An Introduction to Veterinary Epidemiology

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1 Introduction

By the end of this unit you should be able to:

- Compare and contrast clinical approaches and epidemiological approaches to disease management.
- Describe the factors that influence the presence of disease in individuals.
- Describe the factors that influence the presence of disease in populations.
- Explain what is meant by the term causation.

Epidemiology is the study of diseases in populations. Epidemiologists attempt to characterise those individuals in a population with high rates of disease and those with low rates. They then ask questions that help them discover what the high rate group is doing that the low rate group is not or vice versa. This allows the factors influencing the rate of disease to be identified. Once identified, measures can be applied to reduce the level of exposure to these factors — reducing the rate of disease in the population. This allows disease to be controlled even if the precise pathogenic mechanism (or the aetiologic agent) is not known.

It is useful to distinguish epidemiological from clinical approaches to disease management. The **clinical approach** to disease management is focussed on individual animals and is aimed at diagnosing a disease and treating it. It involves physical examination and generation of a list of differential diagnoses. Further examinations, laboratory tests and possibly response to treatment are then used to narrow the list of differential diagnoses to a single diagnosis. In an ideal world this will always be the correct diagnosis. The success of this approach depends on two conditions:

- That the true diagnosis is on the list of differential diagnoses; and
- Clinical signs arise from a single (disease process in the individual.

Research in health professionals has shown that the final diagnosis is nearly always drawn from the initial differential list. If the disease is not on the initial list of differentials then it tends not to become the final diagnosis. Diseases may be omitted from the list because the clinician is not familiar with them (exotic or unusual diseases) or because the disease is ‘new’ and has never been identified before. The single cause idea is true in some diseases (e.g. parvo virus causes a characteristic clinical syndrome in dogs) however in many cases there are multiple causative factors interacting in a complex web that may or may not produce disease.

The **epidemiological approach** to disease management is conceptually different in that there is no dependency on being able to precisely define the aetiological agent. It is based on observing differences and similarities between diseased and non-diseased animals in order to try and understand what factors may be increasing or reducing the risk of disease.
In practice, clinicians unwittingly use a combination of clinical and epidemiological approaches in their day-to-day work. If the problem is relatively clear-cut then an epidemiological approach plays a very minor role. If the condition is new or more complex then the epidemiological approach is preferred since it will provide a better understanding of what makes individuals susceptible to disease and — once these factors are known — the measures required to control the disease become better defined.

1.1 Host, agent, and environment

Whether or not disease occurs in an individual depends often on an interplay of three things:

- The host;
- The agent; and
- The environment

The host is the animal (or human) that may contract a disease. Age, genetic makeup, level of exposure, and state of health all influence a host’s susceptibility to developing disease. The agent is the factor that causes the disease (bacteria, virus, parasite, fungus, chemical poison, nutritional deficiency etc) — one or more agents may be involved. The environment includes surroundings and conditions either within the host or external to it, that cause or allow disease transmission to occur. The environment may weaken the host and increase its susceptibility to disease or provide conditions that favour the survival of the agent.

1.2 Individual, place, and time

The level of disease in a population depends often on an interplay of three factors:

- Individual factors: what types of individuals tend to develop disease and who tends to be spared?
- Spatial factors: where is the disease especially common or rare, and what is different about those places?
- Temporal factors: how does disease frequency change over time, and what other factors are associated with those changes?
1.2.1 Individual

Individuals can be grouped or distinguished on a number of characteristics: age, sex, breed, coat colour and so on. An important component of epidemiological research is aimed at determining the influence of individual characteristics on the risk of disease. Figure 1 shows how mortality rate for drowning varied among children and young adults in the USA during 1999. The rate was highest in those aged 1 - 4 years: an age when children are mobile and curious about everything around them, even though they do not understand the hazards of deep water or how to survive if they fall in. What conclusions do we draw from this? Mortality as a result of drowning is highest in children aged 1 – 4 years: preventive measures should be targeted at this age group.

![Bar chart showing mortality from drowning by age in the USA, 1999](image)


1.2.2 Place

The spatial pattern of disease is typically a consequence of environmental factors. Environmental factors include aspects of climate (temperature, humidity, rainfall) as well as aspects of animal management (management of animals in a certain area of a country may result in high rates of disease that may not be seen in other areas). Geographic Information Systems and easy access to spatial data (e.g. satellite images) have facilitated the ability to conduct spatial epidemiological analyses in recent years. Figure 2 shows the geographical distribution of BSE incidence risk in British cattle from 1986 to 1997. These maps show a higher density of disease in the south of the country, compared with the north.
Temporal patterns of disease in populations are presented graphically using epidemic curves. An epidemic curve consists of bar charts showing time on the horizontal axis and the number of new cases on the vertical axis, as shown in Figure 5. The shape of an epidemic curve can provide important information about the nature of the disease under investigation. An epidemic occurs when there is a rapid increase in the level of disease in a population. An epidemic is usually heralded by an exponential rise in the number of cases in time and a subsequent decline as susceptible animals are exhausted. Epidemics may arise from the introduction of a novel pathogen (or strain) to a previously unexposed (naive) population or as a result of the re-growth of susceptible numbers some time after a previous epidemic due to the same infectious agent. Epidemics may be described as being either common source or propagated.

In a common source epidemic, subjects are exposed to a common noxious influence. If the group is exposed over a relatively short period then disease cases will emerge over one incubation period. This is classified as a common point source epidemic. The epidemic of leukaemia cases in Hiroshima following the atomic bomb blast would be a good example of a point source epidemic. The shape of this curve rises rapidly and
contains a definite peak at the top, followed by a gradual decline. Exposure can also occur over a longer period of time, either intermittently or continuously. This creates either an intermittent common source epidemic or a continuous common source epidemic. The shape of this curve rises rapidly (associated with the introduction of the agent). The down slope of the curve may be very sharp if the common source is removed or gradual if the outbreak is allowed to exhaust itself.

**A propagated epidemic** occurs when a case of disease serves as a source of infection for subsequent cases and those subsequent cases, in turn, serve as sources for later cases. In theory, the epidemic curve of a propagated epidemic has a successive series of peaks reflecting increasing numbers of cases in each generation. The epidemic usually wanes after a few generations, either because the number of susceptibles falls below a critical level, or because intervention measures become effective.

![Figure 3: Epidemic curves. The plot on the left is typical of a propagated epidemic. The curve on the right is typical of a common source epidemic.](image)

Sometimes epidemic curves can show characteristics of being both common source and propagated. **Figure 4** shows the epidemic curve for foot-and-mouth disease in the county of Cumbria (Great Britain) in 2001. This epidemic started as a common (point) source, then become propagative over time.

**Endemic** describes levels of disease which do not exhibit wide fluctuations over time. Epidemic curves for endemic disease might show evidence of seasonal variation (as in the case of monthly reports of human leptospirosis cases in the USA, shown on the left in **Figure 5**). If data are graphed over extended periods, long-term trends might be evident (as in the reported wildlife and dog rabies cases in the USA from 1946 to 1965, shown on the right in **Figure 5**).
1.3 Causation

The basis for most epidemiological investigations is the assumption that disease does not normally occur in a random fashion — something causes it. As a result we can use epidemiological investigations to identify causal relationships and potential risk factors.

Most scientific investigations are aimed at identifying cause-effect relationships. Webster’s dictionary defines a cause as ‘something that brings about an effect or a result.’ A cause of a disease is an event, condition, or characteristic which plays an essential role in producing an occurrence of the disease. Knowledge about cause-and-effect relationships underlies every therapeutic manoeuvre in clinical medicine. The situation is complicated if multiple causes are involved. Koch (1884) provided a framework for identifying causes of infectious disease. He specified that the following criteria (known as Koch’s postulates) had to be met before an agent could be considered as the cause of a disease:

- The agent has to be present in every case of the disease.
- The agent has to be isolated and grown in pure culture.
- The agent has to cause disease when inoculated into a susceptible animal and the agent must then be able to be recovered from that animal and identified.
In the late nineteenth century Koch’s postulates brought a degree of order and discipline to the study of infectious diseases, although the key assumption of ‘one-agent-one-disease’ was highly restrictive (since it failed to take account of diseases with multiple aetiologic factors, multiple effects of single causes, carrier states, and non-agent factors such as age and breed).

Based on John Stuart Mill’s rules of inductive reasoning from 1856, Evan developed a unified concept of causation which is now the generally accepted means for identifying cause-effect relationships in modern epidemiology. Evan’s unified concept of causation includes the following criteria:

- The proportion of individuals with disease should be higher in those exposed to the putative cause than in those not exposed.
- Exposure to the putative cause should be more common in cases than in those without the disease.
- The number of new cases should be higher in those exposed to the putative cause than in those not exposed, as shown in prospective studies.
- Temporally, the disease should follow exposure to the putative cause.
- There should be a measurable biologic spectrum of host responses.
- The disease should be reproducible experimentally.
- Preventing or modifying the host response should decrease or eliminate the expression of disease.
- Elimination of the putative cause should result in lower incidence of disease.
The web of causation is often used to describe modern disease problems where the presence or absence of disease is not just a matter of the agent being present or absent. Using this approach, the occurrence of disease is explained by a complex web of interacting factors involving host, agent, and environment.

A web of causation may be constructed to describe factors influencing the occurrence of rhinitis in swine. This helps researchers to conceptualise the complexity of the system in which this particular disease occurs. Many of the factors will interact and will have a different effect at varying exposure levels. Koch’s postulates do not provide a suitable mechanism for investigating this type of problem.

Much of the work of epidemiologists is aimed at uncovering components of the web of causation. Statistical analysis is often used to identify risk factors for a disease, that is, factors that increase the probability of disease occurring. However, we must also appreciate that statistical association does not prove causality. A statistical association is very likely between sunglasses, ice-cream and drowning (all are a function of outside temperature) but you would not claim that eating ice-cream or wearing sunglasses causes death by drowning.

Figure 6: Descriptive epidemiology of Severe Acute Respiratory Syndrome in Hong Kong, February to April, 2003. A: Temporal pattern of SARS epidemic in Hong Kong by cluster of infection. B: Spatial distribution of population of Hong Kong and district-specific incidence (per 10 000 population) over course of epidemic to date. C: Age distribution of residents of Hong Kong and age-specific incidence (per 10 000 population) over course of epidemic to date. D: Detail of temporal pattern for Amoy Gardens cluster, according to day of admission, and fitted gamma distribution. Reproduced from Donnelly et al. (2004).
If a statistical association is found between a factor and a disease it is important to determine if that factor may be causal. This is done by considering each of the criteria of Evan’s unified concept of causation. This is where the endless process of scientific inference plays such a critical role. Develop a hypothesis and test it: if it is found to be incorrect, modify the hypothesis and test it again.

### 1.4 Historical examples in the development of epidemiology

#### 1.4.1 Ignas Semmelweis

Ignas Semmelweis was director of the Viennese Maternity Hospital in the 1840s. Two clinics made up the Viennese Maternity Hospital: one run by midwives and the second run by doctors and medical students. Perinatal mortality due to puerperal fever (septic metritis) was 3 – 5 times higher in the doctor-run clinic compared with the midwife-run clinic with this relationship remaining constant over a 6 year period. In the 1840s prevailing medical opinion was that disease was essentially an act of God. In an attempt to uncover the reasons for the high mortality rate in the doctor-run clinic Semmelweis performed a series of observational studies and arrived at the following conclusions:

- Mothers were becoming ill within 24 – 36 hours of delivery.
- Illness seemed to be associated with mothers that received a manual examination.
- Doctors and medical students were in the habit of performing necropsies (ungloved) in the morning and then coming straight over to the maternity clinic in the afternoon and performing vaginal examinations with unwashed hands.
• Midwives did not perform necropsies.

Semmelweis instituted a program of washing hands with chlorinated water upon entry to the maternity ward. This was implemented after much argument and opposition and at a time when hygiene was considered to be unrelated to disease. Death rates in the doctor-run clinic decreased immediately.

1.4.2 John Snow

A major outbreak of cholera occurred in a small area of central London (Golden Square) in the 1840s with 500 fatal attacks occurring within a 10-day period. Snow spent much of his life investigating cholera and collected a massive amount of data from this outbreak. He found that most of the affected group had collected their drinking water from a single water pump (the Broad Street pump). Snow applied pressure on the local council to remove the handle from the Broad Street pump, hypothesising correctly that contaminated water from this pump was the source of infection. Snow subsequently provided further evidence of the association between contaminated drinking water and cholera with an eloquent study investigating the relationship between companies supplying household water and cholera rates. During the 1840s London had numerous water companies that competed to supply household water. Customers chose water companies largely at random. One company drew water only from a site on the Thames River above all London sewerage outlets. The others drew water all along the river. Snow showed that those households that used the upriver water company had lower rates of cholera compared with those that used the other companies. This supported Snow’s hypothesis of water borne contamination causing the disease.

It was not until more than 30 years later that the causative organism of cholera (\textit{Vibrio cholerae}) was isolated.
2 Measures of health

By the end of this unit you should be able to:

• Differentiate between ratios, proportions and rates.
• Describe the terms incidence and prevalence, and use them appropriately.
• Describe the difference between risk and rate as applied to measures of incidence.

One of the most fundamental tasks in epidemiological research is to quantify the occurrence of disease. This can be done by counting the number of affected individuals however, to compare levels of disease among groups of individuals, time frames and locations, we need to consider counts of cases in context of the size of the population from which those cases arose.

A ratio defines the relative size of two quantities expressed by dividing one (numerator) by the other (denominator). Proportions, odds, and rates are ratios. Say we have a herd of 100 cattle and 58 are found to be diseased. The odds of disease in this herd is 58:42 or 1.4 to 1.

A proportion is a fraction in which the numerator is included in the denominator. Say we have a herd of 100 cattle and 58 are found to be diseased. The proportion of diseased animals in this herd is $\frac{58}{100} = 0.58 = 58\%$.

A rate is derived from three pieces of information: (1) a numerator: the number of individuals diseased or dead, (2) a denominator: the total number of animals (or animal time) in the study group and/or period; and (3) a specified time period. To continue the above example, we might say that the rate of disease in our herd over a 12-month period was 58 cases per 100 cattle.

The term morbidity is used to refer to the extent of disease or disease frequency within a defined population. Two important measures of morbidity are prevalence and incidence. As epidemiologists we must take care to use these terms correctly.

2.1 Prevalence

Strictly speaking, prevalence refers to the number of cases of a given disease or attribute that exists in a population at a specified time. Prevalence risk is the proportion of a population that has a specific disease or attribute at a specified point in time. Many authors use the term ‘prevalence’ when they really mean prevalence risk, and these notes will follow this convention.

$$\text{Prevalence} = \frac{\text{Number of existing cases}}{\text{Size of population}}$$  \hspace{1cm} (2.1)

Prevalence can be interpreted as the probability of an individual from a population having a disease at a specified point in time.
Two types of prevalence are reported in the epidemiological literature: (1) **point prevalence** equals the number of disease cases in a population at a single point in time (a snapshot), (2) **period prevalence** equals the point prevalence at the beginning of a study period plus the number of new cases that occurred during the remainder of the study period.

In 1944 the cities of Newburgh and Kingston, New York agreed to participate in a study of the effects of water fluoridation for prevention of tooth decay in children (Ast and Schlesinger 1956). In 1944 the water in both cities had low fluoride concentrations. In 1945, Newburgh began adding fluoride to its water — increasing the concentration ten-fold while Kingston left its supply unchanged. To assess the effect of water fluoridation on dental health, a survey was conducted among school children in both cities during the 1954 – 1955 school year. One measure of dental decay in children 6 – 9 years of age was whether at least one of a child’s 12 deciduous cuspids or first or second deciduous molars was missing or had clinical or X-ray evidence of tooth decay.

