Friday



Applied Statistics: Lectures 2 & 3 (1)

2018/19

Applied Statistics Lectures 2 & 3

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Applied Statistics: Lectures 2 & 3 (2)

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Outline

Last lecture

- ₩ Why statistics?
- Introduction to design of experiments
- Parametric versus non-parametric models

This lecture

- Statistical inference
- Introduction to hypothesis testing

OpenIntro Statistics

Chapters 2-4, particularly §4.3



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Probability

Probability (Wikipedia)

Probability describes the plausibility of (random) observed data —
assumed to be described by a statistical model, and specified parameter
value(s) — without reference to any observed data

Different (practical) interpretations of probability

- Classical often connected with gambling, dice, cards, etc
- Frequentist use a large number of trials to estimate probabilities
- Bayesian "reasonable expectation" or "belief" (subjective)

No interpretation fits all physical scenarios!



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Likelihood

Likelihood (Wikipedia)

Likelihood describes the plausibility, given specific observed data, of a parameter value of the statistical model which is assumed to describe that data

- Statistics mostly deals with likelihoods
- Informally probabilities and likelihoods are used interchangeably
- Maximum likelihood for finding parameters of statistical models
- Histograms versus probability density functions

freq 1 John model



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Statistical inference

Definition from Wikipedia

Statistical inference makes propositions about a population, using data drawn from the population with some form of sampling. Given a hypothesis about a population, for which we wish to draw inferences, statistical inference consists of (first) selecting a statistical model of the process that generates the data and (second) deducing propositions from the model.

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Interview selection

A leading FTSE 100 company claims to have a policy of equal opportunity for both men and women on its graduate training scheme. However, upon auditing it was found that during their last recruitment round they appointed 19 women and only 1 man despite roughly equal numbers of men and women applying.

Is the selection process fair to both men and women?

Suhere is the dividing line? Is 15 women - 5 men?

(What about the design of the experiment?!)

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Construct the hypothesis

First decide on a *default position* (e.g., innocent until proven guilty) — this is the null hypothesis

Suhat do we think is true in the absence of any information? (Bayesian: prior belief)

Null hypothesis H_0 the interview selection process is fair to both men and women. Forms the basis of a stehishcal model.

Alternative hypothesis H₁ the interview selection process is biased.

Sopposite Ho

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Interview selection — a statistical model

What is the individual/population/sample?
individual: outcome of a recruitment round.
population: all recruitment rounds
What does fair actually mean?
What does <i>fair</i> actually mean?
Equally probable to select a mon or a women.

What is a statistical model that could generate data for an individual observation?

X: number of women recruited

X ~ Bin (20, 0.5)

trials prob. of success

generates probabilities for each outcome.

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Visualising a statistical model

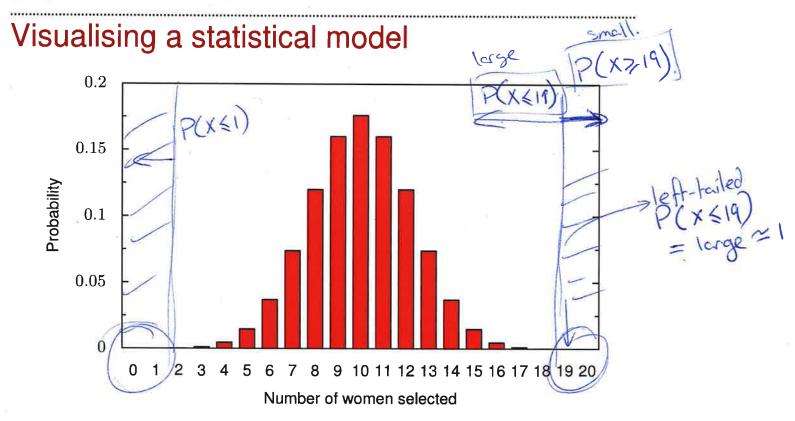
# Women	Probability			
0	9.5×10^{-7}	# Women	Probability	\$
1	$\begin{array}{c c} 3.5 \times 10^{-5} \\ 1.9 \times 10^{-5} \end{array}$	11	0.16	ae of
1	l .	12	0.12	or From zeroth
2	1.8×10^{-4}			S reprobability
3	1.1×10^{-3}	13	0.074	Sign
4	4.6×10^{-3}	14	0.037	
		15	0.015	
5	0.015			prob of this
6	0.037	16	4.6×10^{-3}	outcome
		17	1.1×10^{-3}	=0.
7	0.074		1.8×10^{-4}	
8	0.12	18		C
9		19	1.9×10^{-5}	
	0.16	11 20	9.5×10^{-7}	
10	0.18 e most probe	eble 20	3.5 × 10	
,	mutcome			

Consider probability of the atrone or something even more extreme -> 1.9x105 + 9.5x107 = small!



