23	6.26	7.26	8.55	10.31	11.72	13.03	14.34
16	3.54	3.94	5.14	5.81	6.91	7.96	9.31	11.15	12.62	13.98	15.34
17	3.98	4.42	5.70	6.41	7.56	8.67	10.09	12.00	13.53	14.94	16.34
18	4.44	4.90	6.26	7.01	8.23	9.39	10.86	12.86	14.44	15.89	17.34
19	4.91	5.41	6.84	7.63	8.91	10.12	11.65	13.72	15.35	16.85	18.34
20	5.40	5.92	7.43	8.26	9.59	10.85	12.44	14.58	16.27	17.81	19.34
21	5.90	6.45	8.03	8.90	10.28	11.59	13.24	15.44	17.18	18.77	20.34
22	6.40	6.98	8.64	9.54	10.98	12.34	14.04	16.31	18.10	19.73	21.34
23	6.92	7.53	9.26	10.20	11.69	13.09	14.85	17.19	19.02	20.69	22.34
24	7.45	8.08	9.89	10.86	12.40	13.85	15.66	18.06	19.94	21.65	23.34
25	7.99	8.65	10.52	11.52	13.12	14.61	16.47	18.94	20.87	22.62	24.34
26	8.54	9.22	11.16	12.20	13.84	15.38	17.29	19.82	21.79	23.58	25.34
27	9.09	9.80	11.81	12.88	14.57	16.15	18.11	20.70	22.72	24.54	26.34
28	9.66	10.39	12.46	13.56	15.31	16.93	18.94	21.59	23.65	25.51	27.34
29	10.23	10.99	13.12	14.26	16.05	17.71	19.77	22.48	24.58	26.48	28.34
30	10.80	11.59	13.79	14.95	16.79	18.49	20.60	23.36	25.51	27.44	29.34
40	16.91	17.92	20.71	22.16	24.43	26.51	29.05	32.34	34.87	37.13	39.34
50	23.46	24.67	27.99	29.71	32.36	34.76	37.69	41.45	46.86	49.33	51.66
60	30.34	31.74	35.53	37.48	40.48	43.19	46.46	50.64	53.81	56.62	59.33
100	59.90	61.92	67.33	70.06	74.22	77.93	82.36	87.95	92.13	95.81	99.33

Table for  $\chi^2(f)$  random variable  $F(x) = P(\chi^2(f) \leq x)$ , where  $f$  is a parameter.

$f$	0.60	0.70	0.80	0.90	0.95	0.975	0.99	0.995	0.999	0.9995
1	0.71	1.07	1.64	2.71	3.84	5.02	6.63	7.88	10.83	12.12
2	1.83	2.41	3.22	4.61	5.99	7.38	9.21	10.60	13.82	15.20
3	2.95	3.66	4.64	6.25	7.81	9.35	11.34	12.84	16.27	17.73
4	4.04	4.88	5.99	7.78	9.49	11.14	13.28	14.86	18.47	20.00
5	5.13	6.06	7.29	9.24	11.07	12.83	15.09	16.75	20.52	22.11
6	6.21	7.23	8.56	10.64	12.59	14.45	16.81	18.55	22.46	24.10
7	7.28	8.38	9.80	12.02	14.07	16.01	18.48	20.28	24.32	26.02
8	8.35	9.52	11.03	13.36	15.51	17.53	20.09	21.95	26.12	27.87
9	9.41	10.66	12.24	14.68	16.92	19.02	21.67	23.59	27.88	29.67
10	10.47	11.78	13.44	15.99	18.31	20.48	23.21	25.19	29.59	31.42
11	11.53	12.90	14.63	17.28	19.68	21.92	24.72	26.76	31.26	33.14
12	12.58	14.01	15.81	18.55	21.03	23.34	26.22	28.30	32.91	34.82
13	13.64	15.12	16.98	19.81	22.36	24.74	27.69	29.82	34.53	36.48
14	14.69	16.22	18.15	21.06	23.68	26.12	29.14	31.32	36.12	38.11
15	15.73	17.32	19.31	22.31	25.00	27.49	30.58	32.80	37.70	39.72
16	16.78	18.42	20.47	23.54	26.30	28.85	32.00	34.27	39.25	41.31
17	17.82	19.51	21.61	24.77	27.59	30.19	33.41	35.72	40.79	42.88
18	18.87	20.60	22.76	25.99	28.87	31.53	34.81	37.16	42.31	44.43
19	19.91	21.69	23.90	27.20	30.14	32.85	36.19	38.58	43.82	45.97
20	20.95	22.77	25.04	28.41	31.41	34.17	37.57	40.00	45.31	47.50
21	21.99	23.86	26.17	29.62	32.67	35.48	38.93	41.40	46.80	49.01
22	23.03	24.94	27.30	30.81	33.92	36.78	40.29	42.80	48.27	50.51
23	24.07	26.02	28.43	32.01	35.17	38.08	41.64	44.18	49.73	52.00
24	25.11	27.10	29.55	33.20	36.42	39.36	42.98	45.56	51.18	53.48
25	26.14	28.17	30.68	34.38	37.65	40.65	44.31	46.93	52.62	54.95
26	27.18	29.25	31.79	35.56	38.89	41.92	45.64	48.29	54.05	56.41
27	28.21	30.32	32.91	36.74	40.11	43.19	46.96	49.64	55.48	57.86
28	29.25	31.39	34.03	37.92	41.34	44.46	48.28	50.99	56.89	59.30
29	30.28	32.46	35.14	39.09	42.56	45.72	49.59	52.34	58.30	60.73
30	31.32	33.53	36.25	40.26	43.77	46.98	50.89	53.67	59.70	62.16
40	41.62	44.16	47.27	51.81	55.76	59.34	63.69	66.77	73.40	76.09
50	51.89	54.72	58.16	63.17	67.50	71.42	79.15	79.49	86.66	89.56
60	62.13	65.23	68.97	74.40	79.08	83.30	88.38	91.95	99.61	102.69
100	102.95	106.91	111.67	118.50	124.34	129.56	135.81	140.17	149.45	153.17

