

Analysis of Variance

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6 October – Advanced Biological Statistics

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Outline

2 . 1

Goal

To compare means of something between groups.

2 . 2

Goal

To compare means of something between groups.

Related topics:

- When can you do it, and how well? Power, false positive rate.
- How can experiments best do it? Experimental design.
- Methods: two-sample t -test, (one-way) ANOVA, permutation tests

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Comparing means

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Example:

How different are AirBnB prices between neighbourhoods?

```
airbnb <- read.csv("../Datasets/portland-airbnb-listings.csv", stringsAsFactors=TRUE)
airbnb$price <- as.numeric(gsub("$", "", airbnb$price, fixed=TRUE))
airbnb$neighbourhood[airbnb$neighbourhood == ""] <- NA
(neighbourhood_counts <- sort(table(airbnb$neighbourhood), decreasing=TRUE))
```

```
##
##      Richmond Northwest District Concordia Downtown
##      318      238      230      221
##      Mt. Tabor Irvington Sellwood-Moreland Montavilla
##      145      134      133      129
##      Cully South Portland Woodlawn St. Johns
##      87      87      85      81
## Creston-Kenilworth Hillsdale Old Town/Chinatown Beaumont-Wilshire
##      68      59      59      58
## University Park Foster-Powell Laurelhurst Multnomah
##      48      47      45      45
## Portsmouth Alameda Forest Park Homestead
##      37      35      34      34
## Hillside Ashcreek Pleasant Valley Parkrose Heights
##      25      24      21      19
## Lloyd District Markham Argay Collins View
##      11      11      10      10
## Hayden Island Hollywood Maplewood Marshall Park
##      7      7      7      7
## Woodland Park
##      1      0
```

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Let's take only the ten biggest neighbourhoods:

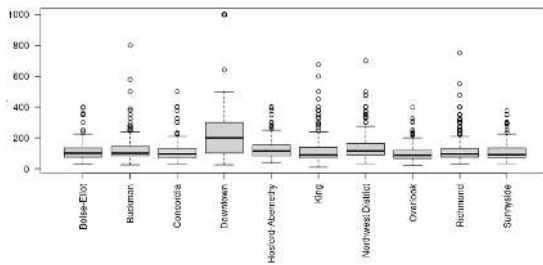
```
big_neighbourhoods <- names(neighbourhood_counts)[1:10]
sub_bnb <- subset(airbnb, !is.na(price) & neighbourhood %in% big_neighbourhoods)
sub_bnb <- droplevels(sub_bnb[, c("price", "neighbourhood", "host_id")])
nrow(sub_bnb)
```

```
## [1] 2023
```

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Look at the data:

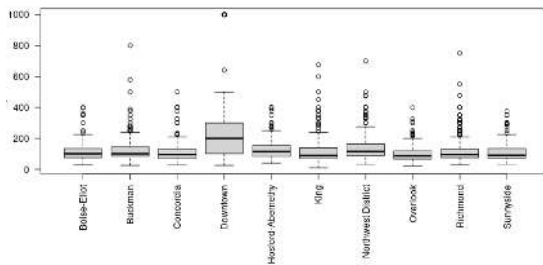
```
par(mar=c(9, 3, 1, 1)+.1)
plot(price ~ neighbourhood, data=sub_bnb, fill=grey(0.8), las=2, xlab='')
```



3 4

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```



Preliminary conclusions? Formal questions?

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ANOVA

4 . 1

The ANOVA model

The *price* P_{ij} of the j th room in neighbourhood i is

$$P_{ij} = \mu + \alpha_i + \epsilon_{ij},$$

where

- μ is the overall mean
- α_i is the mean deviation of neighborhood i from μ
- ϵ_{ij} is what's left over ("error", or "residual")

4 . 2

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In words,

$$(\text{price}) = (\text{group mean}) + (\text{residual})$$

4 . 2

ANOVA

- Stands for ANalysis Of VAriance
- Core statistical procedure in biology
- Developed by R.A. Fisher in the early 20th Century
- Core idea: ask how much variation exists within vs. among groups
- ANOVAs are linear models that have categorical predictor and continuous response variables
- The categorical predictors are often called factors, and can have two or more levels

