A formulae sheet is attached.
Only approved basic scientific calculators may be used.

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Examination for the Module MATH2740
(May-June 2006)

## Environmental Statistics

## Time allowed: 2 hours

Do not answer more than four questions.
All questions carry equal marks.

1. (a) Define, in terms of the mean number of points per unit area and the variance in the number of points per unit area: (i) a random point pattern; (ii) a clustered point pattern; (iii) a regular point pattern.
(b) (i) The number of clusters, $M$, in a given area follows a Poisson distribution with probability generating function:

$$
G_{M}(s)=\exp \left[\lambda_{1}(s-1)\right], \quad\left(\lambda_{1}>0\right)
$$

The number of points in the $j$ th cluster is denoted by $Y_{j}$ for $j=1,2, \ldots, M$ each following an independent Poisson distribution with probability generating function:

$$
G_{Y}(s)=\exp \left[\lambda_{2}(s-1)\right], \quad\left(\lambda_{2}>0\right) .
$$

The total number of points in the given area is calculated as $X=Y_{1}+\cdots+Y_{M}$.
Briefly explain, in terms of the number of points in a cluster, why the Poisson distribution is a theoretically inadequate model.
(ii) Show that the probability generating function of $X, G_{X}(s)$ is given by:

$$
G_{X}(s)=\exp \left[\lambda_{1}\left\{\exp \left[\lambda_{2}(s-1)\right]-1\right\}\right] .
$$

(iii) Use the probability generating function, $G_{X}(s)$ to show that:

$$
\begin{aligned}
\mathrm{P}(X=0) & =\exp \left[\lambda_{1}\left\{\exp \left(-\lambda_{2}\right)-1\right\}\right], \\
\mathrm{E}(X) & =\lambda_{1} \lambda_{2} .
\end{aligned}
$$

(c) The table below gives the distribution of the number of Antirrhinum majus (snapdragon) plants in 300 quadrats each of size 4 square metres:

| Number of plants per quadrat | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | $8+$ | Total |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :--- |
| Number of quadrats | 165 | 62 | 24 | 12 | 13 | 8 | 5 | 11 | 0 | 300 |

Let the random variable $X$ denote the number of plants per quadrat. It has been suggested that the Neyman Type A distribution with parameters $\lambda_{1}$ and $\lambda_{2}$ be used to model $X$.
(i) The Neyman Type A model can be fitted using the method of moments. Show that this leads to parameter estimates $\widehat{\lambda}_{1}=0.6092$ and $\widehat{\lambda}_{2}=1.8877$ (to 4 decimal places).
(ii) The expected frequencies (to 2 decimal places) under the Neyman Type A model are shown below.

| Number of plants per quadrat | 0 | 1 | 2 | 3 | 4 | 5 | 6 |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Number of quadrats | $e_{0}$ | 31.15 | 32.12 | 23.78 | 14.82 | 8.58 | 4.85 |
| Number of plants per quadrat | 7 | $8+$ | Total |  |  |  |  |
| Number of quadrats | 2.70 | $e_{8+}$ | 300 |  |  |  |  |

Calculate the values to be inserted in place of $e_{0}$ and $e_{8+}$.
(iii) Perform a $\chi^{2}$ goodness-of-fit test to assess the proposed model and comment on the result.
2. (a) Explain how departure from a spatially random pattern affects the distance from a randomly chosen point to the nearest neighbouring plant assuming: (i) a clustered pattern; (ii) a regular pattern.
(b) Consider a Poisson forest of plants, with intensity $\lambda$ plants per unit area. Let the random variable $U$ denote the squared distance from a randomly selected point to the nearest plant.
(i) Consider the probability density function of $U$ :

$$
\mathrm{f}(u)=\lambda \pi \exp (-\lambda \pi u), \quad(u \geq 0)
$$

Using integration by parts, or otherwise, show that:

$$
\begin{aligned}
\mathrm{E}(U) & =1 /(\lambda \pi), \\
\operatorname{Var}(U) & =1 /(\lambda \pi)^{2} .
\end{aligned}
$$

(ii) Let the random variable $\bar{U}=\left(U_{1}+\cdots+U_{m}\right) / m$ denote the mean of a set of $m$ independent squared point-plant measurements from the Poisson forest. Use the results from (i) to argue that:

$$
\begin{aligned}
\mathrm{E}(\bar{U}) & =1 /(\lambda \pi), \\
\operatorname{Var}(\bar{U}) & =1 /\left[m(\lambda \pi)^{2}\right] .
\end{aligned}
$$

