Software for transport data science

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Project set-up and tidyverse testing (30 minutes, individually)

- Check your packages are up-to-date with update.packages()
- Create an RStudio project with an appropriate name for this module (e.g. TDS)
- Create appropriate files for code, data and anything else (e.g. images)
- Create a script called learning-tidyverse.R, e.g. with **one** the following commands:

```
file.edit(learning-tidyverse.R) # or
file.edit(code/learning-tidyverse.R)
```

• Read section 5.1 (https://r4ds.had.co.nz/transform.html#filter-rows-with-filter) of R for Data Science and write code that reproduces the results in that section in the script learning-tidyverse.R

Your script will start with something like this:

```
library(tidyverse)
```

library(nycflights13)

Reading-in and processing coffee data

Read-in the coffee data we created last week, e.g. with:

```
u = paste0(
  "https://github.com/ITSLeeds/TDS/",
  "raw/master/sample-data/everyone.csv"
  )
d = read_csv(u)
```

```
## Parsed with column specification:
## cols(
## person_name = col_character(),
## n_coffee = col_double(),
## like_bus_travel = col_logical()
## )
```

Create a new variable called 'n coffee yr' with the following command:

```
d$n_coffee_yr = d$n_coffee * 52
```

Find the mean number of cups of coffee people drink per year (and the total)

Note: the same result can be acheived as follows:

```
d = mutate(d, n_coffee_yr = n_coffee * 52)

# or
d = d %>%
  mutate(n_coffee_yr = n_coffee * 52)
```

- Which do you prefer?
- · Filter-out only those who travel by bus
- Bonus: Create a new dataset that keeps only the person_name and n_coffee_yr variables (hint: use the select() function)
- Bonus: do those who travel by bus drink more or less coffee than those who do not?

Processing a big file and basic visualisation (30 minutes, individually)

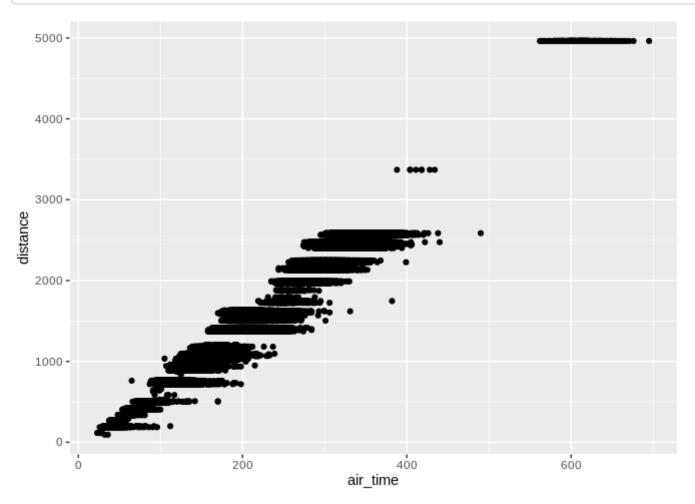
• Take a random sample of 10,000 flights and assign it to an object with the following line of code:

```
flights_sample = sample_n(flights, 1e4)
```

- Find the unique carriers with the unique() function
- Create an object containing flights from United, American, or Delta, and assign it to f, as follows:

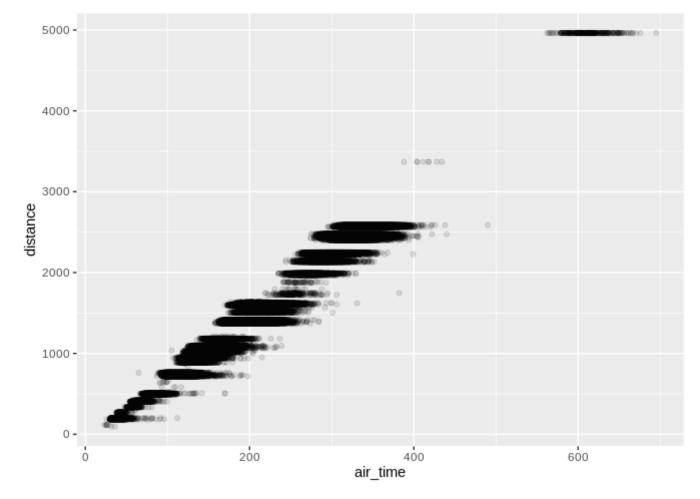
```
f = filter(flights, grepl(pattern = "UA|AA|DL", x = carrier))
```

