

# GEOG 5680

## Introduction to R

### 09: Probabilities and Inference tests

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# Probability

What is this thing called probability?

- Mathematical description of uncertainty
- Tightly linked to statistics for inference
  - Model of population from samples
- Several functions in R for estimating probability
- Also found as part of the inference in test in other functions (e.g. ANOVA, linear models, etc)

# Probability

What is this thing called probability?

- Probability shows what *outcomes* might occur given a *model*
  - Given the animal, what are the footprints?

# Probability

What is this thing called probability?

- Probability shows what *outcomes* might occur given a *model*
  - Given the animal, what are the footprints?
- Statistics show what *models* might result in a given *outcome*
  - Given the footprints, what is the animal?

# Distributions in R

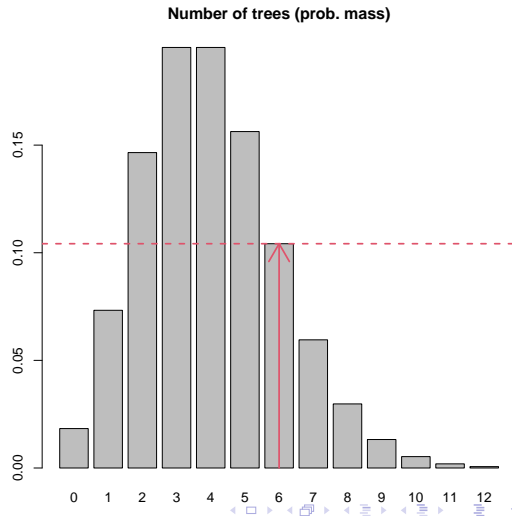
- R comes as standard with approx. 20 well-known probability distribution functions
- Including normal, uniform, binomial, log-normal, beta, gamma,  $t$ ,  $F$ ,  $\chi^2$  etc
- Add-on packages include approx 100+ extra distributions
- Most distribution come with four functions:
  - $d^*$  — density functions (e.g. `dnorm()`)
  - $p^*$  — probability distribution functions (e.g. `pnorm()`)
  - $q^*$  — quantile functions (e.g. `qnorm()`)
  - $r^*$  — random number generation (e.g. `rnorm()`)
- Look at examples with Poisson (discrete, count) and normal (continuous)

# Poisson distribution

- Count data ( $\lambda$  = mean count)
- `d*`: density function, gives the height of the density curve for a given value
- E.g what is the probability of getting 6 trees in a quadrat?

```
dpois(6,lambda=4)
```

```
## [1] 0.1041956
```

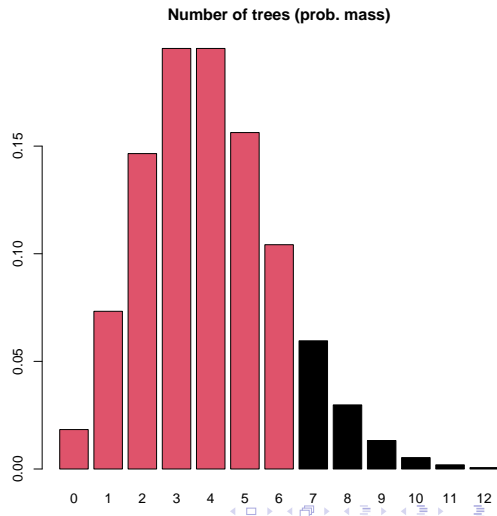


# Poisson distribution

- Count data ( $\lambda$  = mean count)
- `p*`: probability dist. function, gives the integral above or below that value
- E.g what is the probability of getting  $\leq 6$  trees in a quadrat?

```
ppois(6,lambda=4,  
      lower.tail = TRUE)
```

```
## [1] 0.889326
```