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Question 1: what are the means?

```
summary(lm(formula = price ~ neighbourhood, data = sub_bnb))
```

```
##
## Call:
## lm(formula = price ~ neighbourhood, data = sub_bnb)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -211.70   -48.12   -23.16    17.28   762.30
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    118.15625     8.13709   14.521  <2e-16 ***
## neighbourhoodBuckman     11.10976     10.88102    1.021  0.3074
## neighbourhoodConcordia    -5.53016     10.59577   -0.522  0.6018
## neighbourhoodDowntown    118.53940     10.83458   10.941  <2e-16 ***
## neighbourhoodHosford-Abernethy  14.56073     11.52553    1.263  0.2066
## neighbourhoodKing         3.14322     11.08430    0.284  0.7768
## neighbourhoodNorthwest District  23.42358     10.52246    2.226  0.0261 *
## neighbourhoodOverlook    -13.53446     11.58999   -1.169  0.2427
## neighbourhoodRichmond     -0.03638      9.98145   -0.004  0.9971
## neighbourhoodSunnyside    -3.90324     11.40300   -0.342  0.7322
## ---
## Signif. codes:  0. '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

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Question 2: is there group heterogeneity?

I.e.: do mean prices differ by neighborhood?

4 . 5

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I.e.: do mean prices differ by neighborhood?

How would you do this?

Design a statistic that would be big if mean prices are different between neighborhoods, and will be small if all neighborhoods are the same.

Question 2, answered by ANOVA

```
anova(lm(formula = price ~ neighbourhood, data = sub_bnb))

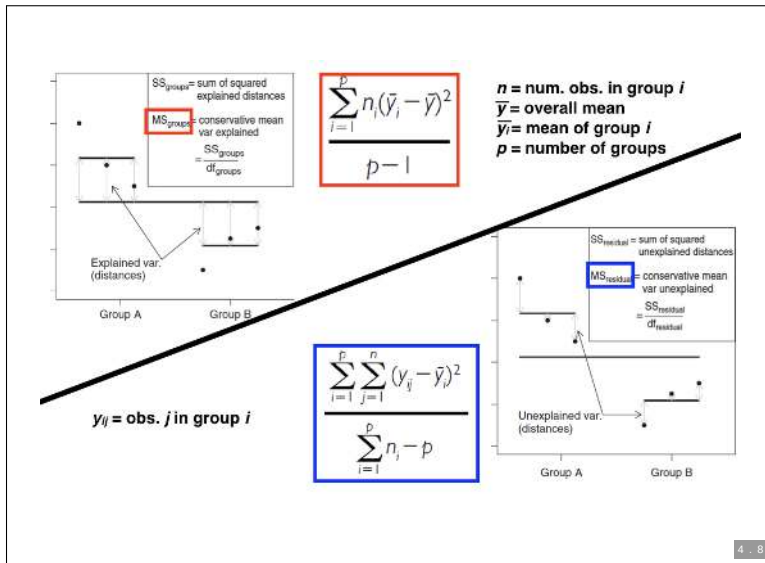
## Analysis of Variance Table
##
## Response: price
##      Df Sum Sq Mean Sq F value    Pr(>F)
## neighbourhood    9  2655967   295107  27.857 < 2.2e-16 ***
## Residuals    2013  21325161   10594
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Table 8.2 ANOVA table for single factor linear model showing partitioning of variation			
Source of	SS	df	MS
Between groups	$\sum_{i=1}^p n_i(\bar{y}_i - \bar{y})^2$	$p - 1$	$\frac{\sum_{i=1}^p n_i(\bar{y}_i - \bar{y})^2}{p - 1}$
Residual	$\sum_{i=1}^p \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_i)^2$	$\sum_{i=1}^p n_i - p$	$\frac{\sum_{i=1}^p \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_i)^2}{\sum_{i=1}^p n_i - p}$
Total	$\sum_{i=1}^p \sum_{j=1}^{n_i} (y_{ij} - \bar{y})^2$	$\sum_{i=1}^p n_i - 1$	

Var. explained by groupings

Var. unexplained by groupings

F-ratio = $\frac{MS_{groups}}{MS_{residuals}}$



One or more predictor variables

- One-way ANOVAs just have a single factor
- Multi-factor ANOVAs
 - Factorial - two or more factors and their interactions
 - Nested - the levels of one factor are contained within another level
 - The models can be quite complex
- ANOVAs use an F -statistic to test factors in a model
 - Ratio of two variances (numerator and denominator)
 - The numerator and denominator d.f. need to be included (e.g. $F_{1,34} = 29.43$)
- Determining the appropriate test ratios for complex ANOVAs takes some work

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Assumptions

- Normally distributed groups
 - robust to non-normality if equal variances and sample sizes
- Equal variances across groups
 - okay if largest-to-smallest variance ratio < 3:1
 - problematic if there is a mean-variance relationship among groups
- Observations in a group are independent
 - randomly selected
 - don't confound group with another factor

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