Of the 216 first-grade children examined in Kingston, 192 had evidence of tooth decay. Of the 184 first-grade children examined in Newburgh 116 had evidence of tooth decay. Assuming complete survey coverage, there were 192 prevalent cases of tooth decay among first-grade children in Kingston at the time of the study. The prevalence of tooth decay was $\frac{192}{216} = 89\%$ in Kingston and $\frac{116}{184} = 63\%$ in Newburgh.


### 2.2 Incidence

Incidence measures how frequently initially susceptible individuals become disease cases as they are observed over time. An incident case occurs when an individual changes from being susceptible to being diseased. The count of incident cases is the number of such events that occur in a defined population during a specified time period. There are two ways to express incidence: **incidence risk** and **incidence rate**.

#### 2.2.1 Incidence risk

Incidence risk (also known as cumulative incidence) is the proportion of initially susceptible individuals in a population who become new cases during a defined time period.

$$\text{Incidence risk} = \frac{\text{Number of incident cases}}{\text{Number of individuals initially at risk}} \quad (2.2)$$

The defined time period may be arbitrarily fixed (e.g. 5-year incidence risk of arthritis) or it may vary among individuals (e.g. the lifetime incidence risk of arthritis). In an investigation of a localised epidemic the defined time period may be simply defined as the duration of the epidemic.

- Individuals have to be disease-free at the beginning of the observation period to be included in the numerator or denominator of this calculation.
The time period to which the risk applies must be specified.

- The quantity is dimensionless and ranges from 0 to 1.

Last year a herd of 121 cattle were tested for tuberculosis using the tuberculin test and all tested negative. This year the same 121 cattle were tested and 25 tested positive. The incidence risk would then be 21 cases per 100 cattle for the 12-month period. We can also say that the risk of an animal becoming positive to the tuberculin test for the 12-month period was 21%. This is an expression of average risk applied to an individual (but estimated from the population).

Populations at risk can be either closed or open. A closed population has no additions during the course of the study and no or few losses to follow-up. An open population is where individuals are recruited (e.g. as births or purchases) and leave (e.g. as sales or deaths) throughout the course of the study period. Incidence risk can be measured directly when the population is closed and all subjects are followed for the entire study period. When the population is open incidence risk cannot be measured directly, but can be estimated by making one of the following adjustments to the denominator:

- Denominator = population size at the mid-point of the study period.
- Denominator = \[N_{\text{start}} + \frac{1}{2}N_{\text{new}}\] - \[\frac{1}{2}N_{\text{lost}}\]
- Denominator = \[N_{\text{start}} + \frac{1}{2}N_{\text{new}}\] - \[\frac{1}{2}(N_{\text{lost}} + N_{\text{cases}})\]. This approach assumes that only one case of disease is considered per individual.

### 2.2.2 Incidence rate

Incidence rate (also known as incidence density) is the number of new cases of disease that occur per unit of individual time at risk, during a defined time period. The denominator of incidence rate is measured in units of animal (or person) time.

\[
\text{Incidence rate} = \frac{\text{Number of incident cases}}{\text{Amount of at-risk experience}} \tag{2.3}
\]

Because the denominator is expressed in units of animal- or person-time at risk those individuals that are withdrawn or are lost to follow up are easily accounted-for. Consider a study of clinical mastitis in five cows over a 12-month period, as shown in Table 1.

On the basis of the data presented in Table 1, the incidence rate of clinical mastitis for the 12-month period is 5 cases per 825 cow-days at risk (equivalent to 2.2 cases of clinical mastitis per cow-year at risk). Note that incidence rate:

- Accounts for individuals that enter and leave the population throughout the period of study.
### Table 1: Hypothetical mastitis data

<table>
<thead>
<tr>
<th>ID</th>
<th>Details</th>
<th>Events</th>
<th>Days at risk</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Calve 01 Aug, mastitis 15 Aug, mastitis 15 Sep, mastitis 15 Oct, sold 15 Nov</td>
<td>3</td>
<td>106</td>
</tr>
<tr>
<td>2</td>
<td>Calve 01 Aug, mastitis 15 Nov, dry off 15 May</td>
<td>1</td>
<td>365</td>
</tr>
<tr>
<td>3</td>
<td>Purchased 01 Dec, mastitis 01 Jan, dry off 15 May</td>
<td>1</td>
<td>243</td>
</tr>
<tr>
<td>4</td>
<td>Calve 01 Aug, Sold 16 Nov</td>
<td>0</td>
<td>107</td>
</tr>
<tr>
<td>5</td>
<td>Calve 01 Oct, Died 05 Oct</td>
<td>0</td>
<td>4</td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td></td>
<td><strong>5</strong></td>
<td><strong>825</strong></td>
</tr>
</tbody>
</table>

- Can account for multiple disease events in the same individual (e.g. cow 1 in Table 1).

To calculate incidence rate correctly, it is necessary to record detailed information for each individual under study. When this is not possible time at risk can be estimated as follows:

- **Denominator** = population size at the mid-point of the study period $\times$ length of study period.

- **Denominator** = $\left[ N_{\text{start}} + \frac{1}{2} N_{\text{new}} \right] - \left[ \frac{1}{2} N_{\text{lost}} \right] \times$ length of study period.

- **Denominator** = $\left[ N_{\text{start}} + \frac{1}{2} N_{\text{new}} \right] - \left[ \frac{1}{2} \left( N_{\text{lost}} + N_{\text{cases}} \right) \right] \times$ length of study period. This approach assumes that only one case of disease is considered per individual.

Gardner et al (1999) studied on-the-job back sprains and strains among 31,076 material handlers employed by a large retail merchandising chain. Payroll data for a 21-month period during 1994 – 1995 were linked with job injury claims. A total of 767 qualifying back injuries occurred during 54,845,247 working hours, yielding an incidence rate of 1.40 back injuries per 100,000 worker-hours.


### 2.2.3 The relationship between prevalence and incidence

Table 2 compares the main features of the three measures of disease frequency that we have described.

Figure 8 provides a worked example for calculating the various measures of disease frequency. The example is based on a herd of 10 animals which are all disease-free at the beginning of the observation period and followed for a 12-month period. Disease status is assessed at monthly intervals.

Providing incidence rate is constant, incidence risk for a defined study period can be estimated from incidence rate as follows:
Table 2: Comparison of prevalence, incidence risk, and incidence rate.

<table>
<thead>
<tr>
<th>Item</th>
<th>Prevalence</th>
<th>Incidence risk</th>
<th>Incidence rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>Numerator</td>
<td>All cases counted on a single occasion</td>
<td>New cases occurring during a specified follow-up period</td>
<td>New cases occurring during a specified follow-up period</td>
</tr>
<tr>
<td>Denominator</td>
<td>All individuals examined - cases and non-cases</td>
<td>All susceptible individuals present at the start of the study</td>
<td>Sum of time periods during which all individuals could have developed disease</td>
</tr>
<tr>
<td>Time</td>
<td>Single point or period</td>
<td>Defined period</td>
<td>Measured for each individual from beginning of study until disease event</td>
</tr>
<tr>
<td>Study</td>
<td>Cross-sectional</td>
<td>Prospective cohort study</td>
<td>Prospective cohort study</td>
</tr>
<tr>
<td>Interpretation</td>
<td>Probability of having disease at a point in time</td>
<td>Risk of developing disease over a specified period</td>
<td>How quickly new cases develop over a specified period</td>
</tr>
</tbody>
</table>

- Closed population: incidence risk = incidence rate \times \text{length of study period}.

- Open population: incidence risk = 1 - \exp(-\text{incidence rate} \times \text{length of study period}).

- Open population (when the study period is short): incidence risk \sim \text{incidence rate} \times \text{length of study period}.

Providing incidence rate is constant, prevalence can be estimated from incidence rate as follows:

- \text{Prevalence} = \frac{\text{incidence rate} \times \text{duration of disease}}{\text{incidence rate} \times \text{duration of disease} + 1}.

The incidence rate of disease is estimated to be 0.006 cases per cow-day at risk. The mean duration of disease is 7 days. The estimated prevalence of disease is \( (0.006 \times 7) / (0.006 \times 7 + 1) = 0.041 \). The estimated prevalence is 4.1 cases per 100 cows.

2.3 Other measures of health

2.3.1 Attack rates

Attack rates are usually used in outbreak situations where the period of risk is limited and all cases arising from exposure are likely to occur within the risk period. Attack rate is defined as the number of cases divided by the number of individuals exposed. ‘Attack risk’ would be a more precise way to describe this parameter.
2.3.2 Secondary attack rates

Secondary attack rates are used to describe infectiousness. The assumption is that there is spread of an agent within an aggregation of individuals (e.g. a herd or a family) and that not all cases are a result of a common-source exposure. Secondary attack rates are the number of cases at the end of the study period less the number of initial (primary) cases divided by the size of the population that were initially at risk.

2.3.3 Mortality

Mortality risk (or rate) is an example of incidence where death is the outcome of interest. Cause-specific mortality risk is the incidence risk of fatal cases of a particular disease in the population at risk of death from that disease. The denominator includes both prevalent cases of the disease (that is, the individuals that haven’t died yet) as well as
individuals who are at risk of developing the disease.

2.3.4 Case fatality

Case fatality risk (or rate) refers to the incidence of death among individuals who develop the disease. Case fatality risk reflects the prognosis of disease among cases, while mortality reflects the burden of deaths from the disease in the population as a whole.

2.3.5 Proportional mortality

As its name implies, proportional mortality is simply the proportion of all deaths that are due to a particular cause for a specified population and time period:

$$\text{Proportional mortality} = \frac{\text{Number of deaths from the disease}}{\text{Number of deaths from all causes}}$$ (2.4)

2.4 Adjusted measures of health

Adjusted rates are used when we want to compare the level of disease in different populations. In human medicine, because the occurrence of many health conditions is related to age, it is common to adjust populations on the basis of age. In veterinary medicine age, breed, and production type (e.g. beef-dairy) are commonly used adjustment variables.

If we have two colonies of mice and observe them for one day we might find the mortality rate in the first colony is 10 per 1,000 and the mortality rate in the second colony is 20 per 1,000. We might initially think that this difference is due to a difference in management, but it might also transpire that the first colony is comprised of mainly young mice and the second colony is comprised of mainly older mice. The two colonies might be exactly the same in terms of standards of care and housing quality and the difference in mortality solely due to a difference in age composition of the two populations.

The age adjustment process removes differences in the age composition of two or more populations to allow comparisons between these populations to be made, independent of their age structure. For example, a county’s age-adjusted death rate is the weighted average of the age-specific death rates observed in that county, with the weights derived from the age distribution in an external population standard. Different standard populations have different age distributions and the choice will affect the resulting age-adjusted rate. If the age-adjusted rates for different counties are calculated with the same weights (that is, using the same population standard), the effect of any differences in the county’s age distributions is removed.

There are two methods for adjusting disease rates: direct adjustment and indirect adjustment.
2.4.1 Stratum-specific rates

- Calculation of stratum-specific rates is recommended before developing adjusted rates. This will identify whether or not the populations being compared show stratum-specific rates that are consistent. If the pattern is not consistent, use of stratum-specific rates, rather than adjusted rates, are recommended.

- Stratum-specific rates are recommended for comparing defined subgroups between or within populations when rates are strongly stratum-dependent.

- Stratum-specific rates are recommended when specific causal or protective factors or the prevalence of risk exposures are different for different levels of strata.

Only compare rates when the numerator and denominator (i.e. events and population) are defined consistently over time and place. Look for:

- Consistency in definition of event.

- Consistency of surveillance intensity over time.

- Consistency of surveillance intensity among areas.

- If comparing stratum-adjusted rates, compare rates that have been adjusted to the same standard population.

- When comparing age-specific rates, if the age categories are relatively large, it is important to consider the possibility of residual confounding by age.

Rates based on small numbers of events can fluctuate widely from year to year for reasons other than a true change in the underlying frequency of occurrence of the event. Calculation of rates is not recommended when there are fewer than five events in the numerator, because the calculated rate is unstable and exhibits wide confidence intervals. Small counts should be included, where possible, even if the rates are not reported, so that the counts can be combined into larger totals (for example, three or five year averages) which would be more stable.

- Directly and indirectly adjusted rates are recommended when making comparisons in the rates of age-related health events between different populations or for comparing trends in a given population over time.

- Age adjusted rates are essential for events that vary with age (e.g. cancer deaths), when comparing populations with different age distributions.

- Directly and indirectly adjusted rates should be used only for the purpose of comparison. Because an adjusted rate is based on an external standard population, it does not reflect the absolute frequency of the event in a population.
2.4.2 Direct adjustment

With direct adjustment the observed stratum-specific rates are known and an estimated population distribution is used as the basis for adjustment. A standard population structure is typically used: if we were stratifying by sex we might say that in a standard population 50% of the total population would be allocated to the male strata and 50% to the female strata. The choice of the standard population for direct adjustment is not crucial; however, where possible it is desirable to select a standard that is demographically sensible. The directly adjusted count for the $i^{th}$ strata is then:

$$\text{Directly adjusted count}_{i} = \text{STD P}_{i} \times \text{OBS R}_{i}$$

(2.5)

Where:

- $\text{STD P}_{i}$: the size of the standard population in the $i^{th}$ strata
- $\text{OBS R}_{i}$: the observed rate in the $i^{th}$ strata

Consider a study of leptospirosis seroprevalence in Scottish dogs, the details of which are shown in Table 3.

<table>
<thead>
<tr>
<th>City</th>
<th>Positive</th>
<th>Sampled</th>
<th>Seroprevalence</th>
</tr>
</thead>
<tbody>
<tr>
<td>Edinburgh</td>
<td>61</td>
<td>260</td>
<td>23%</td>
</tr>
<tr>
<td>Glasgow</td>
<td>69</td>
<td>251</td>
<td>27%</td>
</tr>
<tr>
<td>Total</td>
<td>130</td>
<td>511</td>
<td>25%</td>
</tr>
</tbody>
</table>

The crude prevalence data suggests that Glasgow has a slightly higher seroprevalence of leptospirosis amongst its dog population. However, what about the composition of the two populations that were studied? Male dogs are known to have a higher incidence rate for leptospirosis because of their sexual behaviour, and it might be that more male dogs were sampled in Glasgow. Sex-specific prevalence estimates (Table 4) confirm the role of population structure.

The confounding effect of sex can be removed by producing gender-adjusted prevalence estimates (Table 5). Direct adjustment involves adjusting the crude values to produce estimates which would be expected if the potentially confounding characteristics were similarly distributed in the two study populations.

Direct adjustment involves specifying the frequency of each level of a potential confounder (for example, sex) to produce a ‘standard population.’ In this example, we use a standard population comprised of 250 males and 250 females. The values for each study group are then weighted by the frequency of each level of the confounder.
Table 4: Seroprevalence of leptospirosis in urban dogs, stratified by city and sex.

<table>
<thead>
<tr>
<th>City</th>
<th>Positive</th>
<th>Sampled</th>
<th>Seroprevalence</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Male</td>
<td>Female</td>
<td>Male</td>
</tr>
<tr>
<td>Edinburgh</td>
<td>15</td>
<td>46</td>
<td>48</td>
</tr>
<tr>
<td>Glasgow</td>
<td>53</td>
<td>16</td>
<td>180</td>
</tr>
<tr>
<td>Total</td>
<td>68</td>
<td>62</td>
<td>228</td>
</tr>
</tbody>
</table>

Table 5: Directly adjusted seroprevalence of leptospirosis in urban dogs, stratified by city.

<table>
<thead>
<tr>
<th>City</th>
<th>Positive</th>
<th>Sampled</th>
<th>Seroprevalence</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Male</td>
<td>Female</td>
<td>Male</td>
</tr>
<tr>
<td>Edinburgh</td>
<td>0.31×250=77</td>
<td>0.22×250=55</td>
<td>250</td>
</tr>
<tr>
<td>Glasgow</td>
<td>0.29×250=72</td>
<td>0.22×250=55</td>
<td>250</td>
</tr>
<tr>
<td>Total</td>
<td>77+72=149</td>
<td>55+55=110</td>
<td>500</td>
</tr>
</tbody>
</table>

The directly adjusted prevalence estimates are similar which suggests the difference between the cities is due to the different sex structures of the two populations.