• Create plots that visualise the sample flights, using code from Chapter 3 of the same book, starting with the following plot:



• Add transparency so it looks like this (hint: use alpha = in the geom_point() function call):

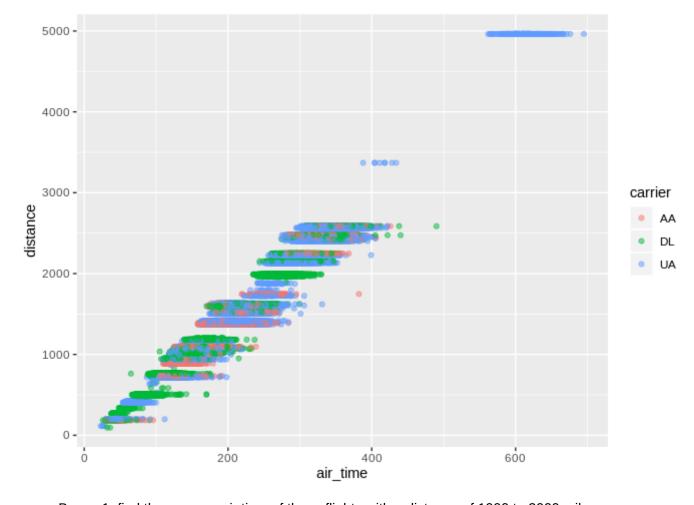
Warning: Removed 2117 rows containing missing values (geom_point).



• Add a colour for each carrier, so it looks something like this:

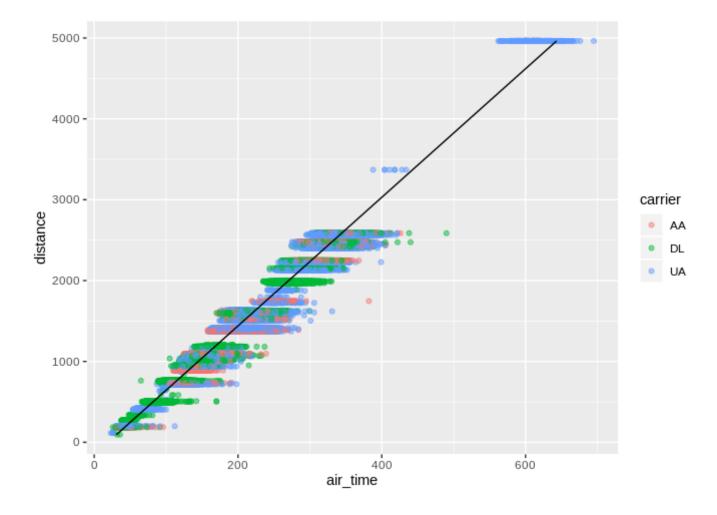
```
ggplot(f) +
  geom_point(aes(air_time, distance, colour = carrier), alpha = 0.5)
```

Warning: Removed 2117 rows containing missing values (geom_point).



- Bonus 1: find the average air time of those flights with a distance of 1000 to 2000 miles
- Bonus 2: use the lm() function to find the relationship between flight distance and time, and plot the results (start the plot as follows, why did we use na.omit()? hint find help with ?na.omit()):

```
f = na.omit(f)
m = lm(air_time ~ distance, data = f)
f$pred = m$fitted.values
```



Homework

- 1. create a reproducible document
- Create an Rmarkdown file with the following command:

file.edit("learning-tidyverse.Rmd")

• Take a read of the guidance on RMarkdown files online and in the following location (or search online for the 'RMarkdown cheatsheet'):

Help > Cheatsheets > RMarkdown

- Put the code you generated for tidyverse.R into the Rmd file and knit it
- Bonus: create a GitHub repo and publish the results of of your work (hint: putting output: github_document may help here!)
- 2. Work-through the remaining exercises of the first sections in R4DS chapters 3 and 5
- Write and R script, with comments, to show your working (and prove you've done it!)
- 3. Create an RMarkdown file containing reproducible code outlining what you learned today
- 4. Identify a dataset you would like to work with for the practical next week.