

ETC5521: Exploratory Data Analysis

Initial data analysis

Lecturer: Emi Tanaka

ETC5521.Clayton-x@monash.edu

Week 3 - Session 2



Linear models in R REVIEW Part 1/3

```
library(tidyverse)
glimpse(cars)

## Rows: 50

## Columns: 2

## $ speed <dbl> 4, 4, 7, 7, 8, 9, 10, 10, 10, 11, 11, 12, 12, 12, 12, 13, 13, 13, 13, 14, 14, 1

## $ dist <dbl> 2, 10, 4, 22, 16, 10, 18, 26, 34, 17, 28, 14, 20, 24, 28, 26, 34, 34, 46, 26, 3

ggplot(cars, aes(speed, dist)) +
    geom_point() +
    geom_smooth(method = "lm", se = FALSE)
```

Linear models in R REVIEW Part 2/3

• We can fit linear models in R with the 1m function:

```
lm(dist \sim speed, data = cars)
is the same as
lm(dist \sim 1 + speed, data = cars)
```

• The above model is mathematically written as

$$y_i = \beta_0 + \beta_1 x_i + e_i$$

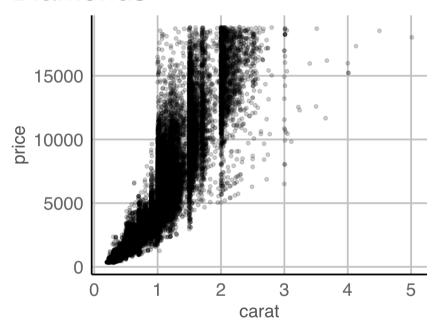
where

- yand xare the stopping distance (in ft) and speed (in mph), respectively, of the ith car;
- And Are intercept and slope, respectively; and
- \dot{a} s the random error; usually assuming $e_i \sim NID(0,\sigma^2)$

2 Model formulation Part 1/2

• Say, we are interested in characterising the price of the diamond in terms of its carat.

Diamonds

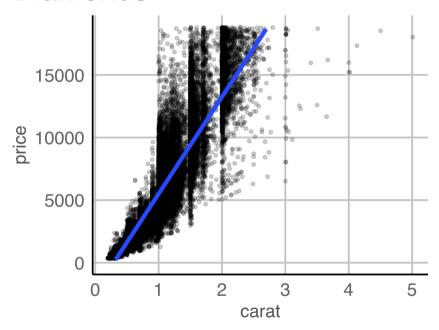


· Looking at this plot, would you fit a linear model with formula

2 Model formulation Part 1/2

• Say, we are interested in characterising the price of the diamond in terms of its carat.

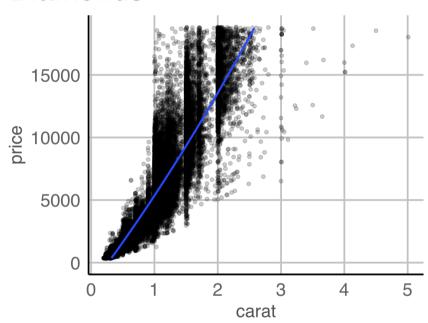
Diamonds



· Looking at this plot, would you fit a linear model with formula

2 Model formulation Part 2/2

Diamonds



What about

$$y_i = \beta_0 + \beta_1 x_i + \beta_2 x_i^2 + e_i$$

- Should the assumption for error distribution be modified if so?
- Should we make some transformation before modelling?
- Are there other candidate models?

2 Model formulation Part 2/2

- Notice that we did no formal statistical inference as we initially try to formulate the model.
- The goal of the main analysis is to characterise the price of a diamond by its carat. This may involve:
 - formal inference for model selection;
 - justification of the selected "final" model; and
 - fitting the final model.
- There may be in fact many, many models considered but discarded at the IDA stage.
- These discarded models are hardly ever reported. Consequently, majority of reported statistics give a distorted view and it's important to remind yourself what might **not** be reported.