(9.4) Table for Binomial random variable  $P(Bin(n, p) \leq k)$  if  $p \leq 0.5$ .  
 If  $p > 0.5$ , then  $P(Bin(n, p) \leq k) = P(Bin(n, 1 - p) \geq n - k)$ .

$n$	$k$	0.05	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.45	0.50
2	0	0.9025	0.8100	0.7225	0.6400	0.5625	0.4900	0.4225	0.3600	0.3025	0.2500
	1	0.9975	0.9900	0.9775	0.9600	0.9375	0.9100	0.8775	0.8400	0.7975	0.7500
3	0	0.8574	0.7290	0.6141	0.5120	0.4219	0.3430	0.2746	0.2160	0.1664	0.1250
	1	0.9928	0.9720	0.9392	0.8960	0.8438	0.7840	0.7183	0.6480	0.5747	0.5000
4	0	0.8145	0.6561	0.5220	0.4096	0.3164	0.2401	0.1785	0.1256	0.0915	0.0625
	1	0.9860	0.9477	0.8905	0.8192	0.7383	0.6517	0.5630	0.4735	0.3910	0.3125
5	0	0.7738	0.5905	0.4437	0.3277	0.2373	0.1681	0.1160	0.0778	0.0503	0.0313
	1	0.9774	0.9185	0.8352	0.7373	0.6328	0.5282	0.4284	0.3370	0.2562	0.1875
6	0	0.7351	0.5314	0.3771	0.2621	0.1780	0.1176	0.0754	0.0467	0.0277	0.0156
	1	0.9672	0.8847	0.7765	0.6554	0.5339	0.4202	0.3191	0.2333	0.1636	0.1094
7	0	0.6983	0.4783	0.3206	0.2097	0.1335	0.0824	0.0490	0.0280	0.0152	0.0078
	1	0.9566	0.8503	0.7166	0.5767	0.4449	0.3294	0.2338	0.1586	0.1024	0.0625
8	0	0.6634	0.4305	0.2725	0.1678	0.1001	0.0576	0.0319	0.0168	0.0084	0.0039
	1	0.9428	0.8131	0.6572	0.5033	0.3671	0.2553	0.1691	0.1064	0.0632	0.0352
9	0	0.6302	0.3874	0.2316	0.1342	0.0751	0.0404	0.0207	0.0101	0.0046	0.0020
	1	0.9288	0.7748	0.5995	0.4362	0.3003	0.1960	0.1211	0.0705	0.0385	0.0195

Table for Binomial random variable  $P(Bin(n, p) \leq k)$  if  $p \leq 0.5$ .  
 If  $p > 0.5$ , then  $P(Bin(n, p) \leq k) = P(Bin(n, 1 - p) \geq n - k)$ .