(iii) Under the null hypothesis $H_{0}$ : spatially random pattern, the test statistic:

$$
\frac{\bar{U}-1 /(\lambda \pi)}{\sqrt{1 /\left[m(\lambda \pi)^{2}\right]}}
$$

will approximate a standard normal distribution. Briefly explain why this is so, and describe the conditions under which the approximation is reasonable.
(c) A quadrat sample reveals that there are, on average, 2.5 Eschscholzia californica (Californian Poppy) plants per 3 square metre quadrat. A set of $m=20$ squared point-plant measurements are taken, giving a mean of $\bar{u}=0.60$ square metres. Use the result of (iii) above to test the null hypothesis that the plants are located at random against an alternative that the locations are non-random. Is the departure from randomness towards clustering or regularity? Do you have any reservations about the test?
3. (a) Briefly explain what is meant by the terms: (i) positive spatial autocorrelation; (ii) negative spatial autocorrelation; (iii) free-sampling.
(b) (i) A plant pathologist is studying the germination of Triticum aestivum (bread wheat). 100 seeds are sown in each of sixteen 10 square metre quadrats, arranged on a contiguous grid. Based on previous work, the pathologist believes that the proportion of wheat seeds that germinate in each quadrat, $p$, can be modelled by the probability density function:

$$
\mathrm{f}(p)=\frac{p}{3}+a, \quad(0<p<1)
$$

Derive the cumulative distribution function of, $p, \mathrm{~F}(p)$. Hence determine the value of $a$ and show that the median proportion of seeds that germinate is given by $m=$ 0.5414 (to 4 decimal places).
(ii) Justify the use of the formula:

$$
B W=\frac{1}{2} \sum_{i} \sum_{j} \delta_{i j} x_{i}\left(1-x_{j}\right),
$$

for the number of black/white (i.e. $B W$ ) joins in a black/white map of spatial pattern, where:

$$
x_{i}= \begin{cases}1 & \text { if cell } i \text { is } B \\ 0 & \text { if cell } i \text { is } W\end{cases}
$$

and $\delta_{i j}$ denotes the contiguity matrix defined as:

$$
\delta_{i j}=\left\{\begin{array}{cc}
1 & \text { if cells } i \text { and } j \text { are joined } \\
0 & \text { otherwise }
\end{array}\right.
$$

with $\delta_{i i}=0$ for all $i$.
(c) (i) The pathologist wants to investigate the effect of spatial autocorrelation on the proportion of seeds that germinate. The proportions that germinate are summarised below:

| 0.0863 | 0.1367 | 0.2294 | 0.3190 |
| :--- | :--- | :--- | :--- |
| 0.1699 | 0.6748 | 0.8184 | 0.5610 |
| 0.9909 | 0.9726 | 0.2980 | 0.2010 |
| 0.9389 | 0.9293 | 0.7436 | 0.6304 |

By denoting quadrats in which the proportion of seeds that germinate is at or above $m$ as $B$ (and all other quadrats as $W$ ), show that the following black/white map is produced:


State whether free-or non-free sampling has been used.
(ii) Diagonal trends are of particular interest to the pathologist, and hence they wish to use the bishop's definition of contiguity. That is, two quadrats are considered joined if and only if they meet at a corner but not at an edge. Determine the number of $B W$ joins using this definition of contiguity. It may be shown that (you do not need to check this) $L=18$ and $K=32$. Assess the evidence for the presence of spatial autocorrelation using $B W$ joins. Comment on the outcome of the test, and the validity of any assumptions in the test procedure.
4. (a) Explain what is meant by: (i) a saturated model; (ii) the hierarchy principle.
(b) Consider the saturated model for a $2 \times 2$ contingency table with observed frequencies $\left\{f_{i j}\right\}$ :

$$
y_{i j}=\ln \left(f_{i j}\right)=\mu+\lambda_{i}^{A}+\lambda_{j}^{B}+\lambda_{i j}^{A B}, \quad(i=1,2), \quad(j=1,2),
$$

subject to cornered constraints:

$$
\lambda_{2}^{A}=\lambda_{2}^{B}=\lambda_{i 2}^{A B}=\lambda_{2 j}^{A B}=0 .
$$

