Modern R with tidyverse [Solutions]

Jonathan de Bruin & Barbara Vreede

Contents

This document is part of the workshop ***Introduction to R & Data*** *by Utrecht University RDM Support.*

```
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 3.4.2
```

```
## -- Attaching packages --------------------------------- tidyverse 1.2.1 --
```

```
## ✓ ggplot2 2.2.1     ✓ purrr   0.2.5
## ✓ tibble  1.4.2     ✓ dplyr   0.7.4
## ✓ tidyr   0.8.1     ✓ stringr 1.2.0
## ✓ readr   1.1.1     ✓ forcats 0.3.0
```

```
## Warning: package 'tibble' was built under R version 3.4.3
```

```
## Warning: package 'tidyr' was built under R version 3.4.4
```

```
## Warning: package 'purrr' was built under R version 3.4.4
```

```
## Warning: package 'dplyr' was built under R version 3.4.2
```

```
## Warning: package 'forcats' was built under R version 3.4.3
```

```
## -- Conflicts ------------------------------------ tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

1. Read and save data

Basic exercise I - Read data into R

a) Find the delimiter of the dataset `Huji_JNF_Crane_israel_GPRS.csv`

Use the data import tool of RStudio or use a text editor to inspect the document. The delimiter is `,`.

b) Read the data `Huji_JNF_Crane_israel_GPRS.csv` into R

```
data_crane <- read_csv('data/Huji_JNF_Crane_israel_GPRS.csv')
```

```
## Parsed with column specification:
## cols(
##   .default = col_double(),
##   visible = col_character(),
##   timestamp = col_datetime(format = ""),
##   battery_charge_percent = col_integer(),
```

```
##   eobs_activity = col_character(),
##   eobs_activity_samples = col_character(),
##   eobs_battery_voltage = col_integer(),
##   eobs_fix_battery_voltage = col_integer(),
##   eobs_start_timestamp = col_character(),
##   eobs_status = col_character(),
##   eobs_temperature = col_integer(),
##   eobs_type_of_fix = col_integer(),
##   eobs_used_time_to_get_fix = col_integer(),
##   gps_fix_type = col_character(),
##   gps_satellite_count = col_integer(),
##   import_marked_outlier = col_character(),
##   orn_transmission_protocol = col_character(),
##   sensor_type = col_character(),
##   individual_taxon_canonical_name = col_character(),
##   individual_local_identifier = col_character(),
##   study_name = col_character()
## )
```

```
## See spec(...) for full column specifications.
```

```
head(data_crane)
```

```
## # A tibble: 6 x 43
##   event_id visible timestamp           location_long location_lat
##      <dbl> <chr>   <dttm>                      <dbl>        <dbl>
## 1   3.79e9 true    2017-10-15 00:00:16          40.3         55.3
## 2   3.80e9 true    2017-10-15 00:00:16          40.3         55.3
## 3   3.79e9 true    2017-10-15 02:00:23          40.3         55.3
## 4   3.80e9 true    2017-10-15 02:00:23          40.3         55.3
## 5   3.79e9 true    2017-10-15 04:00:23          40.3         55.3
## 6   3.80e9 true    2017-10-15 04:00:23          40.3         55.3
## # ... with 38 more variables: acceleration_raw_x <dbl>,
## #   acceleration_raw_y <dbl>, acceleration_raw_z <dbl>,
## #   bar_barometric_height <dbl>, battery_charge_percent <int>,
## #   battery_charging_current <dbl>, eobs_activity <chr>,
## #   eobs_activity_samples <chr>, eobs_battery_voltage <int>,
## #   eobs_fix_battery_voltage <int>,
## #   eobs_horizontal_accuracy_estimate <dbl>, eobs_key_bin_checksum <dbl>,
## #   eobs_speed_accuracy_estimate <dbl>, eobs_start_timestamp <chr>,
## #   eobs_status <chr>, eobs_temperature <int>, eobs_type_of_fix <int>,
## #   eobs_used_time_to_get_fix <int>, external_temperature <dbl>,
## #   gps_fix_type <chr>, gps_hdop <dbl>, gps_satellite_count <int>,
## #   gps_time_to_fix <dbl>, ground_speed <dbl>, heading <dbl>,
## #   height_above_ellipsoid <dbl>, height_above_msl <dbl>,
## #   import_marked_outlier <chr>, gls_light_level <dbl>,
## #   mag_magnetic_field_raw_x <dbl>, mag_magnetic_field_raw_y <dbl>,
## #   mag_magnetic_field_raw_z <dbl>, orn_transmission_protocol <chr>,
## #   tag_voltage <dbl>, sensor_type <chr>,
## #   individual_taxon_canonical_name <chr>,
## #   individual_local_identifier <chr>, study_name <chr>
```

c) Load `readxl` (to read Excel files)