$n$	$k$	0.05	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.45	0.50
10	0	0.5987	0.3487	0.1969	0.1074	0.0563	0.0282	0.0135	0.0060	0.0025	0.0010
	1	0.9139	0.7361	0.5443	0.3758	0.2440	0.1493	0.0860	0.0464	0.0233	0.0107
11	0	0.5688	0.3138	0.1673	0.0859	0.0422	0.0198	0.0088	0.0036	0.0014	0.0005
	1	0.8981	0.6974	0.4922	0.3221	0.1971	0.1130	0.0606	0.0302	0.0139	0.0059
12	0	0.5404	0.2824	0.1422	0.0687	0.0317	0.0138	0.0057	0.0022	0.0008	0.0002
	1	0.8816	0.6590	0.4435	0.2749	0.1584	0.0850	0.0424	0.0196	0.0083	0.0032

Table for Binomial random variable  $P(\text{Bin}(n, p) \leq k)$  if  $p \leq 0.5$ .  
 If  $p > 0.5$ , then  $P(\text{Bin}(n, p) \leq k) = P(\text{Bin}(n, 1 - p) \geq n - k)$ .

$n$	$k$	$p$																				
		0.05	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.45	0.50	0.05	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.45	0.50	
14	0	0.4877	0.2288	0.1028	0.0440	0.0178	0.0068	0.0024	0.0008	0.0002	0.0001	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
	1	0.8470	0.5846	0.3567	0.2079	0.1010	0.0475	0.0205	0.0081	0.0029	0.0009	0.0005	0.0003	0.0002	0.0001	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
	2	0.9699	0.8416	0.6479	0.4481	0.2811	0.1608	0.0839	0.0398	0.0170	0.0065	0.0035	0.0022	0.0013	0.0007	0.0004	0.0002	0.0001	0.0000	0.0000	0.0000	0.0000
	3	0.9958	0.9559	0.8535	0.6982	0.5213	0.3552	0.2205	0.1243	0.0632	0.0287	0.0137	0.0072	0.0041	0.0024	0.0014	0.0008	0.0005	0.0003	0.0002	0.0001	0.0000
	4	0.9996	0.9908	0.9533	0.8702	0.7415	0.5842	0.4227	0.2793	0.1672	0.0898	0.0467	0.0247	0.0132	0.0072	0.0041	0.0024	0.0014	0.0008	0.0005	0.0003	0.0002
	5	1.0000	0.9985	0.9885	0.9561	0.8883	0.7805	0.6405	0.4859	0.3373	0.2120	0.1243	0.0727	0.0417	0.0247	0.0132	0.0072	0.0041	0.0024	0.0014	0.0008	0.0005
	6	1.0000	0.9998	0.9978	0.9884	0.9617	0.9067	0.8164	0.6925	0.5461	0.3953	0.2525	0.1518	0.0906	0.0519	0.0309	0.0163	0.0081	0.0041	0.0021	0.0011	0.0005
	7	1.0000	1.0000	0.9997	0.9976	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897	0.9897
	8	1.0000	1.0000	1.0000	0.9996	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976
	9	1.0000	1.0000	1.0000	1.0000	1.0000	0.9997	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976	0.9976
	10	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998	0.9998
	11	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
	12	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
	13	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
14	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	

Table for Binomial random variable  $P(\text{Bin}(n, p) \leq k)$  if  $p \leq 0.5$ .  
 If  $p > 0.5$ , then  $P(\text{Bin}(n, p) \leq k) = P(\text{Bin}(n, 1 - p) \geq n - k)$ .