Show that:

$$
\begin{gathered}
\mu=y_{22}, \quad \lambda_{1}^{A}=y_{12}-y_{22}, \\
\lambda_{1}^{B}=y_{21}-y_{22}, \quad \lambda_{11}^{A B}=y_{11}-y_{12}-y_{21}+y_{22} .
\end{gathered}
$$

(c) An ecologist studies the attraction/repulsion between two species of butterfly Pieris rapae (cabbage white) and Vanessa atalanta (red admiral) by dividing a large area into 210 quadrats. The variables are $A$ ( $1=$ cabbage white present, $2=$ cabbage white absent) and $B(1=$ red admiral present, $2=$ red admiral absent).

|  | $A_{1}$ | $A_{2}$ |
| :---: | :---: | :---: |
|  | 87 | 8 |
| $B_{1}$ | 87 | 8 |
| $B_{2}$ | 83 | 32 |
|  |  |  |

(i) Fit the saturated model to these data, obtaining parameter estimates $\widehat{\mu}, \widehat{\lambda}_{1}^{A}, \widehat{\lambda}_{1}^{B}$ and $\widehat{\lambda}_{11}^{A B}$.
(ii) Fit the independence model to these data and show that the expected frequencies, $\left\{e_{i j}\right\}$, have an odds ratio $\phi=1$.
(iii) Using the likelihood ratio statistic, $Y^{2}$, test whether the independence model, $A / B$, is a good fit to the data. Interpret the result with reference to the attraction/repulsion of the two species of butterfly. Do you have any reservations about the experiment conducted?
5. A study was conducted on the association between gender, eating habits, and concern over factory farming. Data from the study were analysed using $R$. The variables were gender ( $1=$ male, $2=$ female), meat ( $1=$ eats meat, $2=$ does not eat meat) and farm ( $1=$ concerned about factory faming, $2=$ not concerned about factory farming).

The aim of the analysis was to build a log-linear model which explained the significant factors and interactions present in the data. Consider the following extract from an R session:

```
> xtabs(count~meat+gender+farm,data=factory)
    , , farm = 1
        gender
    meat 1 2
        1 35 34
        2 22 51
    , , farm = 2
        gender
    meat 1 2
        14340
        2 12 27
> xtabs(count~meat+farm+gender,data=factory)
    , , gender = 1
        farm
    meat 1 2
        1 35 43
        2 22 12
    , , gender = 2
        farm
    meat 1 2
        1 34 40
        2 51 27
> xtabs(count~farm+gender+meat,data=factory)
    , , meat = 1
        gender
    farm 1 2
        1 35 34
        24340
    , , meat = 2
        gender
    farm 1 2
        1 22 51
        2 12 27
```

```
> options(contrasts=c("contr.sum","contr.poly"))
> saturated.model = glm(count~gender*meat*farm,family=poisson)
> summary(saturated.model)
        Coefficients:
```



```
                meat1 0.216022 0.067508 3.200 0.00137 **
                farm1 0.109219 0.067508 1.618 0.10569
        gender1:meat1 0.219128 0.067508 3.246 0.00117 **
        gender1:farm1 -0.009148 0.067508 -0.136 0.89221
        meat1:farm1 -0.201312 0.067508 -2.982 0.00286 **
    gender1:meat1:farm1 -0.001685 0.067508 -0.025 0.98009
    Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '', 1
```

(a) Let $A$ denote the variable gender, $B$ denote the variable meat, and $C$ denote the variable farm. Estimate the odds-ratios for the six $2 \times 2$ tables generated by the xtabs() commands. Based on these estimated odds-ratios, use hierarchy notation to suggest a suitable model for the data. How many degrees of freedom are available to test the model you propose?
(b) Write down the expression you would insert in place of the ... in

```
fitted.model=glm(count~ ... ,family=poisson)
```

to fit the model you suggested in (a) using $R$.
(c) Consider the output of the summary () command. List the factors and interactions in the saturated model which are significant at the $5 \%$ level. Interpret any significant factors and interactions with reference to gender, eating habits, and concern over factory farming.