2.4.3 Indirect adjustment

With indirect adjustment the stratum-specific rates are unknown and a known population distribution is used as the basis for adjustment. Indirect adjustment provides an estimate of the expected number of cases, given the stratum-specific population size. It is usual to divide the observed number of disease cases by the expected number to yield a standardised morbidity/mortality ratio (SMR). The indirectly adjusted count for the \(i^{th}\) strata is:

\[
\text{Indirectly adjusted count}_i = \text{STD R}_i \times \text{OBS P}_i
\]  

Where:

\(\text{STD R}_i\): the standard rate in the \(i^{th}\) strata of the population
\(\text{OBS P}_i\): the observed population size in the \(i^{th}\) strata

We know that the prevalence of a given disease throughout a country is 0.01%. If we are presented with a region with 20,000 animals the expected number of cases of disease in this region will be 0.01% \(\times\) 20,000 = 2.

If the actual number of cases of disease in this region is 5, then the standardised mortality (morbidity) ratio is 5 \(\div\) 2 = 2.5. That is, there were 2.5 times more cases of disease in this region, compared with the number of cases we were expecting.
Figure 9: An example of the use of indirect standardisation used to describe the change in spatial distribution of disease risk over time. Choropleth maps of area-level standardised mortality ratios (SMRs) for bovine spongiform encephalopathy in British cattle 1986 – 1997, for (a) cattle born before the 18 July 1988 ban on feeding meat and bone meal to ruminants, and (b) cattle born between 18 July 1988 and 30 June 1997. The above maps show a shift in area-level risk over time (even though the incidence of BSE reduced markedly from 1988 to 1997). Reproduced from Stevenson et al. (2005).
3 Study design

By the end of this unit you should be able to:

- Describe the difference between descriptive and analytical epidemiological studies (giving examples of each).
- Describe the major features of randomised clinical trials, cohort studies, case-control studies, and cross-sectional studies.
- Describe the strengths and weaknesses of clinical trials, cohort studies, case-control studies, and cross-sectional studies.

A study generally begins with a research question. Once the research question has been specified the next step is to choose a study design. A study design is a plan for selecting study subjects and for obtaining data about them. Figure 10 shows the major types of epidemiological study designs. There are three main study types: (1) descriptive studies, (2) analytical studies, and (3) experimental studies.

**Figure 10:** Tree diagram outlining relationships between the major types of epidemiologic study designs.

Descriptive studies are those undertaken without a specific hypothesis. They are often the earliest studies done on a new disease in order to characterise it, quantify its frequency, and determine how it varies in relation to individual, place and time. Analytical studies are undertaken to identify and test hypotheses about the association between an exposure of interest and a particular outcome. Experimental studies are also designed to test hypotheses between specific exposures and outcomes — the major difference is that in experimental studies the investigator has direct control over the study conditions.

3.1 Descriptive studies

The hallmark of a descriptive study is that it is undertaken without a specific hypothesis.
3.1.1 Case reports

A case report describes some ‘newsworthy’ clinical occurrence, such as an unusual combination of clinical signs, experience with a novel treatment, or a sequence of events that may suggest previously unsuspected causal relationships. Case reports are generally reported as a clinical narrative.

<table>
<thead>
<tr>
<th>Trivier at al (2001) reported the occurrence of fatal aplastic anaemia in an 88 year-old man who had taken clopidogrel, a relatively new drug on the market that inhibits platelet aggregation. The authors speculated that his fatal illness may have been caused by clopidogrel and wished to alert other clinicians to a possible adverse effect of the drug.</th>
</tr>
</thead>
</table>

3.1.2 Cases series

Whereas a case report shows that something can happen once, a case series shows that it can happen repeatedly. A case series identifies common features among multiple cases and describes patterns of variability among them.

<table>
<thead>
<tr>
<th>After bovine spongiform encephalopathy (BSE) appeared in British cattle in 1987, there was concern that the disease might spread to humans. A special surveillance unit was set up to study Creutzfeld-Jacob disease (CJD), a rare and fatal progressive dementia that shares clinical and pathological features of BSE. In 1996 investigators at the unit described ten cases that met the criteria for CJD but had all occurred at unusually young ages, showed distinctive symptoms and, on pathological examination, had extensive prion protein plaques throughout the brain similar to BSE.</th>
</tr>
</thead>
</table>

3.1.3 Descriptive studies based on rates

Descriptive studies based on rates quantify the burden of disease on a population using incidence, prevalence, mortality or other measures of disease frequency. Most use data from existing sources (such as birth and death certificates, disease registries or surveillance systems). Descriptive studies can be a rich source of hypotheses that lead later to analytic studies.

<table>
<thead>
<tr>
<th>Schwarz et al (1994) conducted a descriptive epidemiological study of injuries in a predominantly African-American part of Philadelphia. An injury surveillance system was set up in a hospital emergency centre. Denominator information came from US census data. These authors found a high incidence of intentional interpersonal injury in this area of the city.</th>
</tr>
</thead>
</table>
3.2 Analytical studies

Analytical studies are undertaken to test a hypothesis. In epidemiology the hypothesis typically concerns whether a certain exposure causes a certain outcome — e.g. does cigarette smoking cause lung cancer?

The term exposure is used to refer to any trait, behaviour, environmental factor or other characteristic as a possible cause of disease. Synonyms for exposure are: potential risk factor, putative cause, independent variable, and predictor. The term outcome generally refers to the occurrence of disease. Synonyms for outcome are: effect, end-point, and dependent variable.

The hypothesis in an analytic study is whether an exposure actually causes an outcome (not merely whether the two are associated). Each of Hill’s criteria for causation are usually required to be met to support a case for causality, but probably the most important is that exposure must precede the outcome in time.

3.2.1 Ecological studies

In an ecological study the unit of analysis is a group of individuals (such as counties, states, cities, or census tracts) and summary measures of exposure and summary measures of outcome are compared. A key feature of ecological studies is that inference can only be made at the group level, not at the individual level. Ecological studies are relatively quick and inexpensive to perform and can provide clues to possible associations between exposures and outcomes of interest.

Yang et al (1998) conducted an ecological study examining the association between chlorinated drinking water and cancer mortality among 28 municipalities in Taiwan. The investigators found a positive association between the use of chlorinated drinking water and mortality from rectal, lung, bladder, and kidney cancer.


3.2.2 Cross-sectional studies

In a cross-sectional study a random sample of individuals from a population is taken at a point in time. Individuals included in the sample are examined for the presence of disease and their status with regard to the presence or absence of specified risk factors.

Cross sectional studies commonly involve surveys to collect data. Surveys range from simple one-page questionnaires addressing a single variable, to highly complex, multiple page designs. There is a whole sub-field of epidemiology associated with design, implementation and analysis of questionnaires and surveys.

Advantages: Cross-sectional studies are relatively quick to conduct and their cost is moderate, compared with other study designs.
Disadvantages: Cross-sectional studies cannot provide information on the incidence of disease in a population — only an estimate of prevalence. Difficult to investigate cause and effect relationships.

Anderson et al (1998) studied 4,063 children aged 8 to 16 years who had participated in the National Health and Nutrition Examination Survey to assess the relationship between television watching and body-mass index. At a single examination, each child was asked a series of questions about their usual amount of television viewing. Height, weight and a series of other body measurements were taken at the same time. Boys and girls who reported watching four or more hours of television per day had significantly greater body mass indexes than boys and girls who reported watching fewer than two hours of television per day.


3.2.3 Cohort studies

A cohort study involves comparing disease incidence over time between groups (cohorts) that are found to differ on their exposure to a factor of interest. Cohort studies can be distinguished as either prospective or retrospective (Figure [12]).

A prospective cohort study begins with the selection of two groups of non-diseased animals, one exposed to a factor postulated to cause a disease and the other unexposed. The groups are followed over time and their change in disease status is recorded during the study period.

A retrospective cohort study starts when all of the disease cases have been identified. The history of each study participant is carefully evaluated for evidence of exposure to the agent under investigation.

Advantages: Because subjects are monitored over time for disease occurrence, cohort studies provide estimates of the absolute incidence of disease in exposed and non-exposed
individuals. By design, exposure status is recorded before disease has been identified. In most cases, this provides unambiguous information about whether exposure preceded disease. Cohort studies are well-suited for studying rare exposures. This is because the relative number of exposed and non-exposed persons in the study need not necessarily reflect true exposure prevalence in the population at large.

**Disadvantages:** Prospective cohort studies require a long follow-up period. In the case of rare diseases large groups are necessary. Losses to follow-up can become an important problem. Often quite expensive to run.

To assess the possible carcinogenic effects of radio-frequency signals emitted by cellular telephones, Johansen et al (2001) conducted a retrospective cohort study in Denmark. Two companies that operate cellular telephone networks provided names and addresses for all 522,914 of their clients for the period 1982 to 1995. The investigators matched these records to the Danish Central Population Register. After cleaning the data 420,095 cellular telephone subscribers remained and formed the exposed cohort. All other Danish citizens during the study years became the unexposed cohort. The list of exposed and unexposed individuals were then matched with the national cancer registry. The resulting data allowed calculation of cancer incidence rates.

Overall, 3,391 cancers had occurred among cellular telephone subscribers, compared with 3,825 cases expected based on age, gender, and calendar-year distribution of their person time at risk.


### 3.2.4 Case-control studies

A case-control study involves comparing the frequency of past exposure between cases who develop the disease (or other outcome of interest) and controls chosen to reflect the frequency of exposure in the underlying population at risk. Figure 13 shows a diagram of the case-control design.
Advantages: Case-control studies are an efficient method for studying rare diseases. Because subjects have experienced the outcome of interest at the start of the study, case-control studies are quick to run and are considerably cheaper than other study types.

Disadvantages: Case-control studies cannot provide information on the disease incidence in a population. The study is reliant on the quality of past records or recollection of study participants. It can also be very difficult to ensure an unbiased selection of the control group and, as a result, the representativeness of the sample selection process is difficult to guarantee.

Muscat et al (2000) sought to test the hypothesis that cellular telephone use affects the risk of brain cancer. From 1994 to 1998 at five academic medical centres in the USA they recruited 469 cases aged 18 to 80 years with newly diagnosed cancer originating in the brain. Controls (n = 422) were inpatients without brain cancer at those hospitals, excluding those with leukaemia or lymphoma. Controls were sampled to match the cases on age, sex, race and month of admission. Each case and control was then interviewed about any past subscription to a cellular telephone service. Overall 14.1% of cases and 18.0% of controls reported ever having had a subscription for a cellular telephone service. After adjusting for age, sex, race, education, study centre, and month and year of interview, the risk of developing brain cancer in a cellular telephone user was estimated to be 0.85 (95% CI 0.6 – 1.2) times as great as in a non-user.


3.2.5 Hybrid study designs

A nested case-control study is similar to a cohort study with the key difference that a sample of non-cases are selected for analysis (rather than the entire cohort, as in the
An Introduction to Veterinary Epidemiology

A panel study combines the features of cross-sectional and a prospective cohort designs. It can be viewed as a series of cross-sectional studies conducted on the same subjects.

**Figure 14:** Schematic diagram of a nested case-control study.

**Advantages:** Nested case-control studies are useful when it is either too costly or not feasible to perform additional analyses on an entire cohort (e.g., if collection of specimens and laboratory analysis of specimens is expensive). Compared with standard case-control studies, nested studies: 1) can utilise exposure and confounder data originally collected before the onset of the disease, thus reducing potential recall bias and temporal ambiguity, and 2) include cases and controls drawn from the same cohort, decreasing the likelihood of selection bias. The nested case-control study is thus considered a strong observational study, comparable to its parent cohort study in the likelihood of an unbiased association between an exposure and an outcome.

**Disadvantages:** A concern, usually minor, is that the remaining nondiseased persons from whom the controls are selected when it is decided to do the nested study, may not be fully representative of the original cohort due to death or losses to follow-up.

To determine if Helicobacter pylori infection was associated with the development of gastric cancer, Parsonnet et al (1991) identified a cohort of 128,992 persons who had been followed since the mid-1960s. Of the original cohort, 189 patients developed gastric cancer. The investigators carried out a nested case-control study by selecting all of the 189 gastric cancer patients as cases and another 189 cancer-free individuals from the same cohort as controls. *H. pylori* infection status was determined using serum obtained at the beginning of the follow-up period. All total of 84% of the confirmed gastric cancer cases had been infected previously with *H. pylori*, while only 61% of the controls had been infected. This indicated a positive association between *H. pylori* infection and gastric cancer risk.

(the panel) at successive time intervals (sometimes referred to as waves). This design allows investigators to relate changes in one variable to changes in other variables over time.

A repeated survey is a series of cross-sectional studies performed over time on the same study population, but each is sampled independently. Whereas panel studies follow the same individuals from survey to survey, repeated surveys follow the same study population (which may differ in composition from one survey to the next). Repeated surveys are useful for identifying overall trends in health status over time.

### 3.3 Experimental studies

#### 3.3.1 Randomised clinical trials

The randomised clinical trial is the epidemiologic design that most closely resembles a laboratory experiment. The major objective is to test the possible effect of a therapeutic or preventive intervention. The design’s key feature is that a formal chance mechanism is used to assign participants to either the treatment or control group. Subjects are then followed over time to measure one or more outcomes, such as the occurrence of disease. All things being equal, results from randomised trials offer a more solid basis for inference of cause and effect than results obtained from any other study design.

**Advantages**: Randomisation generally provides excellent control over confounding, even by factors that may be hard to measure or that may be unknown to the investigator.
Disadvantages: For many exposures it may not be ethical or feasible to conduct a clinical trial (e.g. exposure to pollution). Expensive. Impractical if long periods of follow-up required.

Bacterial vaginosis affects an estimated 800,000 pregnant women each year in the USA and has been found to be associated with premature birth and other pregnancy complications. To determine whether treatment with antibiotics could reduce the incidence of adverse pregnancy outcomes, Carey et al (2000) screened 29,625 pregnant women to identify 1953 who had bacterial vaginosis, met certain other eligibility criteria, and consented to participate. Women were randomly assigned to receive either: (1) two 2 gram doses of metronidazole, or (2) two doses of a similar-appearing placebo.

Bacterial vaginosis resolved in 78% of women in the treatment group, but in only 37% of women in the placebo group. Pre-term labour, postpartum infections in the mother or infant, and admission to the neonatal intensive care unit were equally common in both groups.


3.3.2 Community trials

Instead of randomly assigning individuals to treatment or control groups, community trials assign interventions to entire groups of individuals. In the simplest situation one group (community) receives the treatment and another serves as a control.

3.4 Comparison of major study designs

Cohort studies involve enumeration of the denominator of the disease measure (individual time at risk) while case-control studies only sample from the denominator. Cohort studies therefore provide an estimate of incidence and risk whereas case-control studies can only estimate ratios. Prospective cohort studies provide the best evidence for the presence of cause-effect relationships, because any putative cause has to be present before disease occurs. Since these study designs are based on observation within a largely uncontrolled environment it is possible that there are still other unmeasured factors which produce cause-effect relationships that might be identified. The prospective cohort study is inefficient for studying rare diseases, which is a particular strength of the case-control study. A carefully designed cross-sectional study is more likely to be representative of the population than a case-control study.
Table 6: Comparison of the features of the cohort, case-control and cross-sectional study design.