```
library(readxl)
```

```
## Warning: package 'readxl' was built under R version 3.4.4
```

```
# open the help function
?read_excel
```

c) Read the additional observations `Huji_JNF_additional_observations.xlsx`.

```
data_crane_additional <- read_excel('data/Huji_JNF_additional_observations.xlsx')

head(data_crane_additional)
```

```
## # A tibble: 6 x 4
##   `event-id` meas_1 meas_2 meas_3
##        <dbl>  <dbl>  <dbl>  <dbl>
## 1 3794510958 0.147  0.832   0.751
## 2 3796034354 0.871  1.23    0.388
## 3 3794510962 0.507  0.0755  1.77
## 4 3796034355 0.654  1.43    1.20
## 5 3794510961 0.0520 0.606   0.307
## 6 3796034356 2.20   1.11    1.90
```

Basic exercise II - Dataset properties

```
glimpse(data_crane)
```

```
## Observations: 20,873
## Variables: 43
## $ event_id                           <dbl> 3794510958, 3796034354, 3794...
## $ visible                            <chr> "true", "true", "true", "tru...
## $ timestamp                          <dttm> 2017-10-15 00:00:16, 2017-1...
## $ location_long                      <dbl> 40.31509, 40.31509, 40.31481...
## $ location_lat                       <dbl> 55.34764, 55.34764, 55.34792...
## $ acceleration_raw_x                 <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ acceleration_raw_y                 <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ acceleration_raw_z                 <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ bar_barometric_height              <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ battery_charge_percent             <int> NA, NA, NA, NA, NA, NA, NA, ...
## $ battery_charging_current           <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ eobs_activity                      <chr> NA, NA, NA, NA, NA, NA, NA, ...
## $ eobs_activity_samples              <chr> NA, NA, NA, NA, NA, NA, NA, ...
## $ eobs_battery_voltage               <int> 3784, 3818, 3784, 3818, 3784...
## $ eobs_fix_battery_voltage           <int> NA, 3801, NA, 3798, NA, 3796...
## $ eobs_horizontal_accuracy_estimate <dbl> NA, 48.90, NA, 2.56, NA, 3.8...
## $ eobs_key_bin_checksum              <dbl> 0, 2787211109, 0, 1067160123...
## $ eobs_speed_accuracy_estimate       <dbl> NA, 1.54, NA, 0.33, NA, 0.34...
## $ eobs_start_timestamp               <chr> NA, "2017,15+Oct", NA, "2017...
## $ eobs_status                        <chr> NA, "A", NA, "A", NA, "A", N...
## $ eobs_temperature                   <int> NA, -7, NA, -7, NA, -8, NA, ...
## $ eobs_type_of_fix                   <int> NA, 3, NA, 3, NA, 3, NA, 3, ...
## $ eobs_used_time_to_get_fix          <int> NA, 15, NA, 22, NA, 21, NA, ...
## $ external_temperature               <dbl> NA, NA, NA, NA, NA, NA, NA, ...
```