## Formulae Sheet

## Poisson

$$
\begin{gathered}
\mathrm{P}(X=x)=\frac{\exp (-\lambda) \lambda^{x}}{x!}, \quad(x=0,1, \ldots) \\
\mathrm{E}(X)=\operatorname{Var}(X)=\lambda, \quad G_{X}(s)=\exp [\lambda(s-1)]
\end{gathered}
$$

## Logarithmic

$$
\begin{gathered}
\mathrm{P}(X=x)=c \frac{q^{x}}{x}, \quad(x=1,2, \ldots), \quad(0<q<1), \quad c=\frac{-1}{\ln (1-q)} . \\
\mathrm{E}(X)=\frac{c q}{(1-q)}, \quad G_{X}(s)=-c \ln (1-q s) .
\end{gathered}
$$

Neyman Type A

$$
\mathrm{E}(X)=\lambda_{1} \lambda_{2}, \quad \operatorname{Var}(X)=\lambda_{1} \lambda_{2}\left(1+\lambda_{2}\right), \quad G_{X}(s)=\exp \left[\lambda_{1}\left\{\exp \left[\lambda_{2}(s-1)\right]-1\right\}\right] .
$$

Negative Binomial

$$
\begin{gathered}
\mathrm{P}(X=x)=\binom{r+x-1}{x} p^{r}(1-p)^{x}, \quad(x=0,1, \ldots), \quad(0<p<1) . \\
\mathrm{E}(X)=\frac{r(1-p)}{p}, \quad \operatorname{Var}(X)=\frac{r(1-p)}{p^{2}}, \quad G_{X}(s)=\left(\frac{p}{1-q s}\right)^{r}, \quad(q=1-p) .
\end{gathered}
$$

## Gamma

$$
\begin{gathered}
\mathrm{f}(x)=\frac{\alpha^{r}}{\Gamma(r)} x^{r-1} \exp (-\alpha x), \quad(x \geq 0), \quad(r, \alpha>0) . \\
\mathrm{E}(X)=\frac{r}{\alpha}, \quad \operatorname{Var}(X)=\frac{r}{\alpha^{2}} .
\end{gathered}
$$

## Weibull

$$
\begin{gathered}
\mathrm{f}(r)=2 \lambda \pi r \exp \left(-\lambda \pi r^{2}\right), \quad(r \geq 0), \quad(\lambda>0) . \\
\mathrm{E}(R)=\frac{1}{2 \sqrt{\lambda}}, \quad \operatorname{Var}(R)=\frac{4-\pi}{4 \lambda \pi} .
\end{gathered}
$$

Testing for spatial randomness

$$
\begin{gathered}
Z=\frac{(n-1) s^{2}}{\bar{x}} \sim \chi_{n-1}^{2} . \\
\mathrm{E}(\bar{r})=\mathrm{E}(R)=\frac{1}{2 \sqrt{\lambda}} . \\
\operatorname{Var}(\bar{r})=\frac{\operatorname{Var}(R)}{m}=\frac{4-\pi}{4 \lambda \pi m} . \\
S=2 m \widehat{\lambda} \pi \bar{u} \sim \chi_{2 m}^{2} . \\
H=\frac{\sum r_{1 i}^{2}}{\sum r_{2 i}^{2}} \sim \mathrm{~F}_{2 m, 2 m} .
\end{gathered}
$$

## Spatial autocorrelation

$$
L=\frac{1}{2} \sum_{i} L_{i} \text { where } L_{i} \text { is the number of cells joined to cell } i, \quad K=\frac{1}{2} \sum_{i} L_{i}\left(L_{i}-1\right) .
$$

## Free-sampling

$$
\begin{aligned}
\mathrm{E}(B B)=L p^{2}, & \mathrm{E}(B W)=2 L p q, \quad \mathrm{E}(W W)=L q^{2} \\
\operatorname{Var}(B B) & =L p^{2}+2 K p^{3}-(L+2 K) p^{4} . \\
\operatorname{Var}(B W) & =2(L+K) p q-4(L+2 K) p^{2} q^{2} . \\
\operatorname{Var}(W W) & =L q^{2}+2 K q^{3}-(L+2 K) q^{4} .
\end{aligned}
$$