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Statistical significance

Testing the hypothesis involves calculating the probability of the sample occurring assuming that the null hypothesis is correct

When is the measurement (or a measurement even more extreme) so improbable that we say the null hypothesis is incorrect? *This is the (statistical) significance level.*

> 5% is very common

11. . 1.1170

Criminal trials in the UK?

Sinnocent until proven guilty.

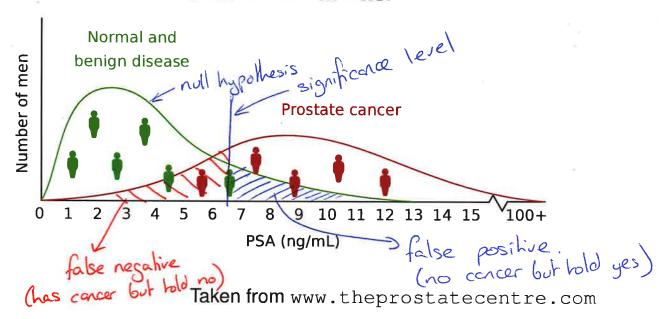
Scientific discoveries (e.g., Higgs Boson)?

6 Physics: 50 N10-7

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Where to draw the line?

PSA cancer marker



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Errors in hypothesis testing

Since hypothesis testing is a probabilistic technique, invariably there can be errors. There are two main types of error to consider.

Type 1 (false positive) These errors occur when you incorrectly reject the null hypothesis, i.e., you think a change has occurred when it hasn't. Since the null hypothesis was actually correct we know that these types of error occur at the same level as the significance level chosen (e.g., 5%).

Type 2 (false negative) These errors occur when you incorrectly accept the null hypothesis, i.e., you think a change has not occurred when it has. Since the null hypothesis was incorrect, we don't know the distribution of errors or the percentage of errors.

Decreasing type 1 errors will almost certainly increase type 2 errors!

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Exercises

For each of the following, decide what the null hypothesis H_0 and the alternative hypothesis H_1 should be.

What would be a type 1 error and a type 2 error in each case?

- 1. Drinking coffee will make us live longer
- 2. The medical treatment improved the patient's condition
- 3. Product A is more efficient than product B



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Testing the hypothesis — p-value 5 Probability of seeing the data or something more extreme.

To test the hypothesis, calculate the p-value

 $p = 2 \min(P(X \ge x), P(X \le x)) - \text{two-tailed (two-sided) test}$ $p = P(X \le x) - \text{left-tailed (one-sided) test}$ $p = P(X \le x) - \text{left-tailed (one-sided) test}$

 $\Rightarrow k p = P(X \leq x)$ — left-tailed (one-sided) test

 $\not k p = P(X \geqslant x)$ — right-tailed (one-sided) test

By default use a two-tailed test unless you're sure that you only care about differences in one direction.

> don't are about being biased bounds but do are about being biased bowerd's women.

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Interview selection — testing the hypothesis

Use a statistical significance level of 5% (i.e. 0.05)

For the interview selection we have

$$X \sim Bin(n, p), \quad P(X = x) = \binom{n}{x} p^x (1-p)^{n-x}$$

with n = 20, p = 0.5. Binomial coefficient is

$$\binom{n}{x} = \frac{n!}{x!(n-x)!}$$

Need to calculate

$$p = 2 \min(P(X \leq \frac{1}{4}), P(X \geq \frac{1}{4})) = 4.0 \times 10^{-5}$$
 < 5%.

Reject null hypothesis.

Accept the alternative hypothesis.



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Interview selection — exercise

A morally dubious colleague is interviewing and they want to know how biased their selection can be (women:men) without coming under suspicion. They are told to select 5 people in total.



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Statistical assumptions — independence

The statistical model for interview selection like many statistical models assumes independence (where?)

What if 16 of the 20 people selected in the last recruitment round had graduated from Mechanical Engineering at Bristol?

Has anything changed?

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Quote of the day

Mark Twain

It ain't what you don't know that gets you into trouble.

It's what you know for sure that just ain't so.



Exercises

¥ 4.17–4.20 from OpenIntro Statistics