```
## $ gps_fix_type                      <chr> NA, NA, NA, NA, NA, NA, NA, ...
## $ gps_hdop                          <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ gps_satellite_count               <int> NA, NA, NA, NA, NA, NA, NA, ...
## $ gps_time_to_fix                   <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ ground_speed                      <dbl> NA, 0.34, NA, 0.01, NA, 12.8...
## $ heading                           <dbl> NA, 340.13, NA, 0.00, NA, 22...
## $ height_above_ellipsoid            <dbl> NA, 127.0, NA, 112.2, NA, 20...
## $ height_above_msl                  <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ import_marked_outlier             <chr> NA, NA, NA, NA, NA, NA, NA, ...
## $ gls_light_level                   <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ mag_magnetic_field_raw_x          <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ mag_magnetic_field_raw_y          <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ mag_magnetic_field_raw_z          <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ orn_transmission_protocol         <chr> NA, NA, NA, NA, NA, NA, NA, ...
## $ tag_voltage                       <dbl> NA, NA, NA, NA, NA, NA, NA, ...
## $ sensor_type                       <chr> "gps", "gps", "gps", "gps", ...
## $ individual_taxon_canonical_name   <chr> "Grus grus", "Grus grus", "G...
## $ individual_local_identifier       <chr> "L6037", "L6037", "L6037", "...
## $ study_name                        <chr> "Huji JNF Crane israel GPRS"...
```

```
glimpse(data_crane_additional)
```

```
## Observations: 20,873
## Variables: 4
## $ `event-id` <dbl> 3794510958, 3796034354, 3794510962, 3796034355, 379...
## $ meas_1     <dbl> 0.14658647, 0.87147653, 0.50707372, 0.65440285, 0.0...
## $ meas_2     <dbl> 0.83185023, 1.22945914, 0.07549742, 1.43160744, 0.6...
## $ meas_3     <dbl> 0.75099164, 0.38835874, 1.77393265, 1.19853341, 0.3...
```

Reading exercise - `readr` versus base R

Optional exercise (+) - Save data to a CSV file with delimiter ;.

```
# create a directory for the output file
if (!dir.exists('tmp')){
  dir.create("tmp")
}

write_delim(data_crane, 'tmp/data_crane_csv_file.csv', delim = ';')
```

Optional exercise (++) - Read SPSS, SAS, Excel and STATA data files

```
library("haven") # to read and write SPSS, STATA and SAS files
```

```
## Warning: package 'haven' was built under R version 3.4.4
```

a) Write tibble `data_crane` to SPSS, SAS, STATA data files.

```
# create a directory
if (!dir.exists('tmp')){
  dir.create("tmp")
}
```

```r
# read and write files
write_sav(data_crane, file.path("tmp", "crane_spss.sav"))
read_sav(file.path("tmp", "crane_spss.sav"))
```

```
## # A tibble: 20,873 x 43
##    event_id visible timestamp           location_long location_lat
##       <dbl> <chr>   <dttm>                      <dbl>        <dbl>
##  1   3.79e9 true    2017-10-15 00:00:16          40.3         55.3
##  2   3.80e9 true    2017-10-15 00:00:16          40.3         55.3
##  3   3.79e9 true    2017-10-15 02:00:23          40.3         55.3
##  4   3.80e9 true    2017-10-15 02:00:23          40.3         55.3
##  5   3.79e9 true    2017-10-15 04:00:23          40.3         55.3
##  6   3.80e9 true    2017-10-15 04:00:23          40.3         55.3
##  7   3.79e9 true    2017-10-15 05:00:14          40.1         55.3
##  8   3.80e9 true    2017-10-15 05:00:14          40.1         55.3
##  9   3.80e9 true    2017-10-15 06:00:23          40.1         55.3
## 10   3.80e9 true    2017-10-15 08:00:18          40.1         55.3
## # ... with 20,863 more rows, and 38 more variables:
## #   acceleration_raw_x <dbl>, acceleration_raw_y <dbl>,
## #   acceleration_raw_z <dbl>, bar_barometric_height <dbl>,
## #   battery_charge_percent <dbl>, battery_charging_current <dbl>,
## #   eobs_activity <chr>, eobs_activity_samples <chr>,
## #   eobs_battery_voltage <dbl>, eobs_fix_battery_voltage <dbl>,
## #   eobs_horizontal_accuracy_estimate <dbl>, eobs_key_bin_checksum <dbl>,
## #   eobs_speed_accuracy_estimate <dbl>, eobs_start_timestamp <chr>,
## #   eobs_status <chr>, eobs_temperature <dbl>, eobs_type_of_fix <dbl>,
## #   eobs_used_time_to_get_fix <dbl>, external_temperature <dbl>,
## #   gps_fix_type <chr>, gps_hdop <dbl>, gps_satellite_count <dbl>,
## #   gps_time_to_fix <dbl>, ground_speed <dbl>, heading <dbl>,
## #   height_above_ellipsoid <dbl>, height_above_msl <dbl>,
## #   import_marked_outlier <chr>, gls_light_level <dbl>,
## #   mag_magnetic_field_raw_x <dbl>, mag_magnetic_field_raw_y <dbl>,
## #   mag_magnetic_field_raw_z <dbl>, orn_transmission_protocol <chr>,
## #   tag_voltage <dbl>, sensor_type <chr>,
## #   individual_taxon_canonical_name <chr>,
## #   individual_local_identifier <chr>, study_name <chr>
```