## Non-free sampling

$$
\begin{gathered}
\mathrm{E}(B B)=L \frac{n_{1}\left(n_{1}-1\right)}{n(n-1)} \\
\mathrm{E}(B W)=2 L \frac{n_{1} n_{2}}{n(n-1)} \\
\mathrm{E}(W W)=L \frac{n_{2}\left(n_{2}-1\right)}{n(n-1)} . \\
\operatorname{Var}(B B)=L \frac{n_{1}\left(n_{1}-1\right)}{n(n-1)}+2 K \frac{n_{1}\left(n_{1}-1\right)\left(n_{1}-2\right)}{n(n-1)(n-2)} \\
+[L(L-1)-2 K] \frac{n_{1}\left(n_{1}-1\right)\left(n_{1}-2\right)\left(n_{1}-3\right)}{n(n-1)(n-2)(n-3)}-\left[L \frac{n_{1}\left(n_{1}-1\right)}{n(n-1)}\right]^{2} \\
\operatorname{Var}(B W)=\frac{2(L+K) n_{1} n_{2}}{n(n-1)}+4[L(L-1)-2 K] \frac{n_{1}\left(n_{1}-1\right) n_{2}\left(n_{2}-1\right)}{n(n-1)(n-2)(n-3)} \\
-4\left[\frac{L n_{1} n_{2}}{n(n-1)}\right]^{2} \\
\operatorname{Var}(W W)= \\
+L \frac{n_{2}\left(n_{2}-1\right)}{n(n-1)}+2 K \frac{n_{2}\left(n_{2}-1\right)\left(n_{2}-2\right)}{n(n-1)(n-2)} \\
+[L(L-1)-2 K] \frac{n_{2}\left(n_{2}-1\right)\left(n_{2}-2\right)\left(n_{2}-3\right)}{n(n-1)(n-2)(n-3)}-\left[L \frac{n_{2}\left(n_{2}-1\right)}{n(n-1)}\right]^{2}
\end{gathered}
$$

## Categorical data

For a $2 \times 2$ contingency table: $Y^{2}=2 \sum_{i j} f_{i j} \ln \left(f_{i j} / e_{i j}\right)$.
For the $A / B$ model: $e_{i j}=\frac{f_{i 0} f_{0 j}}{f_{00}}$.
For the $A$ model: $e_{i j}=\frac{f_{i 0}}{2}$.
For the $B$ model: $e_{i j}=\frac{f_{0 j}}{2}$.

The first table gives

$$
\Phi(x)=\frac{1}{\sqrt{2 \pi}} \int_{-\infty}^{x} e^{-\frac{1}{2} t^{2}} d t
$$

and this corresponds to the shaded area in the figure to the right. $\Phi(x)$ is the probability that a random variable, normally distributed with zero mean amd unit variance, will be less than or equal to $x$. When $x<0$ use $\Phi(x)=1-\Phi(-x)$, as the normal distribution with mean zero is symmetric about zero.


Table 1

| $x$ | $\Phi(x)$ | $x$ | $\Phi(x)$ | $x$ | $\Phi(x)$ | $x$ | $\Phi(x)$ | $x$ | $\Phi(x)$ | $x$ | $\Phi(x)$ |
| ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{0 . 0 0}$ | 0.5000 | $\mathbf{0 . 5 0}$ | 0.6915 | $\mathbf{1 . 0 0}$ | 0.8413 | $\mathbf{1 . 5 0}$ | 0.9332 | $\mathbf{2 . 0 0}$ | 0.9772 | $\mathbf{2 . 5 0}$ | 0.9938 |