<table>
<thead>
<tr>
<th>Criteria</th>
<th>Cohort</th>
<th>Case-control</th>
<th>Cross-sectional</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sampling</td>
<td>Separate samples of exposed and non-exposed individuals</td>
<td>Separate sampled of diseased and non-diseased individuals</td>
<td>Random sample of study population</td>
</tr>
<tr>
<td>Time</td>
<td>Usually prospective (but may be retrospective)</td>
<td>Usually retrospective</td>
<td>Single point</td>
</tr>
<tr>
<td>Causality</td>
<td>Causality through evidence of temporality</td>
<td>Preliminary causal hypothesis</td>
<td>Association between disease and risk factor</td>
</tr>
<tr>
<td>Risk</td>
<td>Incidence density, cumulative incidence</td>
<td>None</td>
<td>Prevalence</td>
</tr>
<tr>
<td>Comparison of risks</td>
<td>Relative risk, odds ratio</td>
<td>Odds ratio</td>
<td>Relative risk, odds ratio</td>
</tr>
</tbody>
</table>
4 Measures of association

By the end of this unit you should be able to:

- Given disease count data, construct a $2 \times 2$ table and explain how to calculate the following measures of association: relative risk, odds ratio, attributable rate, and attributable fraction.
- Interpret the following measures of association: relative risk, odds ratio, attributable rate, and attributable fraction.
- Describe those situations where relative risk is not a valid measure of association between exposure and outcome.

Risk is the probability that an event will happen. A characteristic or factor that influences whether or not an event occurs, is called a risk factor.

- Worn tyres are a risk factor for motor vehicle accidents.
- High blood pressure is a risk factor for coronary heart disease.
- Vaccination is a protective risk factor in that it usually reduces the risk of disease.

If we identify those risk factors that are causally associated with an increased likelihood of disease and those causally associated with a decreased likelihood of disease, then we are in a good position to make recommendations about health management. Much of epidemiological research is concerned with estimating and quantifying risk factors for disease.

Associations between putative risk factors (exposures) and an outcome (usually a disease) can be investigated using analytical observational studies. Consider a study where subjects are disease free at the start of the study and all are monitored for disease occurrence for a specified time period. If both exposure and outcome are binary variables (yes or no), the results can be presented as a $2 \times 2$ table.

<table>
<thead>
<tr>
<th></th>
<th>Diseased</th>
<th>Non-diseased</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Exposed</td>
<td>$a$</td>
<td>$b$</td>
<td>$a+b$</td>
</tr>
<tr>
<td>Non-exposed</td>
<td>$c$</td>
<td>$d$</td>
<td>$c+d$</td>
</tr>
<tr>
<td>Total</td>
<td>$a+c$</td>
<td>$b+d$</td>
<td>$a+b+c+d = n$</td>
</tr>
</tbody>
</table>

Based on data presented in this ‘standard’ format, various measures of association can be calculated. These fall into three main categories: (1) measures of strength, (2) measures of effect, and (3) measures of total effect. To calculate these parameters, it helps to work out some summary parameters:

Incidence risk in the exposed population: $R_E = a/(a+b)$
Incidence risk in the non-exposed population: $R_O = c/(c+d)$
Incidence risk in the total population: $R_{Total} = (a+c)/(a+b+c+d)$
Odds of disease in the exposed population: \( O_E = \frac{a}{b} \)
Odds of disease in the non-exposed population: \( O_O = \frac{c}{d} \)

Observed associations are not always causal and/or may be estimated with bias. The interpretation of the measures of association described below assumes that relationships are causal and have been estimated without bias.

4.1 Measures of strength

4.1.1 Risk ratio

Where incidence risk has been measured, the risk ratio is defined as the ratio of the risk of disease (i.e. the incidence risk) in the exposed group to the risk of disease in the unexposed group. Using the notation defined above, risk ratio (RR) is calculated as:

\[
RR = \frac{R_E}{R_O}
\]

(4.1)

The risk ratio provides an estimate of how many times more likely exposed individuals are to experience disease, relative to non-exposed individuals. If the risk ratio equals 1, then the risk of disease in the exposed and non-exposed groups are equal. If the risk ratio is greater than 1, then exposure increases the risk of disease with greater departures from 1 indicative of a stronger effect. If the risk ratio is less than 1, exposure reduces the risk of disease and exposure is said to be protective. Risk ratio cannot be estimated in case-control studies, as these studies do not allow calculation of risks. Odds ratios are used instead — see below.

Risk ratios range between 0 and infinity.

4.1.2 Incidence rate ratio

In a study where incidence rate has been measured rather than incidence risk, the incidence rate ratio (IRR), also known as the rate ratio, can be calculated. This is the ratio of the incidence rate in the exposed group to that in the non-exposed group. Incidence rate ratio is interpreted in the same way as risk ratio.

The term relative risk is used as a synonym for both risk ratio and incidence rate ratio.

4.1.3 Odds ratio

The odds ratio is the odds of disease, given exposure. The odds ratio (OR) is an estimate of risk ratio and is interpreted in the same way. The odds ratio is calculated as:
When the number of cases of disease is low relative to the number of non-cases (i.e. the disease is rare), then the odds ratio approximates risk ratio. If the incidence of disease is relatively low in both exposed and non-exposed individuals, then $a$ will be small relative to $b$ and $c$ will be small relative to $d$. As a result:

$$RR = \frac{a/(a+b)}{c/(c+d)} \approx \frac{a/b}{c/d} = \frac{ad}{bc} = OR$$  \hspace{1cm} (4.3)$$

### 4.2 Measures of effect in the exposed population

#### 4.2.1 Attributable risk (rate)

Attributable risk (or rate) is defined as the increase or decrease in the risk (or rate) of disease in the exposed group that is attributable to exposure. Attributable risk (unlike risk ratio) describes the absolute quantity of the outcome measure that is associated with the exposure. Using the notation defined above, attributable risk (AR) is calculated as:

$$AR = R_E - R_O$$  \hspace{1cm} (4.4)$$

#### 4.2.2 Attributable fraction

Attributable fraction (also known as the attributable proportion in exposed subjects) is the proportion of disease in the exposed group that is due to exposure. Using the notation defined above, attributable fraction (AF) is calculated as:

$$AF = \frac{(R_E - R_O)}{R_E} = \frac{(RR - 1)}{RR}$$  \hspace{1cm} (4.5)$$

For case-control studies, attributable fraction can be approximated:

$$AF_{est} = \frac{(O_E - O_O)}{O_E} = \frac{(OR - 1)}{OR}$$  \hspace{1cm} (4.6)$$

This approximation is appropriate if: (1) disease incidence is low, or (2) odds ratios were derived from a case-control study where incidence density sampling was used.

In vaccine trials, vaccine efficacy is defined as the proportion of disease prevented by the vaccine in vaccinated individuals (equivalent to the proportion of disease in unvaccinated
individuals due to not being vaccinated), which is the attributable fraction. A casecontrol study investigating the effect of oral vaccination on the presence or absence of rabies in foxes was conducted. The following results were obtained:

The odds of rabies in the unvaccinated group was 2.3 times the odds of rabies in the vaccinated group (OR = 2.30). Fifty six percent of rabies cases in unvaccinated foxes was due to not being vaccinated (AFest = 0.56).

4.3 Measures of effect in the total population

4.3.1 Population attributable risk (rate)

Population attributable risk (or rate) is the increase or decrease in risk (or rate) of disease in the population that is attributable to exposure. Using the notation defined above, population attributable risk (PAR) is calculated as:

\[
PAR = R_{Total} - R_O
\]

(4.7)

4.3.2 Population attributable fraction

Population attributable fraction (also known as the aetiologic fraction) is the proportion of disease in the population that is due to the exposure. Using the notation defined above, the population attributable fraction (PAF) is calculated as:

\[
PAF = \frac{(R_{Total} - R_O)}{R_{Total}}
\]

(4.8)

Methods are available to estimate PAF using data from case-control studies.

A cross sectional study investigating the relationship between dry cat food (DCF) and feline urologic syndrome (FUS) was conducted. The following results were obtained:

<table>
<thead>
<tr>
<th></th>
<th>FUS +</th>
<th>FUS -</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>DCF +</td>
<td>13</td>
<td>2163</td>
<td>2176</td>
</tr>
<tr>
<td>DCF -</td>
<td>5</td>
<td>3349</td>
<td>3354</td>
</tr>
<tr>
<td>Total</td>
<td>18</td>
<td>5512</td>
<td>5530</td>
</tr>
</tbody>
</table>
The incidence risk of FUS in the DCF+ group was 5.97 cases per 1000. The incidence risk of FUS in the DCF- group was 1.49 cases per 1000. The incidence risk of FUS in DCF exposed cats was 4.01 times greater than the incidence risk of FUS in DCF- cats (RR = 4.0).

The incidence risk of FUS in DCF+ cats that may be attributed to DCF is 4.5 per 1000 (AR = 0.0045). In DCF+ cats 75% of FUS is attributable to DCF (AF = 0.75).

The incidence risk of FUS in the cat population that may be attributed to DCF is 1.8 per 1000. That is, we would expect the risk of FUS to decrease by 1.8 cases per 1000 if DCF were not fed (PAR = 0.0018). Fifty-four percent of FUS cases in the cat population are attributable to DCF (PAF = 0.54).

### 4.4 Using the appropriate measure of effect

Table 7 outlines which measures of effect are appropriate for each of the three major study designs (case-control, cohort and cross-sectional studies).

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Case-control</th>
<th>Cohort</th>
<th>Cross-sectional</th>
</tr>
</thead>
<tbody>
<tr>
<td>Measures of strength:</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>RR</td>
<td>No</td>
<td>Yes</td>
<td>Yes (prevalence RR)</td>
</tr>
<tr>
<td>IRR</td>
<td>No</td>
<td>Yes</td>
<td>No</td>
</tr>
<tr>
<td>OR</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes (prevalence OR)</td>
</tr>
<tr>
<td>Measures of effect:</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>AR</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>AF</td>
<td>No</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>AF(est)</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>Measures of effect in population (total effect):</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>PAR</td>
<td>No</td>
<td>Yes*</td>
<td>Yes</td>
</tr>
<tr>
<td>PAF</td>
<td>No</td>
<td>Yes*</td>
<td>Yes</td>
</tr>
<tr>
<td>PAF (est)</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
</tbody>
</table>

* If an estimate of the prevalence of exposure or disease incidence for the population is available from another source.

Members of the public often have a poor understanding of relative and absolute risk. A case in point was a recent news item describing the results of a study of risk factors for leukaemia in children (Draper et al. 2005). Children who lived within 200 metres of high voltage lines at birth had a 70% higher incidence risk of leukaemia compared with those that lived 600 metres or more away. While the facts were correctly reported, the interpretation of the scientific evidence was misguided. If the incidence risk of leukaemia in the general population is around 1 in 20,000 a 70% increase elevates this to around 2 cases per 20,000 (a very minor increase in absolute terms).
Figure 16: Newspaper headline warning of the risk of leukaemia associated with living close to high-voltage electricity lines. Source: The Dominion Post (Wellington, New Zealand) Saturday 4 June 2005.
5 Statistical inference

Experiments and observational studies are carried out to provide data to answer scientific questions, that is, to test hypotheses.

- Do workers in cotton mills have reduced lung function compared with a control group?
- Is a course of exercises beneficial to men suffering from chronic lung disease?

Data on these two questions may be obtained by carrying out an epidemiological study and a randomised controlled trial respectively. The data then have to be analysed in such a way as to answer the original question. This process is called hypothesis testing. The general principles of hypothesis testing are:

- Formulate a null hypothesis that the effect to be tested does not exist.
- Collect data.
- Calculate the probability (P) of these data occurring if the null hypothesis were true.
- If P is large, the data are consistent with the null hypothesis. We conclude that there is no strong evidence that the effect being tested exists (this is not the same as saying that the null hypothesis is true — it may be false but the study was not large enough to detect the departure from the null hypothesis).
- If P is small, we reject the null hypothesis. We conclude that there is a statistically significant effect.

The dividing line between ‘large’ and ‘small’ P values is called the significance level \( \alpha \) (alpha). Usually \( \alpha \) is chosen as 0.05, 0.01, or 0.001 and a significant result is indicated by ‘\( P < 0.05 \)’ or ‘significant at the \( \alpha \) level of 0.05’. On the other hand, \( P > 0.05 \) is usually regarded as not statistically significant (NS).

Note however that when P is small there is in fact a choice of two interpretations:

1. The null hypothesis is true and an event of low probability has occurred by chance.
2. The null hypothesis is untrue and can therefore be rejected in favour of the alternative hypothesis that there actually is an effect.
In the cotton mill example above, the null hypothesis would be that workers in cotton mills have the same lung function as controls. Only if the data appeared inconsistent with this null hypothesis would we feel confident to claim that there was evidence of reduced lung function in cotton workers. In the chronic lung disease example the null hypothesis would be that men allocated to exercises showed no more benefit than the men allocated as controls. We could conclude that the exercises were beneficial only if the data were inconsistent with the null hypothesis.

5.1 Statistical significance and confidence intervals

The use of statistics in biomedical journals over recent decades has increased exponentially. Associated with this increase has been an unfortunate trend away from examining basic results towards an undue concentration on ‘hypothesis testing’. In this approach, data are examined in relation to a statistical ‘null’ hypothesis and the practice has led to a mistaken belief that studies should aim at attaining ‘statistical significance’. Contrary to this paradigm is that most research questions in medicine are aimed at determining the magnitude of some factor(s) of interest on an outcome.

The common statements ‘$P < 0.05$’ and ‘$P = NS$’ convey little information about a study’s findings and rely on an arbitrary convention of using the 5% level of statistical significance to define two alternative outcomes: significant (‘it worked’) or not significant (‘it didn’t work’). Furthermore, even precise P values convey nothing about the sizes of the differences between study groups. In addition, there is a tendency to equate statistical significance with medical importance or biological relevance, however small differences of no real interest can be statistically significant with large sample sizes, whereas clinically important effects may be statistically non-significant only because the number of subjects studied was small.

It is therefore good practice when reporting the results of an analysis involving significance tests to give estimates of the sizes of the effects, both point estimates and confidence intervals. Then readers can make their own interpretation, depending on what they consider to be an important difference (which is not a statistical question).

The five possibilities (as shown in Figure 17) are:

1. The difference is significant and certainly large enough to be of practical importance — ‘definitely important’.
2. The difference is significant but it is unclear whether it is large enough to be important — ‘possibly important’.
3. The difference is significant but too small to be of practical importance — ‘not important’.
4. The difference is not significant but may be large enough to be important — ‘not conclusive’.
5. The difference is not significant and also not large enough to be of practical importance — ‘true negative’.

5.2 Steps involved in testing significance

The full answer to any exercise involving a significance test should include:

1. A statement of the null hypothesis.

2. Calculation of test statistic and its associated P value.

3. A statement of conclusion, which should include: (a) the significance or otherwise of the effect being tested, (b) supporting statistics (the test statistic, degrees of freedom, and P value), and (c) an estimate of effect (the point estimate and its confidence interval).
We wish to compare conception rates among cows where oestrus has been induced using a CIDR device and cows where oestrus has occurred naturally. You have collected the following data:

There were 53 services applied to CIDR-induced oestrus events. Of these 53 services, 23 resulted in conception. There were 124 services applied to natural oestrus events. Of these 124 services, 71 resulted in conception. A chi-squared test will be used to compare the two proportions (that is, to test the hypothesis that the proportions 23/53 and 71/124 do not differ). The null hypothesis is that conception rates for CIDR-induced oestrus events are equal to conception rates for natural oestrus events.

The chi-squared test statistic, calculated from these data is 2.86 (obtained from statistical tables). The number of degrees of freedom is 1. The P value corresponding to this test statistic and degrees of freedom is 0.09.

Since our observed P value is greater than 0.05 we accept the null hypothesis that conception rates for CIDR-induced oestrus events are equal to conception rates for natural oestrus events (chi-squared test statistic = 2.86, df = 1, P = 0.09).