```r
# write_sas(data_crane, file.path("tmp", "crane_sas.sas7bdat"))
# read_sas(file.path("tmp", "crane_sas.sas7bdat"))

# write_dta(data_crane, file.path("tmp", "crane_stata.dta"))
# read_dta(file.path("tmp", "crane_stata.dta"))
```

b) Write tibble `data_crane` to Excel.

This is not possible with `tidyverse` at the moment. `readxl` only support Excel file reading. This is not a problem for a researcher, because we don't use Excel, do we?

Optional exercise (+++) - Parse datetime columns

No solutions available at the moment.

2. Data visualisation

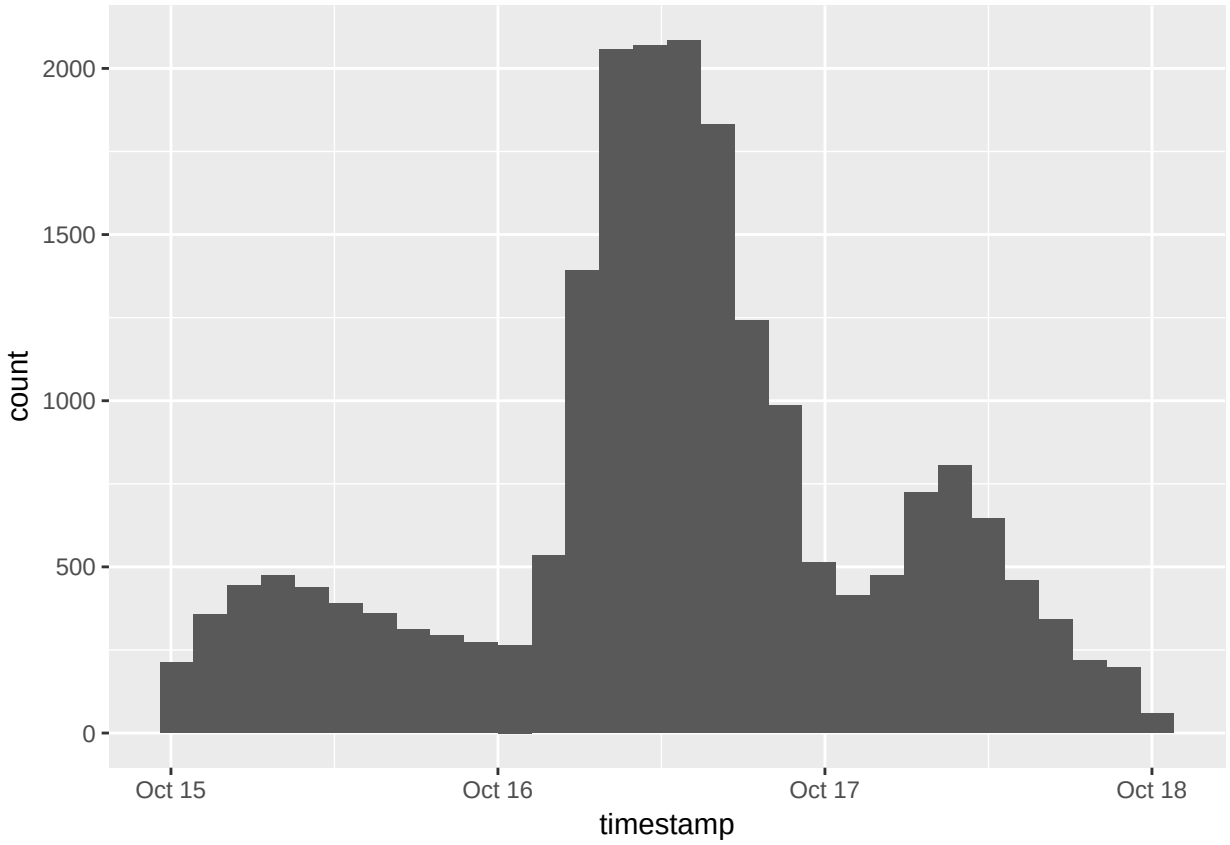
Basic exercise I - Quick plots of the `data_crane`.

a) Single column plots

Make a quick plot of at least 5 columns in the `data_crane` dataset.

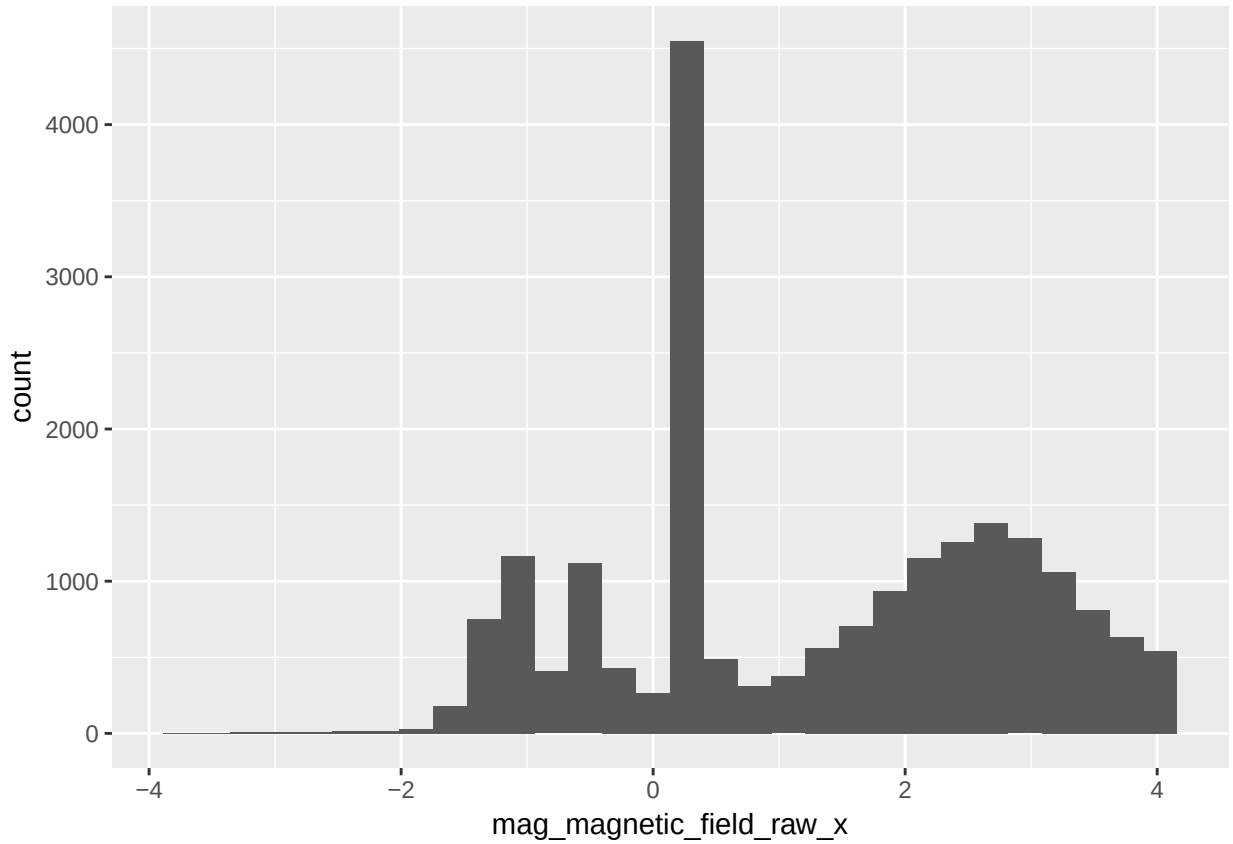
```
qplot(timestamp, data=data_crane)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