| $\mathbf{0 . 0 5}$ | 0.5199 | $\mathbf{0 . 5 5}$ | 0.7088 | $\mathbf{1 . 0 5}$ | 0.8531 | $\mathbf{1 . 5 5}$ | 0.9394 | $\mathbf{2 . 0 5}$ | 0.9798 | $\mathbf{2 . 5 5}$ | 0.9946 |
| $\mathbf{0 . 1 0}$ | 0.5398 | $\mathbf{0 . 6 0}$ | 0.7257 | $\mathbf{1 . 1 0}$ | 0.8643 | $\mathbf{1 . 6 0}$ | 0.9452 | $\mathbf{2 . 1 0}$ | 0.9821 | $\mathbf{2 . 6 0}$ | 0.9953 |
| $\mathbf{0 . 1 5}$ | 0.5596 | $\mathbf{0 . 6 5}$ | 0.7422 | $\mathbf{1 . 1 5}$ | 0.8749 | $\mathbf{1 . 6 5}$ | 0.9505 | $\mathbf{2 . 1 5}$ | 0.9842 | $\mathbf{2 . 6 5}$ | 0.9960 |
| $\mathbf{0 . 2 0}$ | 0.5793 | $\mathbf{0 . 7 0}$ | 0.7580 | $\mathbf{1 . 2 0}$ | 0.8849 | $\mathbf{1 . 7 0}$ | 0.9554 | $\mathbf{2 . 2 0}$ | 0.9861 | $\mathbf{2 . 7 0}$ | 0.9965 |
|  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{0 . 2 5}$ | 0.5987 | $\mathbf{0 . 7 5}$ | 0.7734 | $\mathbf{1 . 2 5}$ | 0.8944 | $\mathbf{1 . 7 5}$ | 0.9599 | $\mathbf{2 . 2 5}$ | 0.9878 | $\mathbf{2 . 7 5}$ | 0.9970 |
| $\mathbf{0 . 3 0}$ | 0.6179 | $\mathbf{0 . 8 0}$ | 0.7881 | $\mathbf{1 . 3 0}$ | 0.9032 | $\mathbf{1 . 8 0}$ | 0.9641 | $\mathbf{2 . 3 0}$ | 0.9893 | $\mathbf{2 . 8 0}$ | 0.9974 |
| $\mathbf{0 . 3 5}$ | 0.6368 | $\mathbf{0 . 8 5}$ | 0.8023 | $\mathbf{1 . 3 5}$ | 0.9115 | $\mathbf{1 . 8 5}$ | 0.9678 | $\mathbf{2 . 3 5}$ | 0.9906 | $\mathbf{2 . 8 5}$ | 0.9978 |
| $\mathbf{0 . 4 0}$ | 0.6554 | $\mathbf{0 . 9 0}$ | 0.8159 | $\mathbf{1 . 4 0}$ | 0.9192 | $\mathbf{1 . 9 0}$ | 0.9713 | $\mathbf{2 . 4 0}$ | 0.9918 | $\mathbf{2 . 9 0}$ | 0.9981 |
| $\mathbf{0 . 4 5}$ | 0.6736 | $\mathbf{0 . 9 5}$ | 0.8289 | $\mathbf{1 . 4 5}$ | 0.9265 | $\mathbf{1 . 9 5}$ | 0.9744 | $\mathbf{2 . 4 5}$ | 0.9929 | $\mathbf{2 . 9 5}$ | 0.9984 |
|  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{0 . 5 0}$ | 0.6915 | $\mathbf{1 . 0 0}$ | 0.8413 | $\mathbf{1 . 5 0}$ | 0.9332 | $\mathbf{2 . 0 0}$ | 0.9772 | $\mathbf{2 . 5 0}$ | 0.9938 | $\mathbf{3 . 0 0}$ | 0.9987 |