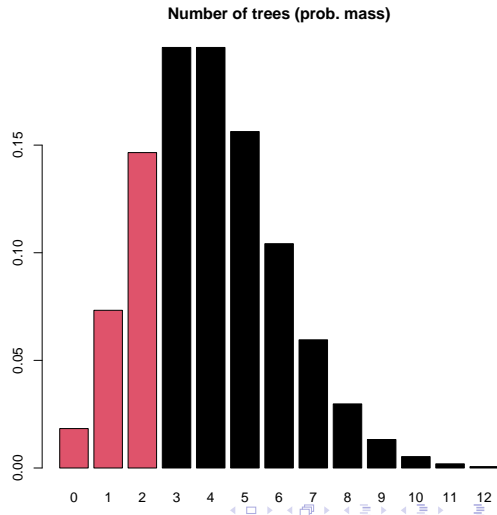


# Poisson distribution

- Count data ( $\lambda$  = mean count)
- `q*`: quantile function, gives the values of  $X$  corresponding to a percentile probability
- E.g how many trees do we expect at the 10 percentile of the distribution?

```
qpois(0.1, lambda=4)
```

```
## [1] 2
```



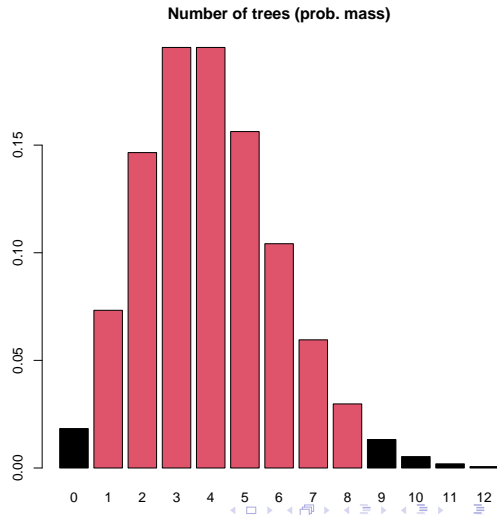


# Poisson distribution

- Count data ( $\lambda$  = mean count)
- `q*`: quantile function, gives the values of  $X$  corresponding to a percentile probability
- E.g what is the 95% CI on the number of trees we expect?

```
qpois(c(0.025,0.975),lambda=4)
```

```
## [1] 1 8
```



# Poisson distribution

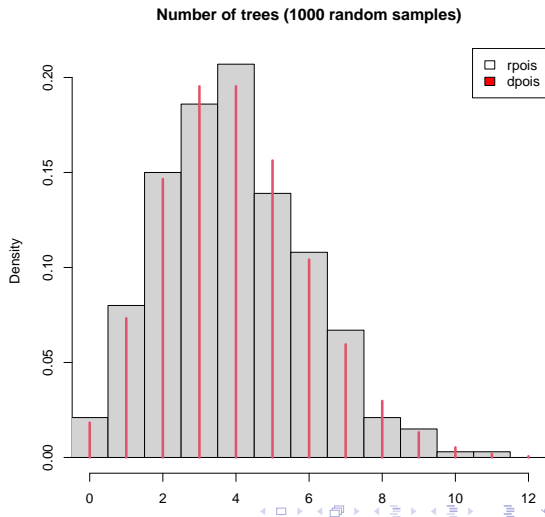
- Count data ( $\lambda = \text{mean count}$ )
- `r*`: random function, generates random samples from the distribution
- E.g how many trees might be found in the next four plots?

```
rpois(4,lambda=4)
```

```
## [1] 4 4 6 3
```

```
rpois(4,lambda=4)
```

```
## [1] 4 5 4 2
```

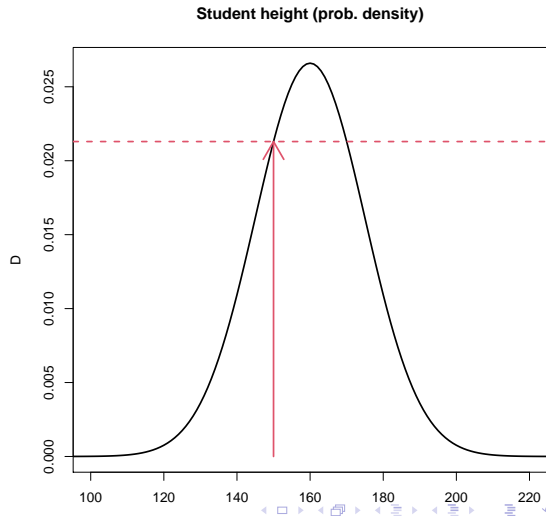


# Normal distribution

- Continuous data ( $\mu$  = mean,  $\sigma$  = std.dev.)
- `d*`: density function, gives the height of the density curve for a given value
- E.g what is the probability density for a height of 150cm?

```
dnorm(150, mean = 160, sd=15)
```

```
## [1] 0.02129653
```