$n$	$k$	$p$																				
		0.05	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.45	0.50	0.05	0.10	0.15	0.20	0.25	0.30	0.35	0.40	0.45	0.50	
17	0	0.4181	0.1668	0.0631	0.0225	0.0075	0.0023	0.0007	0.0002	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
	1	0.7922	0.4818	0.2525	0.1182	0.0501	0.0193	0.0067	0.0021	0.0006	0.0001	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
	2	0.9497	0.7618	0.5198	0.3096	0.1637	0.0774	0.0327	0.0123	0.0041	0.0014	0.0005	0.0002	0.0001	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
	3	0.9912	0.9174	0.7556	0.5489	0.3530	0.2019	0.1028	0.0464	0.0184	0.0064	0.0024	0.0011	0.0005	0.0002	0.0001	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
	4	0.9988	0.9779	0.9103	0.7582	0.5739	0.3887	0.2348	0.1260	0.0596	0.0245	0.0104	0.0045	0.0021	0.0011	0.0005	0.0002	0.0001	0.0000	0.0000	0.0000	0.0000
	5	0.9999	0.9953	0.9681	0.8943	0.7653	0.5668	0.4197	0.2639	0.1471	0.0717	0.0345	0.0165	0.0072	0.0033	0.0016	0.0007	0.0003	0.0001	0.0000	0.0000	0.0000
	6	1.0000	0.9992	0.9922	0.9623	0.8929	0.7752	0.6188	0.4478	0.2902	0.1622	0.0875	0.0445	0.0225	0.0112	0.0055	0.0026	0.0012	0.0006	0.0003	0.0001	0.0000
	7	1.0000	0.9999	0.9983	0.9891	0.9823	0.9752	0.9682	0.9617	0.9552	0.9487	0.9422	0.9357	0.9292	0.9227	0.9162	0.9097	0.9032	0.8967	0.8902	0.8837	0.8772
	8	1.0000	1.0000	0.9997	0.9974	0.9876	0.9797	0.9732	0.9667	0.9602	0.9537	0.9472	0.9407	0.9342	0.9277	0.9212	0.9147	0.9082	0.9017	0.8952	0.8887	0.8822
	9	1.0000	1.0000	1.0000	0.9995	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969	0.9969
	10	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999	0.9999
	11	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
	12	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
	13	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
	14	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
	15	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000

(9.5) Table for Poisson random variable  $P(Po(\mu) \leq k)$ .

$k$	$\mu$									
	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0
0	0.9048	0.8187	0.7408	0.6703	0.6065	0.5488	0.4966	0.4493	0.4066	0.3679
1	0.9953	0.9825	0.9631	0.9384	0.9098	0.8781	0.8442	0.8088	0.7725	0.7358
2	0.9998	0.9989	0.9964	0.9921	0.9856	0.9769	0.9659	0.9526	0.9371	0.9197
3	1.0000	0.9999	0.9997	0.9992	0.9982	0.9966	0.9942	0.9909	0.9865	0.9810
4	1.0000	1.0000	1.0000	0.9999	0.9998	0.9996	0.9992	0.9986	0.9977	0.9963
5	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9998	0.9997	0.9994
6	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999
7	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
$k$	$\mu$									
	1.1	1.2	1.3	1.4	1.5	1.6	1.7	1.8	1.9	2.0
0	0.3329	0.3012	0.2725	0.2466	0.2231	0.2019	0.1827	0.1653	0.1496	0.1353
1	0.6990	0.6626	0.6268	0.5918	0.5578	0.5249	0.4932	0.4628	0.4337	0.4060
2	0.9004	0.8795	0.8571	0.8335	0.8088	0.7834	0.7572	0.7306	0.7037	0.6767
3	0.9743	0.9662	0.9569	0.9463	0.9344	0.9212	0.9068	0.8913	0.8747	0.8571
4	0.9946	0.9923	0.9893	0.9857	0.9814	0.9763	0.9704	0.9636	0.9559	0.9473
5	0.9990	0.9985	0.9978	0.9968	0.9955	0.9940	0.9920	0.9896	0.9868	0.9834
6	0.9999	0.9997	0.9996	0.9994	0.9991	0.9987	0.9981	0.9974	0.9966	0.9955
7	1.0000	1.0000	0.9999	0.9999	0.9998	0.9997	0.9996	0.9994	0.9992	0.9989
8	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9999	0.9998
9	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
$k$	$\mu$									
	2.1	2.2	2.3	2.4	2.5	2.6	2.7	2.8	2.9	3.0
0	0.1225	0.1108	0.1003	0.0907	0.0821	0.0743	0.0672	0.0608	0.0550	0.0498
1	0.3796	0.3546	0.3309	0.3084	0.2873	0.2674	0.2487	0.2311	0.2146	0.1991
2	0.6496	0.6227	0.5960	0.5697	0.5438	0.5184	0.4936	0.4695	0.4460	0.4232
3	0.8386	0.8194	0.7993	0.7787	0.7576	0.7360	0.7141	0.6919	0.6696	0.6472
4	0.9379	0.9275	0.9162	0.9041	0.8912	0.8774	0.8629	0.8477	0.8318	0.8153
5	0.9796	0.9751	0.9700	0.9643	0.9580	0.9510	0.9433	0.9349	0.9258	0.9161
6	0.9941	0.9925	0.9906	0.9884	0.9858	0.9828	0.9794	0.9756	0.9713	0.9665
7	0.9985	0.9980	0.9974	0.9967	0.9958	0.9947	0.9934	0.9919	0.9901	0.9881
8	0.9997	0.9995	0.9994	0.9991	0.9989	0.9985	0.9981	0.9976	0.9969	0.9962
9	0.9999	0.9999	0.9999	0.9998	0.9997	0.9996	0.9995	0.9993	0.9991	0.9989
10	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9999	0.9998	0.9997	0.9996
11	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9999
12	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000

Table for Poisson random variable  $P(Po(\mu) \leq k)$ .