Model selection



All models are approximate and tentative; approximate in the sense that no model is exactly true and tentative in that they may be modified in the light of further data

—*Chatfield (1985)*



All models are wrong but some are useful

—George Box

Case study 4 Wheat yield in South Australia Part 1/9

A wheat breeding trial to test 107 varieties (also called genotype) is conducted in a field experiment laid out in a rectangular array with 22 rows and 15 columns.

```
data("gilmour.serpentine", package = "agridat")
skimr::skim(gilmour.serpentine)
## — Data Summary -
##
                             Values
## Name
                             gilmour.serpentine
## Number of rows
                             330
## Number of columns
## Column type frequency:
   factor
   numeric
## Group variables
                             None
##
```

Case study 4 Wheat yield in South Australia Part 2/9

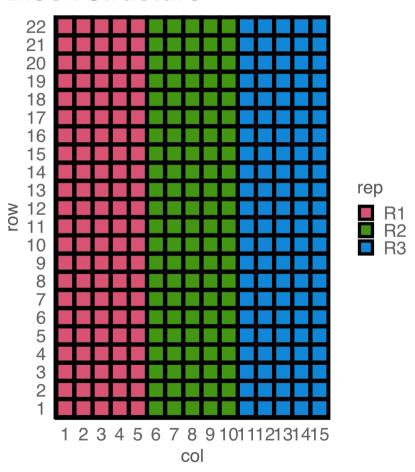
Experimental Design

- The experiment employs what is referred to as a **randomised complete block design** (RCBD) (technically it is *near*-complete and not exactly RCBD due to check varieties have double the replicates of test varieties).
- RCBD means that
 - the there are equal number of replicates for each treatment (here it is gen);
 - each treatment appears exactly once in each block;
 - the blocks are of the same size; and
 - each treatment are randomised within block.
- In agricultural field experiments, blocks are formed spatially by grouping plots within contiguous areas (called rep here).
- The boundaries of blocks may be chosen arbitrary.

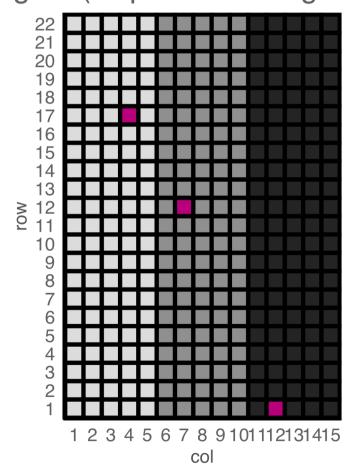
Case study 4 Wheat yield in South Australia Part 3/9

Experimental Design

Block structure



gen: (WqKPWmH*3Ag



Case study 4 Wheat yield in South Australia Part 4/9

Analysis

- In the main analysis, people would commonly analyse this using what is called **two-way ANOVA** model (with no interaction effect).
- The two-way ANOVA model has the form

```
yield = mean + block + treatment + error
```

So for this data,

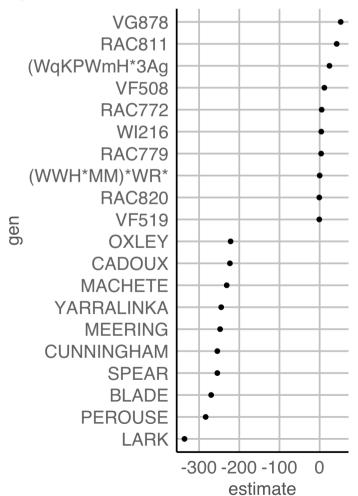
Case study 4 Wheat yield in South Australia Part 5/9

Analysis

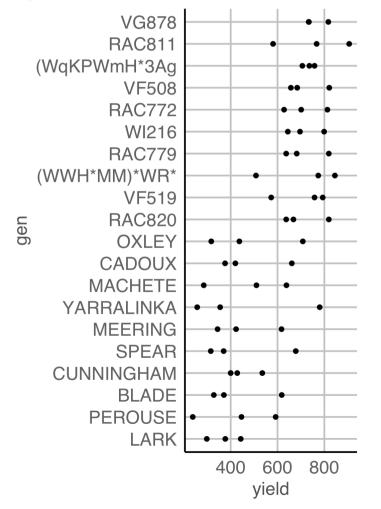
```
summary(fit)
##
## Call:
## lm(formula = yield \sim 1 + rep + gen, data = gilmour.serpentine)
##
## Residuals:
      Min 1Q Median 3Q
##
                                    Max
## -245.070 -69.695 -1.182 71.427 250.652
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 720.248
                           67.335 10.697 < 2e-16 ***
         96.100 15.585 6.166 3.29e-09 ***
## repR2
## repR3 -129.845 15.585 -8.331 8.44e-15 ***
## gen(WqKPWmH*3Ag 24.333 94.372 0.258 0.796766
## genAMERY -93.333 94.372 -0.989 0.323747
```