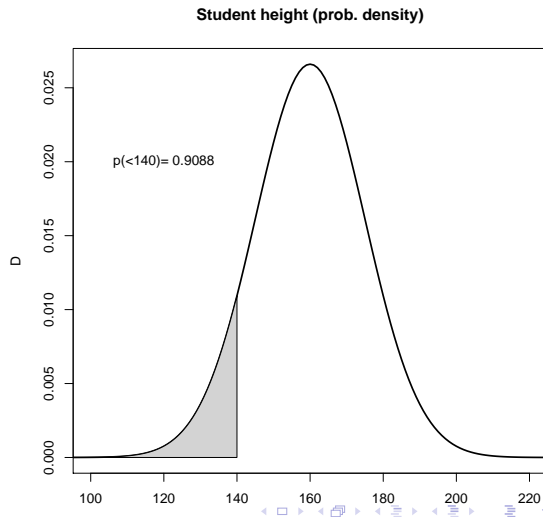


# Normal distribution

- Continuous data ( $\mu$  = mean,  $\sigma$  = std.dev.)
- `p*`: probability dist. function, gives the integral above or below that value
- E.g what is the probability of a student being smaller than 140cm?

```
pnorm(140, 160, 15,  
      lower.tail = TRUE)
```

```
## [1] 0.09121122
```

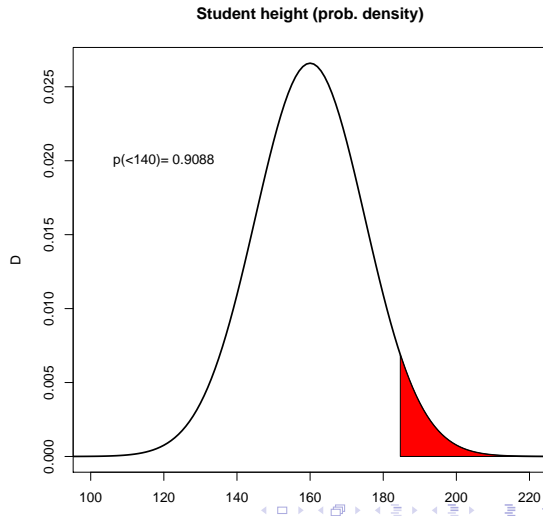


# Normal distribution

- Continuous data ( $\mu$  = mean,  $\sigma$  = std.dev.)
- `q*`: quantile function, gives the values of  $X$  corresponding to a percentile probability
- E.g what cutoff in height gives me the top 5% of students?

```
qnorm(0.95, 160, 15)
```

```
## [1] 184.6728
```



# Normal distribution

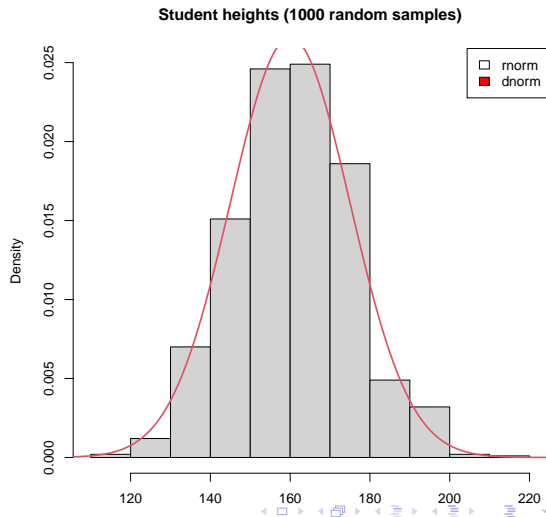
- Continuous data ( $\mu$  = mean,  $\sigma$  = std.dev.)
- `r*`: random function, generates samples from the distribution
- E.g heights of 3 random students?

```
rmnorm(3, 160, 15)
```

```
## [1] 145.6133 168.2708 153.2355
```

```
rmnorm(3, 160, 15)
```

```
## [1] 143.2768 159.0669 166.8958
```