The conception rate for CIDR-induced oestrus events was 43% (95% CI 31% to 57%). The conception rate for natural oestrus events was 57% (95% CI 48% to 66%).
6 Diagnostic tests

By the end of this unit you should be able to:

• Explain what is meant by the terms sensitivity and specificity, as applied to diagnostic tests.
• Given testing results presented in a $2 \times 2$ table, calculate and interpret test sensitivity and specificity.
• Given testing results presented in a $2 \times 2$ table, calculate and interpret test positive and negative predictive value.

A test may be defined as any process or device designed to detect (or quantify) a sign, substance, tissue change, or body response in an animal. Tests included:

• Routine examination of an animal or premises.
• Questions posed during history taking.
• Clinical signs.
• Laboratory findings - haematology, serology, biochemistry, histopathology.
• Post mortem findings.

If tests are to be used in a decision-making context, the selection of an appropriate test should be based on its ability to alter your assessment of the probability that a disease does or does not exist.

6.1 Screening versus diagnosis

In clinical practice, tests tend to be used in two ways:

**Screening** tests are those applied to apparently healthy members of a population to detect seroprevalence of certain diseases, the presence or disease agents, or subclinical disease. Usually, those animals that return a positive to such tests are subject to further in-depth diagnostic work-up, but in other cases (such as national disease control programs) the initial test result is taken as the state of nature.

**Diagnostic** tests are used to confirm or classify disease status, provide a guide to selection of treatment, or provide an aid to prognosis. In this setting, all animals are ‘abnormal’ and the challenge is to identify the specific disease the animal in question has.
6.2 Sensitivity and specificity

Analytic sensitivity of an assay for detecting a given chemical compound refers to the lowest concentration the test can detect. Analytic specificity refers to the capacity of the test to react to only one chemical compound.

Epidemiologic sensitivity and specificity depend on analytic sensitivity and specificity, but are entirely different concepts. Epidemiologic sensitivity answers the question: ‘Of all individuals that actually had disease X, what proportion tested positive? Epidemiologic specificity answers the question: ‘Of all individuals that were free of disease X, what proportion tested negative? Figure 18 presents this concept diagramatically.

Figure 18: Test results measured on a continuous scale, showing the distribution of results that might be obtained for healthy and diseased individuals. The cut-off value for the test is shown by the vertical dashed line: those individuals with a result less than the cut-off value are diagnosed as non-diseased, those individuals with a result greater than the cut-off value are diagnosed as diseased. Using this diagnostic test, disease-positive individuals with a test result in the area marked ‘A’ will be false negatives. Disease-negative individuals with a test result in the area marked ‘B’ will be false positives.

6.3 Accuracy and precision

The accuracy of a test relates to its ability to give a true measure of the substance being measured. To be accurate, a test need not always be close to the true value, but if repeat tests are run, the average of the results should be close to the true value. An accurate test will not over- or under-estimate the true value. Results from tests can be ‘corrected’ if the degree of inaccuracy can be measured and the test results adjusted accordingly.

The precision of a test relates to how consistent the results of the test are. If a test always gives the same value for a sample (regardless of whether or not it is the correct value), it is said to be precise.
6.3.1 Accuracy

Assessment of test accuracy involves running the test on samples with a known quantity of substance present. These can be field samples for which the quantity of substance present has been determined by another, accepted reference procedure. Alternatively, the accuracy of a test can be determined by testing samples to which a known quantity of a substance has been added. The presence of background levels of substance in the original sample and the representativeness of these ‘spiked’ samples make this approach less desirable for evaluating tests designed for routine field use.

6.3.2 Precision

Variability among test results might be due to variability among results obtained from running the same sample within the same laboratory (repeatability) or variability between laboratories (reproducibility). Regardless of what is being measured, evaluation of test precision involves testing the same sample multiple times within and/or among laboratories.

6.4 Test evaluation

The two key requirements of a diagnostic test are: (1) the test will detect diseased animals correctly, and (2) the test will detect non-diseased animals correctly. To work out how well a diagnostic test performs, we need to compare it with a ‘gold standard.’ A gold standard is a test or procedure that is absolutely accurate. It diagnoses all diseased animals that are tested and misdiagnoses none. Once samples are tested using the gold standard and the test to be evaluated, a $2 \times 2$ table can be constructed, allowing test performance to be quantified.

<table>
<thead>
<tr>
<th></th>
<th>Diseased</th>
<th>Non-diseased</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Test positive</td>
<td>$a$</td>
<td>$b$</td>
<td>$a + b$</td>
</tr>
<tr>
<td>Test negative</td>
<td>$c$</td>
<td>$d$</td>
<td>$c + d$</td>
</tr>
<tr>
<td>Total</td>
<td>$a + c$</td>
<td>$b + d$</td>
<td>$a + b + c + d$</td>
</tr>
</tbody>
</table>

6.4.1 Sensitivity

The sensitivity of a test is defined as the proportion of subjects with disease that test positive [$p(T^+|D^+)$]. A sensitive test will rarely misclassify animals with the disease. Sensitivity is a measure of accuracy for predicting events.

$$\text{Sensitivity} = \frac{a}{(a + c)} \quad (6.1)$$

Sensitivity is:
• The conditional probability of a positive test, given the presence of disease.
• The likelihood of a positive test in a diseased animal.
• The proportion of animals with disease that have a positive test for the disease.
• The true positive rate (relative to all animals with disease).

6.4.2 Specificity

The specificity of a test is defined as the proportion of subjects without the disease that test negative \( p(T^-|D^-) \). A highly specific test will rarely misclassify animals without the disease.

\[
\text{Specificity} = \frac{d}{(b + d)} \tag{6.2}
\]

Specificity is:

• The conditional probability of a negative test, given the absence of disease.
• The likelihood of a negative test in an animal without disease.
• The proportion of animals without the disease that have a negative test for the disease.
• The true negative rate (relative to all animals without disease).

Sensitivity and specificity are inversely related and in the case of test results measured on a continuous scale they can be varied by changing the cut-off value. In doing so, an increase in sensitivity will often result in a decrease in specificity, and vice versa. The optimum cut-off level depends on the diagnostic strategy. If the primary objective is to find diseased animals (that is, to minimise the number of false negatives and accept a limited number of false positives) a test with a high sensitivity and good specificity is required. If the objective is to make sure that every test positive is ‘truly’ diseased (minimise the number of false positives and accept a limited number of false negatives) the diagnostic test should have a high specificity and good sensitivity.

6.4.3 Positive predictive value

The positive predictive value is the proportion of subjects with positive test results which have the disease.

\[
\text{Positive predictive value} = \frac{a}{(a + b)} \tag{6.3}
\]

Positive predictive value is:
• The predictive value of a positive test.
• The post test probability of disease following a positive test.
• The posterior probability of disease following a positive test.

6.4.4 Negative predictive value

The negative predictive value is the proportion of subjects with negative test results which do not have the disease.

\[
\text{Negative predictive values} = \frac{d}{(c + d)} \quad (6.4)
\]

Negative predictive value is:

• The predictive value of a negative test.
• The posttest probability of no disease following a negative test.
• The posterior probability of no disease following a negative test.

Predictive values quantify the probability that a test result for a particular animal correctly identifies the condition of interest. Estimation of predictive values requires knowledge of sensitivity, specificity and the prevalence of the disease in the population. It is important to remember that predictive values are used for interpretation at the individual animal level and cannot be used to compare tests. The effect of prevalence on predictive values is considerable. Given a prevalence of disease in a population of around 30% and we are using a test with 95% sensitivity and 90% specificity, the predictive value of a positive test would be 80% and the predictive value of a negative test would be 98%. If the prevalence of disease is only 3% and the test characteristics remain the same, the predictive value of a positive and negative test will be 23% and 99.8%, respectively.

Remember the following general rules about diagnostic tests:

• Sensitivity and specificity are generally independent of prevalence.
• If the prevalence increases, positive predictive value increases and negative predictive value decreases.
• If the prevalence decreases, positive predictive value decreases and negative predictive value increases.
• The more sensitive a test, the better its negative predictive value.
• The more specific a test, the better its positive predictive value.
6.5 Prevalence estimation

The estimate of disease prevalence determined on the basis of an imperfect test is called the apparent prevalence. Apparent prevalence is the proportion of all animals that give a positive test result. It can be more than, less than, or equal to the true prevalence. If sensitivity and specificity of a test are known, then the true prevalence can be calculated using the following formula:

\[
p(D^+) = \frac{AP - (1 - Sp)}{1 - [(1 - Sp) + (1 - Se)]} = \frac{AP + Sp - 1}{Se + Sp - 1}
\]

Where:

\( AP \): apparent prevalence
\( Se \): sensitivity (0 - 1)
\( Sp \): specificity (0 - 1)

Individual cow somatic cell counts (ICSCC) are used as a screening test for subclinical mastitis in dairy cattle. This test has a sensitivity of 0.80 and a specificity of 0.80. The apparent prevalence of mastitis in this herd using the screening test is 23 cases per 100 cows. True prevalence \( p(D^+) \) may be calculated as follows:

\[
AP = 0.23 \\
Se = 0.80 \\
Sp = 0.80 \\
p(D^+) = \frac{0.23 + 0.80 - 1}{0.80 + 0.80 - 1} \\
p(D^+) = 0.03 / 0.6 \\
p(D^+) = 0.05
\]

The true prevalence of mastitis in this herd is 5 cases per 100 cows.

**Figure 19:** Relationship between prevalence and positive predictive value for tests of different sensitivities and specificities.
6.6 Diagnostic strategies

Clinicians commonly perform multiple tests to increase their confidence that a patient has a particular diagnosis. When multiple tests are performed and all are positive, the interpretation is straightforward: the probability of disease being present is relatively high. It is far more likely however, that some of the tests return a positive result and others will be negative. Interpretation, in this case, is more complicated.

Multiple test results can be interpreted in parallel or series.

6.6.1 Parallel interpretation

Parallel interpretation means that when multiple tests are run an individual is declared positive if at least one of the multiple tests returns a positive result. Interpreting test results in parallel increases the sensitivity and therefore the negative predictive value for a given disease prevalence. However, specificity and positive predictive value are lowered. As a consequence, if a large number of tests are performed and interpreted in this way then virtually every individual will be considered positive.

6.6.2 Serial interpretation

Series interpretation means that when multiple tests are run an individual is declared positive if all tests return a positive result. Series interpretation maximises specificity and positive predictive value which means that more confidence can be attributed to positive results. It reduces sensitivity and negative predictive value, and therefore it becomes more likely that diseased animals are being missed.

6.7 Screening and confirmatory testing

With a screening and confirmatory test strategy (as often used in disease control schemes) a test is applied to every animal in the population to screen the population for positives. Ideally, this test should be easy to apply and low in cost. It also should be a highly sensitive test so that it misses only a small number of diseased or infected animals. Its specificity should still be reasonable, so that the number of false positives subjected to the confirmatory test remains economically justifiable.

Individuals that return a negative result to the screening test are considered definitive negatives and not submitted to any further examination. Any animal positive to the screening test is subjected to a confirmatory test. The confirmatory test can require more technical expertise and more sophisticated equipment, and be more expensive, because it is only applied to a reduced number of samples. But it has to be highly specific, and any positive reaction to the confirmatory test is considered a definitive positive.
The same principles apply to disease control and eradication schemes. We firstly apply a test to detect disease: individuals identified as positive are removed from the population. To efficiently identify positives we need a highly sensitive test. During this early phase of a program the apparent prevalence will be higher than the true prevalence, as a consequence of test specificity being less than 100%. As the program continues, test positive animals are identified and culled. The population prevalence of disease declines. As prevalence declines, the positive predictive value of testing declines which increases the gap between apparent and true prevalence. The proportion of false positives will then increase. At this stage a highly specific test is required. In some cases it may become necessary to use a number of tests interpreted in series to increase specificity.

Rules of thumb:

- If the objective is to find disease (e.g. diagnose neoplasia early in a much-loved pet) use a highly sensitive test.
- If the objective is to confirm the absence of disease (e.g. testing a cow for brucellosis before it is imported into New Zealand) use a highly specific test.

6.8 Multiple testing

Multiple testing is common in clinical practice. Blood samples from patients are sent to a laboratory and for a fixed fee a range of haematological and biochemical analyses are performed. The objective is to identify normal and abnormal blood parameters. The technique is useful for establishing patterns which are suggestive of a particular disease. The approach becomes questionable if it is used as part of a ‘fishing expedition’ for a diagnosis. We need to keep in mind that a cut-off for a single test is typically set such that it includes 95% of the normal population, which means the test will produce 5% false positives. As an example, with 12 diagnostic tests measuring different blood parameters, each of them will have a 0.95 probability of diagnosing a normal animal correctly as negative. But it also means that the overall chance of a correct negative diagnosis on all tests is \(0.95^{12} = 54\%\). There is, as a result, a 46% chance that a normal animal has at least one abnormal (false positive) value among the 12 tests.

6.9 Likelihood ratios

Diagnostic testing is often undertaken to help us decide whether or not an individual is diseased. Because diagnostic tests are imperfect (that is, false positives and false negatives occur) we should move away from the ‘test positive = disease positive’, ‘test negative = disease negative’ paradigm and think about testing as a process that provides us with a probability estimate of the presence of disease in the tested individual. Likelihood ratios offer a means for doing this.
The likelihood ratio for a positive test tells us how likely we are to find a positive test result in a diseased individual compared with a non-disease individual. The likelihood ratio for a positive test is estimated on the basis of dividing the probability of a particular test result in the presence of disease (sensitivity) by the probability of the test result in the absence of disease (1 - specificity). The likelihood ratio for a negative test equals (1 - sensitivity) divided by the specificity. Thus:

\[
LR^+ = \frac{Se}{1 - Sp} \quad (6.6)
\]

\[
LR^- = \frac{1 - Se}{Sp} \quad (6.7)
\]

Where:

\(Se\): sensitivity (0 - 1)

\(Sp\): specificity (0 - 1)

Likelihood ratios (LR) can be calculated using single cut-off values, so that one obtains only one pair of likelihood ratios, one for a positive (LR+) and another for a negative test result (LR-). More powerful information can be extracted from the diagnostic test by using multilevel likelihood ratios. In this case ranges of test results will have associated likelihood ratio values.

Likelihood ratios provide a quantitative measure of the diagnostic information contained in a particular test result. If we consider the expectation of the likelihood that an animal has a certain condition (= pre-test odds of disease) the likelihood ratio of the test multiplied by the pre-test odds gives us a revised estimate of the odds of disease (= post-test odds). This result can be re-expressed as a probability to make it more interpretable. To convert odds to probability and vice versa, use the following equations:

\[
\text{Odds of event} = \frac{\text{Probability of event}}{1 - \text{Probability of event}} \quad (6.8)
\]

\[
\text{Probability of event} = \frac{\text{Odds of event}}{1 + \text{Odds of event}} \quad (6.9)
\]

Individual cow somatic cell counts (ICSCC) are used as a screening test for sub-clinical mastitis in dairy herds. A client has a herd of dairy cows where the prevalence of subclinical mastitis is estimated to be around 5%. You receive the following data from herd testing:

<table>
<thead>
<tr>
<th></th>
<th>Mastitis +</th>
<th>Mastitis -</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>ICSCC &gt; 200</td>
<td>40</td>
<td>190</td>
<td>230</td>
</tr>
<tr>
<td>ICSCC &lt; 200</td>
<td>10</td>
<td>760</td>
<td>770</td>
</tr>
<tr>
<td>Total</td>
<td>50</td>
<td>950</td>
<td>1000</td>
</tr>
</tbody>
</table>
At a later date you examine an individual cow from this herd and note that she has an ICSCC of 320,000 cells/mL. What is the probability that this cow has mastitis?