$k$	$\mu$									
	3.2	3.4	3.6	3.8	4.0	4.2	4.4	4.6	4.8	5.0
0	0.0408	0.0334	0.0273	0.0224	0.0183	0.0150	0.0123	0.0101	0.0082	0.0067
1	0.1712	0.1468	0.1257	0.1074	0.0916	0.0780	0.0663	0.0563	0.0477	0.0404
2	0.3799	0.3397	0.3027	0.2689	0.2381	0.2127	0.1851	0.1626	0.1425	0.1247
3	0.6025	0.5584	0.5152	0.4735	0.4335	0.3954	0.3594	0.3257	0.2942	0.2650
4	0.7806	0.7442	0.7064	0.6678	0.6288	0.5898	0.5512	0.5132	0.4763	0.4405
5	0.8946	0.8705	0.8441	0.8156	0.7851	0.7531	0.7199	0.6858	0.6510	0.6160
6	0.9534	0.9421	0.9267	0.9091	0.8893	0.8675	0.8436	0.8180	0.7908	0.7622
7	0.9832	0.9769	0.9682	0.9599	0.9489	0.9361	0.9214	0.9049	0.8867	0.8666
8	0.9943	0.9917	0.9883	0.9840	0.9786	0.9721	0.9642	0.9549	0.9442	0.9319
9	0.9982	0.9973	0.9960	0.9942	0.9919	0.9889	0.9851	0.9805	0.9749	0.9682
10	0.9995	0.9992	0.9987	0.9981	0.9972	0.9959	0.9943	0.9922	0.9896	0.9863
11	0.9999	0.9998	0.9996	0.9994	0.9991	0.9986	0.9980	0.9971	0.9960	0.9945
12	1.0000	0.9999	0.9999	0.9999	0.9998	0.9997	0.9996	0.9993	0.9990	0.9986
13	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9998	0.9997	0.9995	0.9993
14	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9999	0.9998
15	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999
$k$	$\mu$									
	5.2	5.4	5.6	5.8	6.0	6.5	7.0	7.5	8.0	8.5
0	0.0055	0.0045	0.0037	0.0030	0.0025	0.0015	0.0009	0.0006	0.0003	0.0002
1	0.0342	0.0289	0.0244	0.0206	0.0174	0.0113	0.0073	0.0047	0.0030	0.0019
2	0.1088	0.0948	0.0824	0.0715	0.0620	0.0430	0.0286	0.0203	0.0138	0.0093
3	0.2381	0.2133	0.1906	0.1700	0.1512	0.1118	0.0818	0.0591	0.0424	0.0301
4	0.4061	0.3733	0.3422	0.3127	0.2851	0.2237	0.1730	0.1321	0.0996	0.0744
5	0.5809	0.5461	0.5119	0.4783	0.4457	0.3690	0.3007	0.2414	0.1912	0.1496
6	0.7324	0.7017	0.6703	0.6384	0.6063	0.5265	0.4497	0.3782	0.3134	0.2562
7	0.8449	0.8217	0.7970	0.7710	0.7440	0.6728	0.5987	0.5246	0.4530	0.3856
8	0.9181	0.9027	0.8857	0.8672	0.8472	0.7916	0.7291	0.6620	0.5925	0.5231
9	0.9603	0.9512	0.9409	0.9292	0.9161	0.8774	0.8305	0.7764	0.7166	0.6530
10	0.9823	0.9775	0.9718	0.9651	0.9574	0.9332	0.9015	0.8622	0.8159	0.7634
11	0.9927	0.9904	0.9875	0.9841	0.9799	0.9661	0.9467	0.9208	0.8881	0.8487
12	0.9972	0.9962	0.9949	0.9932	0.9912	0.9840	0.9730	0.9573	0.9362	0.9091
13	0.9990	0.9986	0.9980	0.9973	0.9964	0.9929	0.9872	0.9784	0.9658	0.9486
14	0.9999	0.9995	0.9993	0.9990	0.9986	0.9970	0.9943	0.9897	0.9827	0.9726
15	0.9999	0.9998	0.9998	0.9996	0.9995	0.9988	0.9976	0.9954	0.9918	0.9862
16	1.0000	0.9999	0.9999	0.9999	0.9998	0.9996	0.9990	0.9980	0.9963	0.9934
17	1.0000	1.0000	1.0000	1.0000	0.9999	0.9998	0.9996	0.9992	0.9984	0.9970
18	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9997	0.9993	0.9987
19	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9997	0.9995
20	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9998
21	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999
22	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000

Table for Poisson random variable  $P(Po(\mu) \leq k)$ .