Case study 4 Wheat yield in South Australia Part 6/9

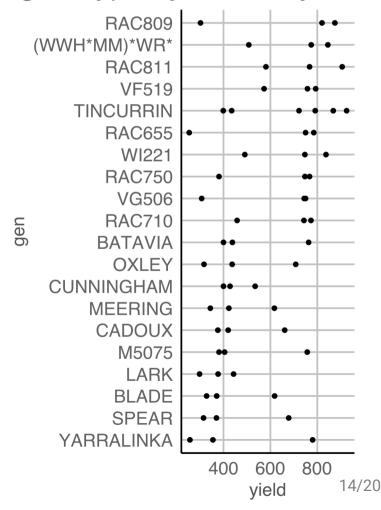
Top 10 and bottom 10 genotype by model est.



Top 10 and bottom 10 genotype by mean yield



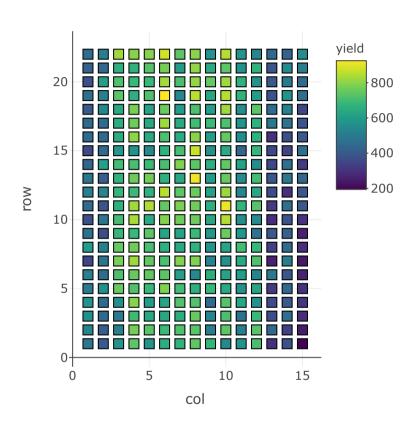
Top 10 and bottom 10 genotype by median yield

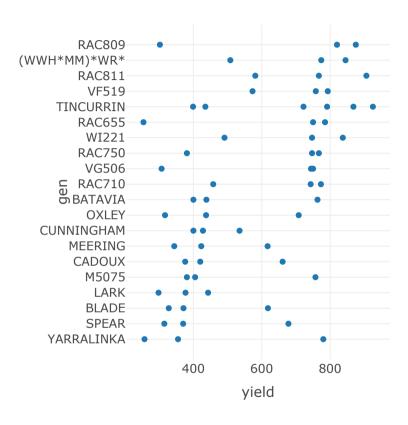


Case study 4 Wheat yield in South Australia Part 7/9

Do you notice anything from below?

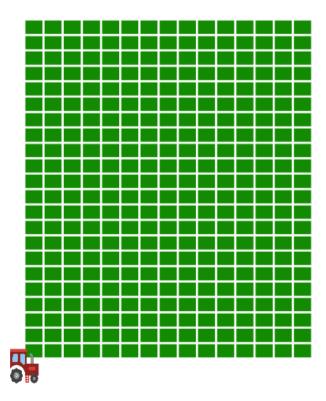
Case study 4 Wheat yield in South Australia Part 8/9

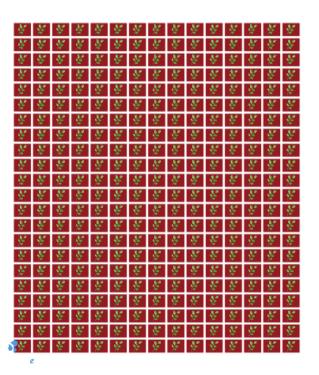




Case study 4 Wheat yield in South Australia Part 9/9

- It's well known in agricultural field trials that spatial variations are introduced in traits; this could be because of the fertility trend, management practices or other reasons.
- In the IDA stage, you investigate to identify these spatial variations you cannot just simply fit a two-way ANOVA model!





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"Teaching of Statistics should provide a more balanced blend of IDA and inference"

Chatfield (1985)

Yet there is still very little emphasis of it in teaching and also at times in practice.

So don't forget to do IDA!

Take away messages

- ✓ Initial data analysis (IDA) is a modelfocussed exploration of data with two main objectives:
 - data description including scrutinizing for data quality, and
 - model formulation without any formal statistical inference.
- ✓ IDA hardly sees the limelight even if it's the very foundation of what the main analysis is built on.





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