The inverse function $\Phi^{-1}(p)$ is tabulated below for various values of $p$.
Table 2

| $p$ | $\mathbf{0 . 9 0 0}$ | $\mathbf{0 . 9 5 0}$ | $\mathbf{0 . 9 7 5}$ | $\mathbf{0 . 9 9 0}$ | $\mathbf{0 . 9 9 5}$ | $\mathbf{0 . 9 9 9}$ | $\mathbf{0 . 9 9 9 5}$ |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| $\Phi^{-1}(p)$ | 1.2816 | 1.6449 | 1.9600 | 2.3263 | 2.5758 | 3.0902 | 3.2905 |

## Percentage Points of the $\chi^{2}$-Distribution

This table gives the percentage points $\chi_{\nu}^{2}(P)$ for various values of $P$ and degrees of freedom $\nu$, as indicated by the figure to the right.

If $X$ is a variable distributed as $\chi^{2}$ with $\nu$ degrees of freedom, $P / 100$ is the probability that $X \geq \chi_{\nu}^{2}(P)$.

For $\nu>100, \sqrt{2 X}$ is approximately normally distributed with mean $\sqrt{2 \nu-1}$ and unit variance.


|  | Percentage points $P$ |  |  |  |  |  |
| ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| $\boldsymbol{\nu}$ | $\mathbf{9 9}$ | $\mathbf{9 7 . 5}$ | $\mathbf{9 5}$ | $\mathbf{5}$ | $\mathbf{2 . 5}$ | $\mathbf{1}$ |
| $\mathbf{1}$ | 0.000 | 0.001 | 0.004 | 3.841 | 5.024 | 6.635 |
| $\mathbf{2}$ | 0.020 | 0.051 | 0.103 | 5.992 | 7.378 | 9.210 |
| $\mathbf{3}$ | 0.115 | 0.216 | 0.352 | 7.815 | 9.348 | 11.345 |
| $\mathbf{4}$ | 0.297 | 0.484 | 0.711 | 9.488 | 11.143 | 13.277 |
| $\mathbf{5}$ | 0.554 | 0.831 | 1.145 | 11.070 | 12.833 | 15.086 |
|  |  |  |  |  |  |  |
| $\mathbf{6}$ | 0.872 | 1.237 | 1.635 | 12.592 | 14.449 | 16.812 |
| $\mathbf{7}$ | 1.239 | 1.690 | 2.167 | 14.067 | 16.013 | 18.475 |
| $\mathbf{8}$ | 1.646 | 2.180 | 2.733 | 15.507 | 17.535 | 20.090 |
| $\mathbf{9}$ | 2.088 | 2.700 | 3.325 | 16.919 | 19.023 | 21.666 |
| $\mathbf{1 0}$ | 2.558 | 3.247 | 3.940 | 18.307 | 20.483 | 23.209 |
|  |  |  |  |  |  |  |
| $\mathbf{1 1}$ | 3.053 | 3.816 | 4.575 | 19.675 | 21.920 | 24.725 |
| $\mathbf{1 2}$ | 3.571 | 4.404 | 5.226 | 21.026 | 23.337 | 26.217 |
| $\mathbf{1 3}$ | 4.107 | 5.009 | 5.892 | 22.362 | 24.736 | 27.688 |
| $\mathbf{1 4}$ | 4.660 | 5.629 | 6.571 | 23.685 | 26.119 | 29.141 |
| $\mathbf{1 5}$ | 5.229 | 6.262 | 7.261 | 24.996 | 27.488 | 30.578 |
|  |  |  |  |  |  |  |
| $\mathbf{1 6}$ | 5.812 | 6.908 | 7.962 | 26.296 | 28.845 | 32.000 |
| $\mathbf{1 7}$ | 6.408 | 7.564 | 8.672 | 27.587 | 30.191 | 33.409 |
| $\mathbf{1 8}$ | 7.015 | 8.231 | 9.390 | 28.869 | 31.526 | 34.805 |
| $\mathbf{1 9}$ | 7.633 | 8.907 | 10.117 | 30.144 | 32.852 | 36.191 |
| $\mathbf{2 0}$ | 8.260 | 9.591 | 10.851 | 31.410 | 34.170 | 37.566 |
|  |  |  |  |  |  |  |
| $\mathbf{2 5}$ | 11.524 | 13.120 | 14.611 | 37.652 | 40.646 | 44.314 |
| $\mathbf{3 0}$ | 14.953 | 16.791 | 18.493 | 43.773 | 46.979 | 50.892 |
| $\mathbf{4 0}$ | 22.164 | 24.433 | 26.509 | 55.758 | 59.342 | 63.691 |
| $\mathbf{5 0}$ | 29.707 | 32.357 | 34.764 | 67.505 | 71.420 | 76.154 |
| $\mathbf{8 0}$ | 53.540 | 57.153 | 60.391 | 101.879 | 106.629 | 112.329 |
| $\mathbf{1 5}$ |  |  |  |  |  |  |

### 2.5 Percent Points of the $\boldsymbol{F}$-Distribution

This table gives the percentage points $F_{\nu_{1}, \nu_{2}}(P)$ for $P=0.025$ and degrees of freedom $\nu_{1}, \nu_{2}$, as indicated by the figure to the right.

The lower percentage points, that is the values $F_{\nu_{1}, \nu_{2}}^{\prime}(P)$ such that the probability that $F \leq$ $F_{\nu_{1}, \nu_{2}}^{\prime}(P)$ is equal to $P / 100$, may be found using the formula