# Statistical Inference

## Statistical Inference and hypothesis testing

- Test some assumptions about a population of interest, using data drawn or sampled from that population
- Compared to descriptive statistics, inference gives significance of a statistical observation
- Examples
  - Do two sets of observations have the same characteristics (mean, variance)?
  - Are two variables correlated among a set of observations?
  - Are observations distributed equally or not?

# Student's $t$ -test

- A  $t$ -test is used to compare an observed sample mean ( $\mu_1$ ) to a hypothesized value ( $\mu_0$ ) (one sample  $t$ -test)
- Or to compare two sample means (two sample  $t$ -test)

$$t = \frac{\mu_1 - \mu_2}{s_{\mu_1 - \mu_2}} \quad (1)$$

- One-tailed ( $\mu_1 < \mu_2$  or  $\mu_1 > \mu_2$ )
- Two-tailed ( $\mu_1 \neq \mu_2$ )

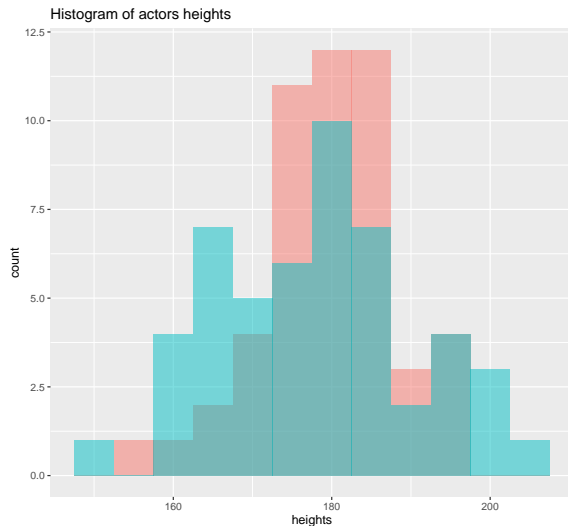




# Student's $t$ -test

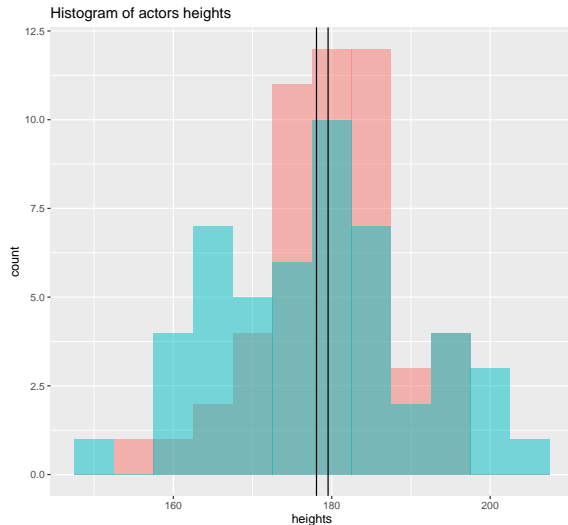


- Two samples ( $n = 50$ ) of actors who auditioned for the role of Aragorn in Lord of the Rings in two different locations
- Is there a difference in heights?



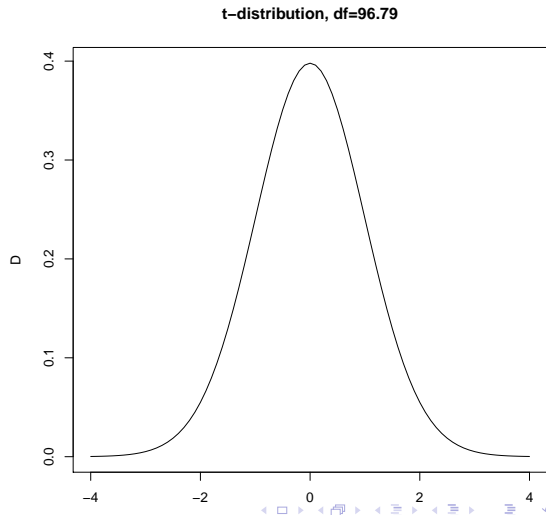
# Student's $t$ -test

- Is there a difference in mean height?
- Loc. 1 mean = 179.54
- Loc. 2 mean = 178.07
- Difference = 1.46
- $t$ -statistic = 0.6886