Using a fixed ICSCC threshold 200,000 cells/mL to classify individuals as mastitic or not, and assuming that ICSCC testing has a sensitivity of 80% and a specificity of 80%, the calculated positive predictive value is $40 \div 230 = 17\%$. On the basis of these calculations we reckon that if a cow has an ICSCC value greater than 200,000 cell/mL the probability that she really has mastitis is around 17%.

Your herd testing authority provides you with the following likelihood ratios for categories of ICSCC values:

<table>
<thead>
<tr>
<th>ICSCC</th>
<th>&lt; 100</th>
<th>100 – 200</th>
<th>200 – 300</th>
<th>300 – 400</th>
<th>&gt; 400</th>
</tr>
</thead>
<tbody>
<tr>
<td>LR (+)</td>
<td>0.14</td>
<td>0.37</td>
<td>2.50</td>
<td>14.50</td>
<td>40.80</td>
</tr>
</tbody>
</table>

The posterior probability of mastitis is determined as follows:

1. The pre-test probability of mastitis is $50 \div 1000 = 0.05$.
2. The pre-test odds of mastitis is $0.05 \div (1 - 0.05) = 0.053$.
3. The post-test odds of mastitis given a positive test result is pre-test odds $\times$ LR(+) $= 0.053 \times 14.5 = 0.76$.
4. The post-test probability of mastitis given a positive test result: $0.76 / (1 + 0.76) = 0.43$.

The post-test probability of a cow with a ICSCC of 320,000 cells/mL being mastitic is around 43%.

Post-test probabilities can be quickly determined in practice by using a nomogram, as shown in Figure [20]. On the left hand side of the nomogram (labeled ‘Prior prob’ in Figure [20]) we mark the pre-test probability that the individual being examined has disease. We next identify the point defining the likelihood ratio of a positive test result along the middle scale. Finally, we draw a straight line from the pre-test probability estimate through the likelihood ratio value to the corresponding post-test probability value on the right-hand side of the chart.

A nice feature of this approach to evaluating test information is that sequential testing can be easily handled. If we are using serial interpretation, the post-test probability of disease from the first test becomes the pre-test probability for the second test.

To continue the mastitis example described above lets imagine that we examine our cow and as part of that examination we test milk from each quarter using a rapid mastitis test (RMT). We are told that the sensitivity and specificity of the RMT is 70% and
80%, respectively. Our cow returns a positive result to the RMT. What now is this cow’s probability of being mastitic?

The likelihood ratio of a positive RMT is 3.5 (= 0.70 / 1 - 0.80). If the pre-test probability of disease is 0.43 we can use a nomogram to estimate the posterior probability of disease, given a positive test, as 0.72. We are now much more certain that this cow has mastitis.

The advantage of the likelihood ratio method of test interpretation is that we can better appreciate the value (i.e. the increase in post-test probability) provided by each diagnostic test that is applied (in the above example, ICSCC provided more information compared with the RMT). If the cost of each test applied is known the cost per unit increase in post-test probability can be determined, enabling us to be more objective in our use of diagnostic resources.
Figure 21: Diagram showing how the estimated probability of disease changes after applying a series of diagnostic tests. In our example of the cow with mastitis, we had a prior belief that the probability of the cow being mastitic was 5%. After considering the ICSCC result this probability increased to 43%. After applying a rapid mastitis test and getting a positive result, the probability of the cow having mastitis increased to 72%.
7 Sampling populations

By the end of this unit you should be able to:

- Explain the key features of simple random sampling, systematic random sampling, stratified random sampling, and cluster sampling.
- Describe the advantages of disadvantages of simple random sampling, systematic random sampling, stratified random sampling, and cluster sampling.
- Describe ways to reduce error when making inferences from sampled data.

Epidemiologists frequently examine populations to:

- Detect the presence of a disease;
- Demonstrate that a disease is not present within a population; and
- Establish the level of occurrence of a disease within a population.

To produce accurate estimates of disease we must be able to measure populations effectively. The exact level of disease within a population will be obtained if every individual within the population is examined (and if there was no measurement error). This technique is a census. However, in many situations a census is impossible and/or excessively expensive. Usually an accurate estimate can be obtained by examining some of the animals (a sample) from the population.

7.1 Probability sampling methods

A probability sample is one in which every element in the population has a known non-zero probability of being included in the sample.

7.1.1 Simple random sampling

Simple random sampling occurs when each subject in the population has an equal chance of being chosen.

7.1.2 Systematic random sampling

With systematic random sampling, the selection of sampling units occurs at a predefined equal interval (known as the sampling interval). This process is frequently used when the total number of sampling units is unknown at the time of sampling (e.g. in a study where patients that enter an emergency department of a hospital on a given day are to
Figure 22: Simple random sampling. If a sample of five cows was required, five random numbers between 1 and 10 would be generated and cows selected on the basis of the generated random numbers.

be sampled — at the start of the study day we do not know the total number of patients seen by the end of the day).

Suppose we are studying inpatient medical records on an ongoing basis for a detailed audit. The total number of records in the population is not likely to be known in advance of the sampling since the records are to be sampled on an ongoing basis (and so it would not be possible to use simple random sampling). However, it would be possible to guess the approximate number of records that would be available per time period and to select a sample of one in every \( k \) records as they become available.

We require a total of 300 records over a 12-month period to complete the study. If there are, on average, ten new discharge records available per day then total number of records available per year is estimated to be \( 10 \times 365 = 3650 \). To obtain the required number of records per year in the sample, the sampling interval \( k \) should be the largest integer in the quotient \( 3650 \div 300 \). Since the value of the quotient is 12.17, the sampling interval \( k \) would be 12. Thus, we would take a sample of 1 from every 12 records.

One way to implement this procedure is to identify each record as it is created with a consecutive number. At the beginning of the study a random number between 1 and 12 is chosen as the starting point. Then, that record and every twelfth record beyond it is sampled. If the random number chosen is 4, then the records in the sample would be 4, 16, 28, 40, 52, and so on.

7.1.3 Stratified random sampling

Stratified sampling occurs when the sampling frame is divided into groups (strata) and a random selection within each stratum are selected. Stratified sampling is frequently undertaken to ensure that there is adequate representation of all groups in the population in the final sample. The simplest form is proportional stratified random sampling, where the number sampled within each stratum is proportional to the total number within the stratum.

Suppose that you wish to determine the prevalence of disease in the pig population of a region. Previous surveys have indicated that 70% of the regions pigs are located in very large, intensive specialised pig farms, 20% of pigs are found within smaller farming units (frequently as a secondary enterprise on large dairy farms), and 10% of pigs are kept singly within small plots around towns (by people whose major occupation is not farming). With proportional stratification, a sample would be selected at random from within each stratum such that the aggregated sample would consist of 70% pigs obtained from the large intensive farms, 20% pigs obtained from the smaller pig farms, and 10% pigs obtained from small plots near towns.

In some situations obtaining a sample from a particular stratum is more difficult or costly than for other strata. In the example described it may be more costly to sample from
the pigs held in small plots around towns. This may be due to an incomplete register of smallholdings, difficulties in contacting pig owners and arranging suitable times to visit and perhaps extra travel requirements. In this situation, a technique known as non-proportional sampling may adopted.

An advantage of stratified sampling is that the precision of parameter estimates is improved. If the population can be divided into logical strata whereby the variation within each stratum is small compared with the variation between strata a more precise estimate will be obtained.

We wish to determine average total lactation milk volume (total litres) produced by dairy cows in a region. The region contains two breeds of cattle. One breed (Friesian) is characterised by production of large volumes of milk with low concentrations of milk solids. The other breed (Jersey) is characterised by production of small volumes of milk with high concentrations of milk solids. By dividing the population into breed strata and sampling within each stratum, the average lactation milk volume production of each breed can be estimated with accuracy. The mean milk production for cows within the region can also be estimated by calculation of a weighted mean based upon each stratum mean and the stratum size.

**Figure 23:** Stratified random sampling. A group of animals are stratified by breed and a random sample within each breed taken.

### 7.1.4 Cluster sampling

Cluster sampling occurs when the sampling frame is divided into logical aggregations (clusters) and a random selection of clusters is performed. The individual sampling units (known as primary sampling units) within the selected clusters are then examined. Clustering may occur in space or time. For example, a litter of piglets is a cluster formed within a sow, a herd of dairy cows is a cluster within a farm, and a fleet of fishing boats is a cluster formed within space (that is, a port or harbour).
Although cluster sampling has a number of advantages (including the advantage of being economical) it has the disadvantage that the standard errors of estimates are often high compared with those obtained from samples of the same number of listing units chosen by other sampling designs. The reason for this is that listing units within the same cluster are tend to be more homogenous than those listing units from different clusters. There are two types of cluster sampling:

- One stage cluster sampling occurs when clusters are selected by simple random sampling and then, once selected, all of the listing units within the cluster are examined.

- Two stage cluster sampling occurs when clusters are selected by simple random sampling and then, once selected, a random sample of listing units within each cluster are selected for examination. Estimation of population characteristics is straightforward in this situation when each cluster has the same number of listing units. Estimation of population characteristics is not straightforward when each cluster contains different numbers of listing units (in this case, you will need to consult a statistician).

The number of clusters to sample and the number of listing units within each cluster to sample will depend upon the relative variation of the factor of interest between clusters, compared with within clusters, and the relative cost of sampling clusters compared with the cost of sampling individual listing units.

- When the between-cluster variation is large relative to the within-cluster variation, you will have to sample many more clusters to get a precise estimate.

- When the between-cluster variation is small relative to the within-cluster variation, you will have to sample many more individual listing units within each cluster to get a precise estimate.

### 7.2 Non-probability sampling methods

Non-probability sampling occurs when the probability of selection of an individual within a population is not known and some groups within the population are more or less likely than other groups to be selected. Non-probability sampling methods include:

- Convenience sampling: where the most accessible or amenable sampling units are selected;

- Purposive sampling: where the most desired sampling units are selected; and
• Haphazard sampling: where sampling units are selected using no particular scheme or method. Inherent in this type of sampling is the problem that subconscious forces may influence the person selecting the units in an attempt to 'balance' the sample. For example, a young animal may be preferred for the next selection immediately after an older animal has been selected.

Non-probability sampling will produce biased population estimates, and the extent of that bias cannot be quantified.

### 7.3 Sources of error and how to reduce error

When you derive an estimate from a sample you want it to be precise and accurate. A precise estimate has confidence intervals that are small. An accurate has confidence intervals that are centred on the true population value. There are two types of error that can exist within a sample estimate: random errors and bias. The difference between random error and bias may be explained using the following diagram:

![Figure 24: The distribution of bullets fired at the target on the left show little evidence of random error and bias. The distribution of the bullets fired at the centre target show a high degree of random error and a low degree of bias. The distribution of the bullets fired at the target on the right show a low degree of random error and a high degree of bias.](image)

#### 7.3.1 Random error

Random error is caused by chance. A random selection of individuals taken to make up a sample will differ slightly from each other. These differences will result in sample estimates that differ slightly from each other and also from the target population. Random error is the inherent error that arises from using a sample to make a measurement of a population. The influence of random error may be reduced by:

1. Increasing the size of the sample taken. Using the central limit theorem it can be demonstrated that a fourfold increase in sample size will result in a halving of the confidence interval.

2. Modifying the sample selection procedure to ensure that only the target group is sampled. For example, you may be interested in the performance of only one
particular breed of dairy cow. You can design the study to ensure that you sample animals only from farms that contain this breed of cow. Stratified sampling is a technique that reduces sample variance by dividing the population into individual strata. Each stratum contains individuals that are similar, and so the variance within strata is less than the variation between strata. You would typically obtain samples from individual strata that have less variation than similar-sized samples obtained from the whole (unstratified) population.

3. Using an appropriate scale of measurement. Ratio estimators may result in a reduction in confidence intervals in some situations. Suppose, for example, that you wish to determine whether farmed lambs have reached the correct weight for sale. You could take a sample of lambs and estimate the average weight of the sample and from that an associated confidence interval. If the weight of lambs in the population is quite variable and you do not select a large sample it is likely that the associated confidence interval will be wide (and will include the target value). An alternative is to dichotomously classify each lambs weight within the sample with respect to the target weight (i.e. describe it as either above or below target weight). You can then calculate an estimate of the proportion of lambs that have obtained target weight (along with associated confidence intervals). You are more likely to produce narrow confidence intervals for this ratio estimate and are thus able to make a more confident decision regarding the sale of the lambs.

7.3.2 Bias

Bias is caused by systematic error, a systematic error being one that is inherent to the technique being used that results in a predictable and repeatable error for each observation. Bias may present itself in two ways:

1. Non-observational errors are due to inappropriate sample selection. These errors may arise from failure to include an important group of individuals within the sampling frame (resulting in their exclusion from selection), or as a result of missing data. In some situations data may be missing from a particular group of individuals within the sample.

2. Observational errors are due to inappropriate measurements. These may be attributable to false responses (i.e. participants make untrue statements) or to measurement errors.

7.4 Sampling techniques

Random sampling means that each unit of interest within the population has the same probability of selection into the sample as every other unit. The probability of selection
of individual units must not differ. This is irrespective of accessibility, ease of collection or other differences that may exist between individuals. There are several important considerations to take into account before collecting a random sample:

- The target population must be identified and defined.
- A study population that is representative of the target population must be identified. The study population must not differ in composition from the target population.
- A sampling frame is produced. The sampling frame is a means of identifying every unit of interest (sampling unit) within the study population.
- Sampling units are selected from the sampling frame using a random (probabilistic) approach such that each sampling unit within the sampling frame has an equal probability of selection.

7.4.1 Methods of randomisation

There are two principal techniques for random sampling, physical randomisation and the use of random numbers. Physical randomisation is a process where sampling units are selected using physical systems that contain random elements. These include the selection of numbered marbles from a bag, the use of a die, or the toss of a coin.

Random numbers are a sequence of numbers comprising individual digits with an equal chance that any number from 0 to 9 will be present. Tables of random numbers can be used for sample selection. Some computer programs can generate random numbers. These programs use algorithms to produce the sequence of numbers. The sequence of numbers that is generated depends upon the value chosen as the starting value for the algorithm (the seed value). Whilst there is an equal probability that any digit from 0 to 9 will be present in a position chosen at random from the sequence, the actual digit present at each point of the sequence is determined by the seed value. In other words, the exact sequence of random numbers can be reproduced if the process is repeated using the same seed value. Computer-generated random numbers are frequently called pseudo-random numbers for this reason.

7.4.2 Replacement

Samples may be taken in one of two ways: sampling with replacement or sampling without replacement. In sampling with replacement, each selected unit is examined and recorded and then returned to the sampling frame. These units may then be selected into the sample again.
In sampling without replacement, each selected unit is examined and recorded and then withdrawn from the sampling frame. These units are excluded from selection into the sample again. Intuitively, sampling without replacement is the most logical; it is better to have different information from new animals as opposed to having copies of information obtained from the repeated sampling of a single animal. However, there are statistical reasons why sampling with replacement may be employed in certain circumstances. These reasons relate to the mathematics of the estimation process. In sampling with replacement the probability of selection of a unit remains the same from the first selection through to the last selection. The distribution of results within the final sample is described by the binomial distribution. In sampling without replacement, the probability of selection of the next unit changes each time a selection is made. This is due to a reduction in size of the denominator as each unit is drawn. The distribution of results is described by the (more complex) hypergeometric distribution.

The difference between the two sampling procedures is not important when samples are drawn from large populations. Often, the binomial distribution is used to approximate the hypergeometric distribution when analysing the results of samples drawn without replacement from large populations.

7.5 Sample size

The choice of sample size involves both statistical and non-statistical considerations. Non-statistical considerations include the availability of time, money, and resources. Statistical considerations include the required precision of the estimate, and the variance expected in the data. In descriptive studies we need to specify the desired level of confidence that the estimate obtained from sampling is close to the true population value \((1 - \alpha)\). In analytical studies we may also be interested in the power \((1 - \beta)\) of the study to detect real effects.