$$
F_{\nu_{1}, \nu_{2}}^{\prime}(P)=1 / F_{\nu_{1}, \nu_{2}}(P)
$$



|  |  |  |  |  | $\boldsymbol{\nu}_{1}$ |  |  |  |  |  |  |  |  |  |  |
| ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\boldsymbol{\nu}_{2}$ | $\mathbf{1}$ | $\mathbf{2}$ | $\mathbf{3}$ | $\mathbf{4}$ | $\mathbf{5}$ | $\mathbf{6}$ | $\mathbf{1 2}$ | $\mathbf{2 4}$ | $\boldsymbol{\infty}$ |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{2}$ | 38.506 | 39.000 | 39.165 | 39.248 | 39.298 | 39.331 | 39.415 | 39.456 | 39.498 |  |  |  |  |  |  |
| $\mathbf{3}$ | 17.443 | 16.044 | 15.439 | 15.101 | 14.885 | 14.735 | 14.337 | 14.124 | 13.902 |  |  |  |  |  |  |
| $\mathbf{4}$ | 12.218 | 10.649 | 9.979 | 9.605 | 9.364 | 9.197 | 8.751 | 8.511 | 8.257 |  |  |  |  |  |  |
| $\mathbf{5}$ | 10.007 | 8.434 | 7.764 | 7.388 | 7.146 | 6.978 | 6.525 | 6.278 | 6.015 |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{6}$ | 8.813 | 7.260 | 6.599 | 6.227 | 5.988 | 5.820 | 5.366 | 5.117 | 4.849 |  |  |  |  |  |  |
| $\mathbf{7}$ | 8.073 | 6.542 | 5.890 | 5.523 | 5.285 | 5.119 | 4.666 | 4.415 | 4.142 |  |  |  |  |  |  |
| $\mathbf{8}$ | 7.571 | 6.059 | 5.416 | 5.053 | 4.817 | 4.652 | 4.200 | 3.947 | 3.670 |  |  |  |  |  |  |
| $\mathbf{9}$ | 7.209 | 5.715 | 5.078 | 4.718 | 4.484 | 4.320 | 3.868 | 3.614 | 3.333 |  |  |  |  |  |  |
| $\mathbf{1 0}$ | 6.937 | 5.456 | 4.826 | 4.468 | 4.236 | 4.072 | 3.621 | 3.365 | 3.080 |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{1 1}$ | 6.724 | 5.256 | 4.630 | 4.275 | 4.044 | 3.881 | 3.430 | 3.173 | 2.883 |  |  |  |  |  |  |
| $\mathbf{1 2}$ | 6.554 | 5.096 | 4.474 | 4.121 | 3.891 | 3.728 | 3.277 | 3.019 | 2.725 |  |  |  |  |  |  |
| $\mathbf{1 3}$ | 6.414 | 4.965 | 4.347 | 3.996 | 3.767 | 3.604 | 3.153 | 2.893 | 2.595 |  |  |  |  |  |  |
| $\mathbf{1 4}$ | 6.298 | 4.857 | 4.242 | 3.892 | 3.663 | 3.501 | 3.050 | 2.789 | 2.487 |  |  |  |  |  |  |
| $\mathbf{1 5}$ | 6.200 | 4.765 | 4.153 | 3.804 | 3.576 | 3.415 | 2.963 | 2.701 | 2.395 |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{1 6}$ | 6.115 | 4.687 | 4.077 | 3.729 | 3.502 | 3.341 | 2.889 | 2.625 | 2.316 |  |  |  |  |  |  |
| $\mathbf{1 7}$ | 6.042 | 4.619 | 4.011 | 3.665 | 3.438 | 3.277 | 2.825 | 2.560 | 2.247 |  |  |  |  |  |  |
| $\mathbf{1 8}$ | 5.978 | 4.560 | 3.954 | 3.608 | 3.382 | 3.221 | 2.769 | 2.503 | 2.187 |  |  |  |  |  |  |
| $\mathbf{1 9}$ | 5.922 | 4.508 | 3.903 | 3.559 | 3.333 | 3.172 | 2.720 | 2.452 | 2.133 |  |  |  |  |  |  |
| $\mathbf{2 0}$ | 5.871 | 4.461 | 3.859 | 3.515 | 3.289 | 3.128 | 2.676 | 2.408 | 2.085 |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{2 5}$ | 5.686 | 4.291 | 3.694 | 3.353 | 3.129 | 2.969 | 2.515 | 2.242 | 1.906 |  |  |  |  |  |  |
| $\mathbf{3 0}$ | 5.568 | 4.182 | 3.589 | 3.250 | 3.026 | 2.867 | 2.412 | 2.136 | 1.787 |  |  |  |  |  |  |
| $\mathbf{4 0}$ | 5.424 | 4.051 | 3.463 | 3.126 | 2.904 | 2.744 | 2.288 | 2.007 | 1.637 |  |  |  |  |  |  |
| $\mathbf{5 0}$ | 5.340 | 3.975 | 3.390 | 3.054 | 2.833 | 2.674 | 2.216 | 1.931 | 1.545 |  |  |  |  |  |  |
| $\mathbf{1 0 0}$ | 5.179 | 3.828 | 3.250 | 2.917 | 2.696 | 2.537 | 2.077 | 1.784 | 1.347 |  |  |  |  |  |  |
| $\mathbf{\infty}$ |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| $\mathbf{\infty}$ | 5.024 | 3.689 | 3.116 | 2.786 | 2.567 | 2.408 | 1.945 | 1.640 | 1.003 |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |