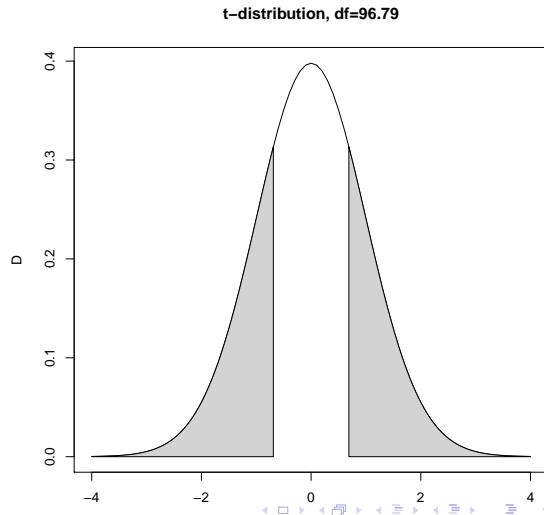
# Student's $t$ -test

- Compare  $t$ -statistic to  $t$ -distribution
- Represents the range of  $t$ -statistics expected through normal random variation
- If observed  $t$  has a low probability (i.e. in one of the tails), it is less likely to have occurred by chance ( $p$ -value)



# Student's $t$ -test

- Two-tail test:
- The  $p$ -value represents the probability that this *difference* (positive or negative) could have occurred by chance
  - $p$ -value is integral of curve  $< -|t|$  plus integral of curve  $> |t|$
  - $p$ -value = 0.4929



# Student's $t$ -test

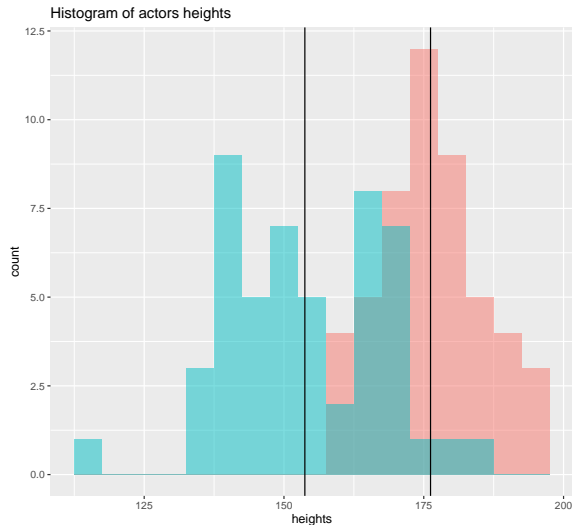
$t$ -test in R using the `t.test()` function:

```
t.test(apop1, apop2, alternative = "two.sided")

##
##  Welch Two Sample t-test
##
## data:  apop1 and apop2
## t = 0.68864, df = 87.394, p-value = 0.4929
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -2.761618  5.690039
## sample estimates:
## mean of x mean of y
## 179.5392 178.0750
```

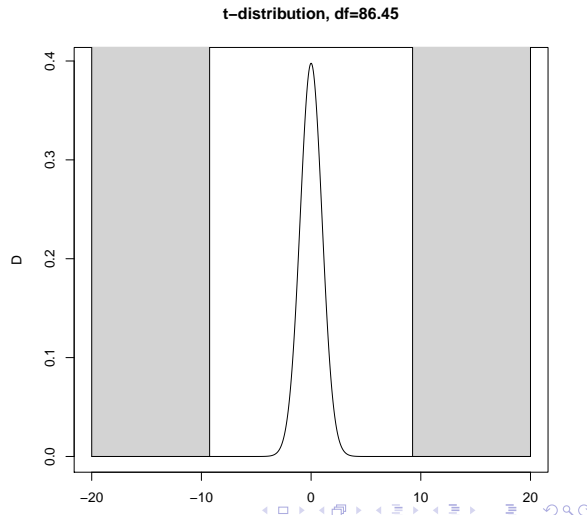
# Student's $t$ -test

- If we also had 50 actors who auditioned for Gimli + 50 who auditioned for Aragorn
- Difference = 22.48
- $t$ -statistic = 9.2512