7.5.1 Simple and systematic random sampling

The following formulae may be used to derive sample sizes appropriate to estimate population parameters (population total, mean, and proportion) on the basis of a simple random sample. From: Levy PS, and Lemeshow S (1999). Sampling of Populations Methods and Applications. London: Wiley Series in Probability and Statistics; p 74.

\[
\text{Total: } n \geq \frac{4V^2_x}{\epsilon^2} \tag{7.1}
\]

\[
\text{Mean: } n \geq \frac{4V^2_x}{\epsilon^2} \tag{7.2}
\]

\[
\text{Proportion: } n \geq \frac{4(1 - P_y)P_y}{\epsilon^2} \tag{7.3}
\]
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Where:

- $z$: the reliability coefficient (e.g. $z = 1.96$ for an alpha level of 0.05)
- $V^2$: the relative variance (the estimated population variance divided by the square of the estimated population mean)
- $\epsilon$: the maximum relative difference between our estimate and the unknown population value
- $P_0$: the unknown population proportion

Suppose that a survey of retail pharmacies is to be conducted in a state with 2500 pharmacies. The purpose of the survey is to estimate the average retail price of 20 tablets of a commonly used vasodilator. An estimate is needed that is within 10% of the true value of the average retail price in the state. As a rough guess we reckon that 95% of the values will lie between $4.20$ and $9.80$. How many pharmacies should be included in the survey to be 95% confident that the surveyed value will be within 10% of the average retail price in the state?

The range of the estimated 95% confidence interval is $9.80 - 4.20 = 5.40$. We assume that the range of the 95% confidence interval is equal to 4 times the population standard deviation. The estimated population standard deviation is $5.40 \div 4 = 1.40$.

The estimated population variance is $1.40 \times 1.40 = 1.96$.

The estimated population mean is $7.00$.

$V^2 = \frac{1.96}{7 \times 7} = 0.04$.

Sample size $= \left(4 \times 0.04\right) \div \left(0.1 \times 0.1\right) = 16$.

A sample of 16 pharmacies are required to meet the requirements of the survey.

7.5.2 Sampling to detect disease

Veterinarians are frequently asked to test groups of animals to confirm the absence of disease. The number of animals that should be tested to provide a specified level of confidence that disease is detected is given by:

$$n = (1 - \alpha^{\frac{1}{2}}) \times \left(N - \frac{D - 1}{2}\right)$$  \hspace{1cm} (7.4)

Where:

- $N$: the population size
- $\alpha$: 1 - confidence level (usually $\alpha = 0.05$)
- $D$: the estimated minimum number of diseased animals in the group (population size $\times$ the minimum expected prevalence)

What is the approximate number of individuals that should be tested in a herd of 200 to confirm the presence of disease if the expected prevalence is 20%?

$N = 200$

$a = 0.05$

$D = 0.20 \times 200 = 40$

$n = \left(1 - 0.05^{\frac{1}{2}}\right) \times \left(200 - \frac{40 - 1}{2}\right)$

$n = 0.072 \times 180.5$

$n = 13$

A minimum of 13 individuals need to be tested.
7.5.3 Sampling to prove that disease is not present

The probability of failing to detect disease (when it actually exists) is given by:

\[ p = \left(1 - \frac{D}{N}\right)^n \]  

(7.5)

Where:

- \( N \): the population size
- \( d \): the number of diseased animals present
- \( n \): the number of animals tested

<table>
<thead>
<tr>
<th>We estimate the prevalence of brucellosis in a herd of 200 to be around 5%. What is the probability of failing to detect brucellosis if we test 28 animals?</th>
</tr>
</thead>
<tbody>
<tr>
<td>( d = 0.05 \times 200 = 10 )</td>
</tr>
<tr>
<td>( N = 200 )</td>
</tr>
<tr>
<td>( n = 28 )</td>
</tr>
<tr>
<td>( p = (1 - \frac{10}{200})^{28} = 0.23 )</td>
</tr>
</tbody>
</table>

There is a 23% chance that we will fail to detect disease if we sample 28 cattle from a herd of 200.
8 Outbreak investigation

An outbreak is a series of disease events clustered in time. During an outbreak the investigator asks the questions:

- What is the problem?
- Can something be done to control it?
- Can future occurrences be prevented?

These notes outline an approach to investigating outbreaks of disease in animal populations. Although the term outbreak implies a sudden (and possibly spectacular) event (e.g. an outbreak of botulism in feedlot cattle), be aware that outbreaks can be of a more insidious nature: some causing subclinical losses in a population of animals over an extended period before being identified, characterised and investigated.

8.1 Verify the outbreak

8.1.1 What is the illness?

Once a suspected outbreak is identified, identifying the specific nature of the illness is an important early step. An attempt should be made to characterise cases (leading towards a formal case definition, see below). Usually it will not be possible to make a definitive diagnosis at this stage. What is required is a working definition of the disease or syndrome: for example ‘ill thrift in recently weaned calves’ or ‘sudden death in grower pigs.’

8.1.2 Is there a true excess of disease?

The first issue to be certain of is whether or not the outbreak is genuinely an unusual event worthy of special attention. The number of cases per unit time should be substantially greater than what is normal for the group of individuals under investigation. It is common to have owners and others concerned about a possible outbreak which is transient increase in the normal level of endemic disease.
8.2 Investigating an outbreak

8.2.1 Establish a case definition

A case definition is the operational definition of a disease for study purposes. A good case definition has two parts: (1) it specifies characteristics shared by all members of the class being defined, and (2) it specifies what distinguishes them from all outside the class. A case definition ensures that the outcome of interest is consistently defined across space (e.g. among different investigation centres in a large scale outbreak) and over time.

In an outbreak of this severe and often fatal pneumonia in delegates attending the 58th annual meeting of the American Legion, Department of Pennsylvania a case was considered Legionnaires’ disease if it met clinical and epidemiologic criteria. The clinical criteria required that a person have onset between 1 July and 18 August 1976, an illness characterised by cough and fever (temperature of 38.9 degrees or higher) or any fever and chest x-ray evidence of pneumonia. To meet the epidemiologic criteria, a patient either had to have attended the American Legion Convention held 21 – 24 July 1976, in Philadelphia, or had to have entered Hotel A between 1 July 1976 and the onset of illness.


8.2.2 Enhance surveillance

When it is suspected that an outbreak is occurring, enhanced surveillance can be useful to identify additional cases. Enhanced surveillance may involve both heightening awareness to increase passive case reports and implementing targeted surveillance. Techniques include directly contacting field practitioners by telephone, facsimile or email, via health department web pages and email discussion groups. For large outbreaks media releases (print, television, radio) can be extremely effective.

8.2.3 Describe outbreak according to individual, place and time

Collect historical, clinical and productivity data on those individuals that are affected (cases) and those that are not affected (non-cases). It is a mistake to concentrate exclusively on diseased animals. If possible, all cases of diseased animals should be included in the investigation. If there are large numbers of unaffected individuals you may select a representative sample of unaffected individuals for examination (controls). You may consider matching controls with some characteristic of the cases e.g. age and gender.

Plot an epidemic curve by identifying the first case (index case) and then graphing subsequent numbers of cases per day or per week from the index case through to the end of the outbreak. An extremely rapid increase in the number of cases from the index case suggests a common source epidemic (all the diseased animals were exposed to the source at about the same time). If the number of disease animals is increasing
over time, this is more indicative of a propagated epidemic which is more typical of contagious disease or prolonged exposure to the agent via vectors or toxins.

Location is often an important risk factor for disease. Draw a sketch map of the area or the layout of the pens and the number of cases within pens. This includes examination of animal movements and recent additions to the herd or flock. The investigator should inspect the drawing for possible interrelationships among cases, and between location of cases and other physical features.

8.2.4 Develop hypotheses about the nature of exposure

At this stage, you will probably have some suspicions about what has caused the outbreak — that is, you will have started to form some hypotheses. Your next job is to test these hypotheses using the various analytical techniques described below.

8.2.5 Conduct analytical studies

Part of the data collection procedure above will have entailed collecting individual-level details such as age, sex, breed, date of parturition, stage of production. Individuals should be categorised according to the presence of each attribute. Attack rate tables divide the cohort of interest into exposed and non-exposed groups. Attack rates are then calculated for each exposure by dividing the number diseased by the group size (Table 8).

<table>
<thead>
<tr>
<th>Food</th>
<th>Exposed</th>
<th>Unexposed</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Ill</td>
<td>Well</td>
</tr>
<tr>
<td>Ham</td>
<td>36</td>
<td>5</td>
</tr>
<tr>
<td>Salad</td>
<td>40</td>
<td>4</td>
</tr>
<tr>
<td>Prawns</td>
<td>16</td>
<td>15</td>
</tr>
</tbody>
</table>

The exposure which is most likely to have served as a vehicle for an outbreak is that with the greatest difference in attack rate for exposed and unexposed individuals. An alternative is to calculate the risk ratio of disease for each exposure. Essentially this is the attack rate for the exposed individuals divided by the attack rate for unexposed individuals — the exposure with the highest risk ratio being the likely vehicle for the outbreak. It is also useful to calculate the population attributable fraction for each exposure. This will identify the percent of the risk of disease in the exposed group that is due to exposure. The closer this value is to 100% the more likely the exposure accounted for the outbreak.

8.3 Implement disease control interventions

At this stage it may be possible to produce a hypothesis regarding the cause of the outbreak. If further investigation is warranted then other epidemiological studies (case-control, prospective cohort etc) may be designed and implemented. You may also use more complex analytical techniques to analyse data already collected (multivariate techniques).
9 Appraising the literature

By the end of this unit you should be able to:

- Describe, in your own words, the four main areas that should be considered when appraising the scientific literature.

Reading the literature is necessary to keep up to date with new developments and to learn more about a particular area of science that interests us.

Fortunately, there appears to be no shortage of literature available to read, and our ability to source this literature easily has been facilitated by the Internet (either in the form of peer-reviewed articles published on-line by established journals or as pre-print publications published by individuals on their own web pages). Although the Internet allows information to be widely disseminated, the quality of that information varies widely. As a result, as good scientists, we need to be discerning about what we read and (more importantly) what we believe. A systematic method of appraising (or evaluating) the literature helps us to do this. These notes outline a systematic approach to appraising the epidemiological literature, which consists of:

- Describing the evidence;
- Assessing the internal validity of the study;
- Assessing the external validity of the study; and
- Comparing the results with other available evidence.

9.1 Description of the evidence

The first step in evaluating a scientific article is to understand exactly what relationship was being evaluated and what hypothesis was being tested. The reader should be able to identify the exposure variable(s) and the outcome variable. It is also necessary to categorise the study in terms of its design (survey, case-control, observational cohort, intervention cohort). Definition of the subjects that were studied in terms of source populations, the eligibility criteria, and the participation rates of the different groups that are being compared.

Having defined the topic of study, it is then useful to summarise the main result — what is the result in terms of the association between exposure and outcome? It should be possible to express the main result in a simple table and obtain from the paper the means to calculate the appropriate measure of association (relative risk, odds ratio, difference in proportions) and the appropriate test of statistical significance.
9.2 Internal validity

9.2.1 Non-causal explanations

Having described the study the next step is to assess its internal validity — that is, for the subjects who were studied, does the evidence support a causal relationship between the exposure and the outcome? We consider the three possible non-causal mechanisms which could produce the observed results:

- Are the results likely to be affected by bias?
- Are the results likely to be affected by confounding?
- Are the results likely to be affected by chance variation?

It is useful to consider each of these aspects separately. The order of these non-causal explanations is important. If there is severe observation bias, no analytical manipulation of the data will overcome the problem. If there is confounding, then appropriate analysis will (in most cases) overcome the problem. The assessment of chance variation should be made on the main result of the study, after considering issues of bias and confounding.

9.2.2 Positive features of causation

Is there a correct temporal relationship? For a relationship to be causal, the putative exposure must act before the outcome occurs. In a prospective study design where exposed and non-exposed subjects are compared, this requirement is established by ensuring that subjects do not already have the outcome of interest when the study starts. The ability to clarify time relationships is weaker in retrospective studies, and care is required to ensure that possible causal factors did in fact occur before the outcome of interest.

A difficulty in all study designs, but more so in retrospective studies, is that the occurrence in biological terms of the outcome of interest may precede the recognition and documentation of that outcome by a long and variable period of time (e.g. some cancers).

Is the relationship strong? A stronger association, that is a larger relative risk, is more likely to reflect a causal relationship. As a measured factor gets closer to a biological event on the causal pathway, the relative risks become larger.

The fact that a relationship is strong does not protect us against certain non-causal relationships, however if the relationship that is observed is due to bias, then the bias must be large and therefore easy to identify. If a strong relationship is due to confounding, either the association of the exposure with the confounder must be very close, or the association of the confounder with the outcome must be very strong.
Is there a dose-response relationship? In some circumstances the demonstration of a smooth dose-response relationship may be a strong argument against an identified relationship arising as a result of bias. In general, we should expect uni-directional dose-effect relationships and evidence that this is not the case should be considered carefully.

Consistency of the association? A causal relationship will be expected to apply across a wide range of subjects. An association identified in one study that is consistent with the same association identified in a different groups of subjects is supportive of causation. The difficulty with consistency is that very large data sets are required to assess the similarity or otherwise of associations in different subgroups of subjects. Even with adequate numbers, the subgroups to be compared need to be defined on a priori grounds.

Specificity of association? It has been argued that a specific association between one causal factor and one outcome (i.e. exposure to the defined causal factor results in a specific syndrome), is good evidence for causality.

An argument against the negative health effects of smoking arose from the observation that smoking was shown to be associated with the occurrence of a number of cancers and other serious diseases and therefore demonstrated non-specificity of action, making the hypothesis of a causal link with lung cancer less likely.

Specificity may be useful, if we do not make it an absolute criterion, as one causal agent may in truth produce various outcomes, and one outcome may result from various agents. The concept is often useful in study design: as a check on response bias we may deliberately collect information on factors which we expect to be the same in groups that we are comparing (similar results across groups will indicate a lack of observation bias).

9.3 External validity

If the internal validity of a study is poor, then there is no point in proceeding further — if the results are not valid for the subjects that were studied, its application to other groups of subjects is irrelevant.

9.3.1 Can the results be applied to the eligible population?

The relationship between the study population (those that participated in the study) and the eligible population (those that met the study inclusion criteria but did not take part) should be well documented. Losses due to non-participation have to be considered carefully as they are likely to be non-random, and the reasons for the losses may be related to the exposure or the outcome.
9.3.2 Can the results be applied to the source population?

The important issue is not whether the subjects studied are ‘typical’, but whether the association between outcome and exposure given by the study participants is likely to apply to other groups. In assessing this applicability, we need to be specific about the factors which are likely to affect the association.

Most clinical trials are done on patients in teaching hospitals. If a new therapy for a particular type of neoplasia is shown to be effective in such a trial, we would readily apply the results to patients in a district hospital who had a similar stage and type of tumour and were of similar age, even though the trial patients cannot be said to be representative of district hospital patients in a general or statistical sense.

9.3.3 Can the results be applied to other relevant populations?

In general, the difficulties of applying results from one groups of subjects to another will be minimal for issues of basic physiology and maximal for effects in which cultural and psycho-social aspects are dominant.

9.4 Comparison of the results with other evidence

For many clinical questions a large amount of evidence is available which comes from different types of studies. In these circumstances it is useful to consider a hierarchy of evidence. Given that studies are adequately performed within the limitations of the design used, the reliability of the information from them can be ranked as follows:

1. Randomised trials.
2. Cohort and case-control studies.
3. Other comparative studies.

Randomised clinical trials, if properly performed on adequate numbers of subjects, provide greatest evidence because of the unique advantages in overcoming problems of bias and confounding.