$k$	$\mu$																
	9.0	9.5	10.0	11.0	12.0	13.0	14.0	15.0	16.0	17.0							
0	0.0001	0.0001	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000							
1	0.0012	0.0008	0.0005	0.0002	0.0001	0.0000	0.0000	0.0000	0.0000	0.0000							
2	0.0062	0.0042	0.0028	0.0012	0.0005	0.0002	0.0001	0.0000	0.0000	0.0000							
3	0.0212	0.0149	0.0103	0.0049	0.0023	0.0011	0.0005	0.0002	0.0001	0.0000							
4	0.0550	0.0403	0.0293	0.0151	0.0076	0.0037	0.0018	0.0009	0.0004	0.0002							
5	0.1157	0.0885	0.0671	0.0375	0.0203	0.0107	0.0055	0.0028	0.0014	0.0007							
6	0.2068	0.1649	0.1301	0.0786	0.0458	0.0259	0.0142	0.0076	0.0040	0.0021							
7	0.3239	0.2687	0.2202	0.1432	0.0895	0.0540	0.0316	0.0180	0.0100	0.0054							
8	0.4557	0.3918	0.3328	0.2320	0.1550	0.0998	0.0621	0.0374	0.0220	0.0126							
9	0.5874	0.5218	0.4579	0.3405	0.2424	0.1658	0.1094	0.0699	0.0433	0.0261							
10	0.7060	0.6453	0.5830	0.4599	0.3472	0.2517	0.1757	0.1185	0.0774	0.0491							
11	0.8030	0.7520	0.6968	0.5793	0.4616	0.3532	0.2600	0.1848	0.1270	0.0847							
12	0.8758	0.8364	0.7916	0.6887	0.5760	0.4631	0.3585	0.2676	0.1931	0.1350							
13	0.9261	0.8981	0.8645	0.7813	0.6815	0.5730	0.4644	0.3632	0.2745	0.2009							
14	0.9585	0.9400	0.9165	0.8540	0.7720	0.6751	0.5704	0.4657	0.3675	0.2808							
15	0.9780	0.9665	0.9513	0.9074	0.8444	0.7636	0.6694	0.5681	0.4667	0.3715							
16	0.9889	0.9823	0.9730	0.9441	0.8987	0.8355	0.7559	0.6641	0.5660	0.4677							
17	0.9947	0.9911	0.9857	0.9678	0.9370	0.8905	0.8272	0.7489	0.6593	0.5640							
18	0.9976	0.9957	0.9928	0.9823	0.9626	0.9302	0.8826	0.8195	0.7423	0.6550							
19	0.9989	0.9980	0.9965	0.9907	0.9787	0.9573	0.9235	0.8752	0.8122	0.7363							
20	0.9996	0.9991	0.9984	0.9953	0.9884	0.9750	0.9521	0.9170	0.8682	0.8055							
21	0.9998	0.9996	0.9993	0.9977	0.9939	0.9859	0.9712	0.9469	0.9108	0.8615							
22	0.9999	0.9999	0.9997	0.9990	0.9970	0.9924	0.9833	0.9673	0.9418	0.9047							
23	1.0000	0.9999	0.9999	0.9995	0.9985	0.9960	0.9907	0.9805	0.9633	0.9367							
24	1.0000	1.0000	1.0000	0.9998	0.9993	0.9980	0.9950	0.9888	0.9777	0.9594							
25	1.0000	1.0000	1.0000	0.9999	0.9997	0.9990	0.9974	0.9938	0.9869	0.9748							
26	1.0000	1.0000	1.0000	1.0000	0.9999	0.9995	0.9987	0.9967	0.9925	0.9848							
27	1.0000	1.0000	1.0000	1.0000	0.9999	0.9998	0.9994	0.9983	0.9959	0.9912							
28	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9997	0.9991	0.9978	0.9950							
29	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9996	0.9994	0.9986							
30	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9997	0.9993							
31	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9997	0.9996							
32	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9996							
33	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999	0.9999	0.9999							
34	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	0.9999							
35	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000							