# Student's $t$ -test

- The  $p$ -value represents the probability that this value (or larger) could have occurred by chance
- Two-tail test:
  - $p$ -value is integral of curve  $< -|t|$  plus integral of curve  $> |t|$
  - $p\text{-value} = 1.4492108 \times 10^{-14}$



# Student's $t$ -test

$t$ -test in R using the `t.test()` function:

```
t.test(aragorn, gimli, alternative = "two.sided")

##
##  Welch Two Sample t-test
##
## data:  aragorn and gimli
## t = 9.2512, df = 86.451, p-value = 1.449e-14
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  17.64770 27.30697
## sample estimates:
## mean of x mean of y
##  176.2119  153.7345
```



# Student's $t$ -test

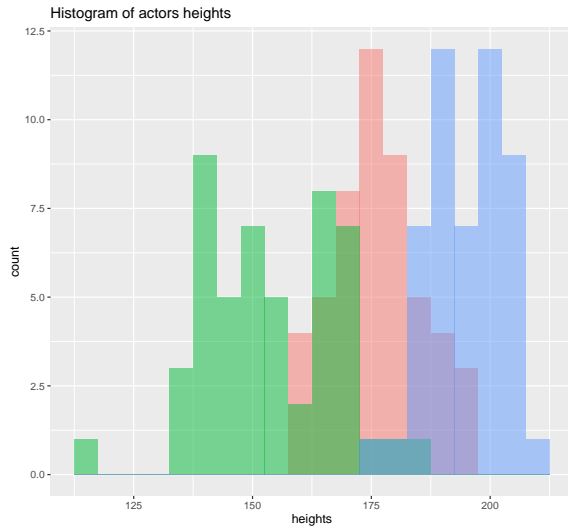
$t$ -test in R using the `t.test()` function (one-sided):

```
t.test(aragorn, gimli, alternative = "greater")

##
##  Welch Two Sample t-test
##
## data:  aragorn and gimli
## t = 9.2512, df = 86.451, p-value = 7.246e-15
## alternative hypothesis: true difference in means is greater than 0
## 95 percent confidence interval:
##  18.43762      Inf
## sample estimates:
## mean of x mean of y
##  176.2119  153.7345
```

# ANOVA

- What if we have more than two groups?
- If we also have 50 actors who auditioned for Legolas



# ANOVA

The  $F$ -statistic is used to test for significance in the split of variance:

$$F = \frac{BSS/(t-1)}{ESS/(n-t-1)} \quad (2)$$

- Ratio of how much of the variance is between the groups to how much is within the groups
- Compare to an  $F$ -distribution, using degrees of freedom based on the number of groups ( $t$ ) and the number of observations ( $n$ )

# ANOVA

- We can use the R function `aov()` to calculate ANOVA for the three groups. Note this uses the model syntax ( $\sim$ )

```
roles = c(rep("Aragorn",50), rep("Gimli",50), rep("Legolas",50))
heights = c(aragorn, gimli, legolas)
summary(aov(heights ~ roles))

##              Df Sum Sq Mean Sq F value Pr(>F)
## roles          2  42767    21383   181.5 <2e-16 ***
## Residuals    147   17319      118
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

# Other inference tests

- $F$ -test: test if difference in ratio of variance of two samples
  - `var.test()`
- Wilcoxon rank sum test: Non-parametric test for the equality of medians
  - `wilcox.test()`
- Correlation tests: tests of *covariation*
  - `cor.test()`
  - Pearson's vs. Spearman's
- Chi-squared ( $\chi^2$ ) tests: tests of *distribution* and *association*
  - `chisq.test()`