9.4.1 Consistency

This is the most important characteristic used in the judgement that an association is causal. To say that the result is consistent requires that the association has been observed in a number of different studies, each of which individually can be interpreted as showing a causal explanation, and which have enough variation in their methodology and study populations to make it unlikely that the same biases or confounding factors apply in all the studies. Lack of consistency argues against causality.
9.4.2 Specificity

Whether a difference in results between two studies is interpreted as inconsistency or as specificity depends on whether the difference is anticipated by a hypothesis set up before the comparison is made. If not, but a plausible mechanism can be found or if the difference itself found consistently, then the hypothesis may be modified to take into account the specificity which has been shown.

9.4.3 Plausibility

Plausibility refers to the observed association being biologically understandable on the basis of current knowledge concerning its likely mechanisms.

However, any dramatically new observation may be in advance of current biological thinking and its lack of plausibility may reflect deficiencies in biological knowledge rather than error in observation. For example:

- John Snow effectively prevented cholera in London 25 years before the isolation of the cholera bacillus and the general acceptance of the principle that the disease could be spread by water.

- Percival Pott demonstrated the causal relationship between exposure to soot and scrotal cancer some 150 years before the relevant carcinogen was isolated.

9.4.4 Coherency

An association is regarded as coherent if it fits the general features of the distribution of both the exposure and the outcome under assessment; thus if lung cancer is due to smoking, the frequency of lung cancer in different populations and in different time periods should relate to the frequency of smoking in those populations at earlier relevant time periods.

If the exposure variable under study causes only a small proportion of the total disease, the overwhelming influence of other factors may make the overall pattern inconsistent.
10 Exercise: outbreak investigation

This exercise has been adapted from Gardner (1990b).

A veterinarian in a mixed practice has been investigating an ongoing diarrhoea problem in neonatal pigs in a 150-sow breeding/finishing herd. In the 12 months prior to the outbreak, 7% of litters had diarrhoea but over recent weeks the proportion of litters affected has increased to about 40%. As part of the investigation the veterinarian submitted 3 acutely affected pigs to the regional diagnostic laboratory. Of the 3 pigs, 1 was infected with \textit{E. coli} serotype 08 but other pathogenic bacteria and viruses were not isolated from the other 2 pigs. Lesions in all 3 pigs were consistent with an acute enteritis. The veterinarian asks you to assist.

As background to the problem, the veterinarian provides you with a map showing the layout of the sheds, a description of normal management procedures, and recent records for farrowing sows as detailed below:

10.1 The problem

\textit{Shed design.} The shed has 16 concrete-floored pens (oriented in a single row in a west-east direction. Pen 1 is near the entrance door at the western end of the shed and pens run in numerical sequence to pen 16 which is located near the extraction fans. The pit underneath the sows is flushed at least twice daily. During the study, pen 14 was under repair and was not used.

\textit{Management - treatments.} Sows are moved into cleaned and disinfected pens in the farrowing shed on about day 110 of gestation. Sows farrow with minimal supervision. On the first day of life, pigs have their needle teeth clipped and are provided with heat lamps. No vaccines are given to sows or baby pigs for control of enteric disease. Sows are fed \textit{ad libitum} during lactation with a high energy ration (15.5 MJ DE/kg). During gestation, they are fed about 2.0 to 2.5 kg of a lower energy ration plus about 0.5 kg/day of recycled manure for control of enteric infections and parovirus. Piglets in litters with diarrhoea are treated with oral furazolidone and electrolytes are offered \textit{ad libitum} in shallow bowls in each pen.

\textit{Records.} Records are provided from a recent set of 26 farrowings (April 2002) for you to examine before your visit. Before April 2002 the records of diarrhoea were insufficiently detailed to be of value in the current investigation.

10.2 Question 1

How valid are owner-diagnoses of scours-related deaths? How could you improve their validity in the future?
<table>
<thead>
<tr>
<th>Litter</th>
<th>Pen</th>
<th>Sow</th>
<th>Parity</th>
<th>Farrow</th>
<th>Born</th>
<th>Weaned</th>
<th>Death due to</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td>1</td>
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<td>169</td>
<td>4</td>
<td>30 Apr 02</td>
<td>11</td>
<td>10</td>
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</tr>
<tr>
<td>Total</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>253</td>
<td>205</td>
<td>16</td>
</tr>
</tbody>
</table>

a Sow sick at farrowing.

### 10.3 Question 2

Estimate the following rates from the data:

- The scours-specific mortality rate.
- The proportional mortality rate for scours.
- The case fatality rate for scours.
- The proportion of litters affected with scours.
- The preweaning mortality rate.
10.4 Question 3

Outline your approach to investigating this diarrhoea problem (at this stage there is no need to calculate any factor-specific rates). What initial conclusions or hypotheses did you formulate after examining the history and laboratory findings, and temporal and spatial patterns of disease?

10.5 Question 4

Analyse the records from the 26 April farrowings and calculate some factor-specific rates or relative risks either by hand or by using computer software available for that purpose. For example:

- What was the relative risk of scours in parity 1 litters, compared with litters from all other parities?
- What was the relative risk of scours in litters from sick sows, compared with litters from healthy sows?
- What was relative risk of scours in large litters, compared with small litters?
- What was the relative risk of scours in litters born in pens 1 - 8, compared with litters born in pens 9 - 16?

Test the statistical significance of the difference between the two rates in each case. How helpful are the data in allowing you to formulate better hypotheses? Could confounding be a problem and how would you deal with it at this stage of the study?

Data may be presented in a $2 \times 2$ table format as follows:

<table>
<thead>
<tr>
<th></th>
<th>Diseased</th>
<th>Non-diseased</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Exposed</td>
<td>$a$</td>
<td>$b$</td>
<td>$a + b$</td>
</tr>
<tr>
<td>Non-exposed</td>
<td>$c$</td>
<td>$d$</td>
<td>$c + d$</td>
</tr>
<tr>
<td>Total</td>
<td>$a + c$</td>
<td>$b + d$</td>
<td>$a + b + c + d$</td>
</tr>
</tbody>
</table>

We are interested in testing the hypothesis that the proportion of exposed individuals that are disease positive differs from the proportion of non-exposed individuals that are disease positive. Because this is nominal (count) data, a chi-squared test is the appropriate method to test this hypothesis. This involves three steps:

1. A statement of the null hypothesis: ‘The proportion of exposed individuals that are diseased does not differ from the proportion of non-exposed individuals that are diseased.’

2. Calculation of a chi-squared test statistic. Using the above notation, the formula for the chi-squared test statistic for data presented in a $2 \times 2$ table is:
\[ \chi^2 = \frac{n(ad - bc)^2}{(a + c)(b + d)(a + b)(c + d)} \]  

3. We will use an alpha level of 0.05 to test this hypothesis and apply a one-tailed test. Specifying an alpha level of 0.05 means that there is a 5% probability of incorrectly rejecting the null hypothesis (when it is in fact true). The critical value that separates the upper 5% of the \( \chi^2 \) distribution with 1 degree of freedom from the remaining 95% is 3.841 (from statistical tables). Thus, if our calculated chi-squared test statistic is greater than 3.841 we can reject the null hypothesis and accept the alternative hypothesis, concluding that the proportions diseased among exposed and non-exposed individuals differ.

### 10.6 Question 5

What recommendations, if any, would you make to your colleague and to his client based on your findings (without the data from the clinical trial or cohort study)?

### 10.7 Question 6

Design either a clinical trial or a prospective cohort study to test one of your hypotheses in detail.

### 10.8 Question 7

Estimate the financial impact of the losses due to diarrhoea in this set of 26 litters. The following data has been provided:

<table>
<thead>
<tr>
<th>Item</th>
<th>Value</th>
<th>Target</th>
</tr>
</thead>
<tbody>
<tr>
<td>Percent of litters with scours in 12 months before outbreak</td>
<td>7%</td>
<td>&lt; 5%</td>
</tr>
<tr>
<td>Preweaning mortality in 12 months before outbreak</td>
<td>11.5%</td>
<td>&lt; 12%</td>
</tr>
<tr>
<td>Post weaning mortality</td>
<td>5%</td>
<td>&lt; 3%</td>
</tr>
<tr>
<td>Gross margin per pig marketed</td>
<td>$35.00</td>
<td>-</td>
</tr>
<tr>
<td>Treatment costs per litter</td>
<td>$10.00</td>
<td>-</td>
</tr>
<tr>
<td>E. coli vaccine</td>
<td>2 \times $2.50</td>
<td>-</td>
</tr>
<tr>
<td>Labour cost to vaccinate one pig</td>
<td>$0.30</td>
<td>-</td>
</tr>
</tbody>
</table>
# 11 Review questions

You are discharging a 2 year-old male domestic shorthair cat who has spent 10 days in your clinic recovering from the complications associated with obstruction of the urinary tract. As the cat’s owner is writing out a cheque for $1500 he asks ‘will my cat experience another attack of FUS in the future and what can I do to prevent it?’ What advise would you give, from an epidemiological perspective?

Think about three or four health problems or diseases that you or your friends have had. List each of the host, agent, and environmental factors that may have been causative for each disease you have listed.

Can you think if circumstances when exposure to a causal factor does not change disease incidence?

List five or six broad and fundamental influences on health and disease, that is, those influences that change the population patterns of disease.

Reflect on some medical and public health activities which were widely practiced but are now known to be wrong, some dangerously so. Your reflection should include some historical activities say, before the turn of the twentieth century and more recent ones. Also reflect on some current policies and practices that may meet the same fate.

Imagine you are in a country where no animal demographic data is available. An epidemic of pneumonia is suspected in the cattle population. You are asked to develop a plan to prevent and control the epidemic. Which questions do you need to answer to start a rational control strategy for this disease? Which epidemiological data do you need to answer the questions?

What benefits are there from investigating the changes in disease frequency in a population over time?

Consider the reasons why a variation in disease pattern might be artefact rather than real. Can you group them into three or four categories of explanation? What explanations can you think of for a real change in disease frequency? Can you group these into three or four categories of explanation?

Imagine you are asked to describe the health status of a population of animals to a senior public servant. The person you are talking to has no previous background in animal (or human) health. What kinds of measures would you choose to portray the health of the animal population? Consider not only the specific types of data, but also the qualities of the data you would seek out.

Imagine a population of 10,000 new army recruits. You are interested in studying the incidence and prevalence rate of gunshot wounds on war duty. Assume all gunshot wounds lead to permanent visible damage. You follow the recruits for one year. All of the study population survive, all medical records are available, and all recruits are available to interview and examination. Assume the occurrence of gunshot wounds is
An Introduction to Veterinary Epidemiology

spread evenly through the year, and that at the time of entering the army, no recruits had gunshot wounds. Over the year you determine that 20 recruits had a gunshot wound.

• What is the incidence risk of gunshot wounds? What is the incidence rate of gunshot wounds?
• What is the point prevalence rate of having had a gunshot wound at the beginning, middle, and end of the year?
• What is the period prevalence rate over the year?
• If the incidence rate remains the same over time, what is the prevalence rate of ever being scarred by the end of five years?
• What is the average duration of a gunshot wound, among those scarred, by the end of the first year?
• What is the estimated point prevalence rate over the five-year period?

What might be your denominator for a study defining the incidence rate of:

• Calf mortality.
• Clinical mastitis.
• Bovine spongiform encephalopathy.

Reflect on the terms ‘risk factor’ and ‘cause of disease’. What is the difference between these terms?

Consider why the risk ratio might provide a false picture of the effect of a risk factor on disease and hence the strength of association.

Imagine that the incidence of chronic obstructive pulmonary disease (COPD) in horse is compared in two areas of a country: one with polluted air (A) and the other not (B). In the polluted area there were 20 cases of COPD in a population of 100,000. In the other area there were 10 cases in a population of 100,000.

• What is the risk ratio of COPD in area (A)?
• What is the risk ratio of COPD in area (B)?
• Do we know the precision of these estimates of risk ratio?
• What explanations are there for the risk ratio estimate in area (A)?
• What questions will you need to consider before concluding that there is a real association between pollution and COPD?

Imagine that exposure to a dry cat food triples the incidence of a feline urologic syndrome (FUS), that is, the risk ratio is 3. This disease has a baseline incidence of 1 per cent per year in the non-exposed group. Imagine also that the baseline incidence is double in castrated male cats (that is, 2 per cent) and that the risk ratio associated with exposure to dry cat food is the same, three. You follow 100 entire and 100 castrated male cats that are fed dry cat food, and an equivalent number of cats fed moist food. The study lasts for 5 years. Create a $2 \times 2$ table to show the data for castrates and entire male cats and calculate the odds ratio of disease in the exposed group in relation to those not exposed. Compare the odds ratio with the risk ratio of 3.

The Ministry of Health has made available a sum of $100,000 for a health promotion programme to reduce coronary heart disease mortality. We can spend it on encouraging people to stop smoking or encouraging them to do more exercise. Assume the risk ratio associated with both risk factors is 2, that changes in prevalence rate are equally permanent, and that the cardioprotective effect occurs quickly. Which choice will give a better return in lives saved?

• First, make a judgement on which of the two preventive programs you prefer.
• Now consider which is more common: smoking or lack of exercise?
• Calculate the population attributable risk when the prevalence rate of smoking is 20%, 30%, 40% and 50% and the prevalence rate of lack of exercise is 60%, 70%, and 80% (these are realistic prevalence rates in industrialised countries). Has the result altered or substantiated your earlier judgement?

Imagine a cohort study which aims to determine the incidence of arthritis in large breeds of dogs. The follow-up period for the study is five years. Describe the advantages and disadvantages of the two approaches for measuring incidence.

Imagine a study of the incidence of congestive heart disease in large breeds of dogs, base on post mortem records collected at a University teaching hospital over a five-year period. Again, consider the advantages and disadvantages of the two approaches for measuring incidence.

Is there a difference between a clinical case series and a population case series?

How might epidemiology study the potential role in disease causation of factors which vary little between individuals within a region or country. For example: fluoride content of water, hardness or softness of water supplies, annual exposure to sunshine?

What is the essential feature that differentiates a cross-sectional study from a cohort study?
Explain what you understand by the term ‘error’. What is the difference, if any, between error and bias?

A client of your manages a study beef herd which, for the past ten years, has consistently tested negative for tuberculosis. A positive reactor has been found after the latest round of testing. What would you advise?
12 Epidemiological resources

EpiCentre, Massey University  
http://epicentre.massey.ac.nz/

University of Guelph, Department of Pop Medicine  
http://www.ovc.uoguelph.ca/PopMed/

Atlantic Veterinary College Epidemiology Group  
http://www.upei.ca/~avc/health/epi.htm

Royal Veterinary College, University of London  
http://www.rvc.ac.uk/

University of Michigan School of Public Health  
http://www.sph.umich.edu/epid/

Epidemiology Monitor  
http://www.epimonitor.net/

Association of Teachers of Veterinary Public Health  
http://www.cvm.uiuc.edu/atvphpm/

Epidemiology for the uninitiated — BMJ  
http://www.bmj.com/epidem/

Centers for Disease Control and Prevention  
http://www.cdc.gov/

EXCITE  
http://www.cdc.gov/excite/

Epidemiology Supercourse  
http://www.pitt.edu/~super1/

VEIN links: Evidence Based Medicine  
http://vein.library.usyd.edu.au

Post Graduate Foundation in Veterinary Science  
http://www.pgf.edu.au/

EBM Resources  
http://www.dartmouth.edu/~biomed/

MAF, New Zealand  
http://www.maf.govt.nz

AFFA, Australia  
http://www.affa.gov.au

Canadian Food Inspection Agency  
http://www.inspection.gc.ca

Health Canada  
http://www.hc-sc.gc.ca/

International EpiLab  
http://www.dfvf.dk/

The Cochrane Collaboration  
http://www.cochrane.org/index0.htm
References