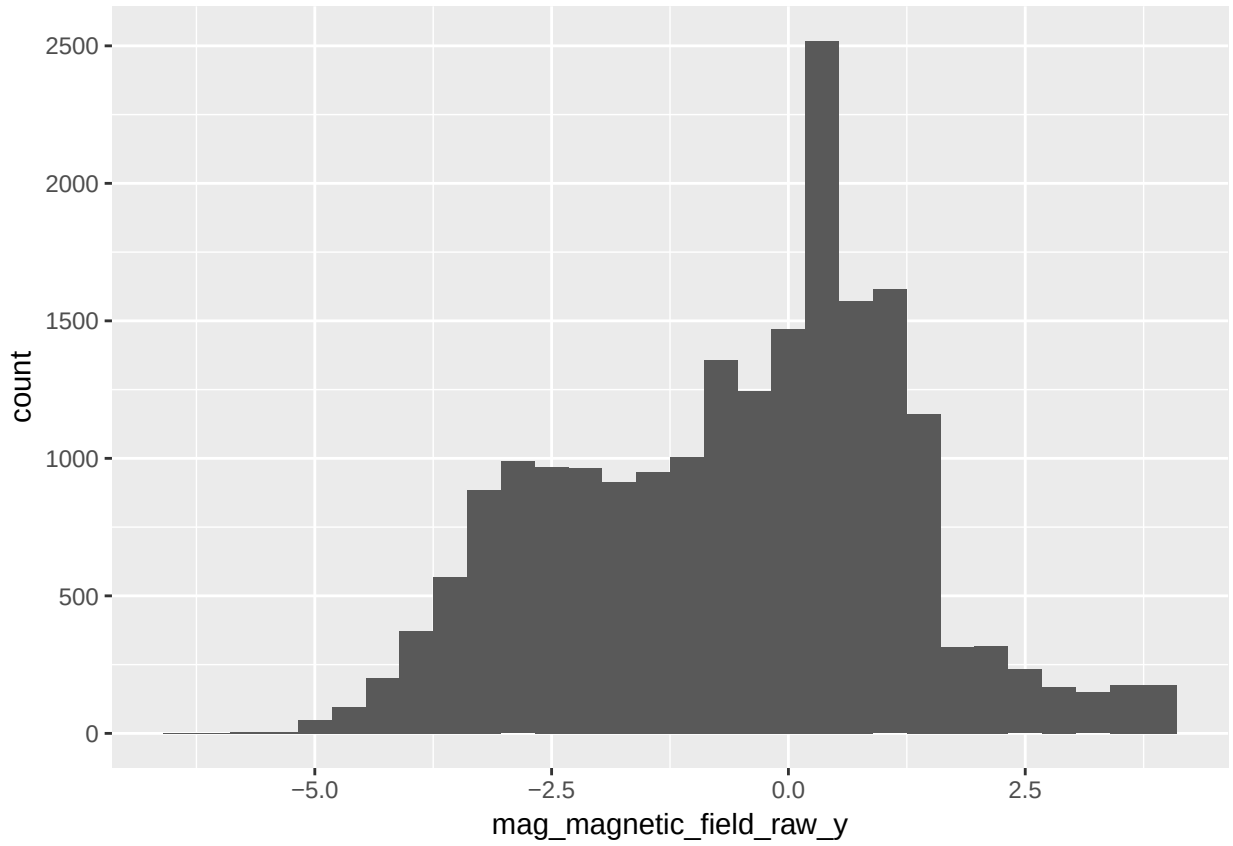

```
qplot(mag_magnetic_field_raw_x, data=data_crane)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## Warning: Removed 432 rows containing non-finite values (stat_bin).
```



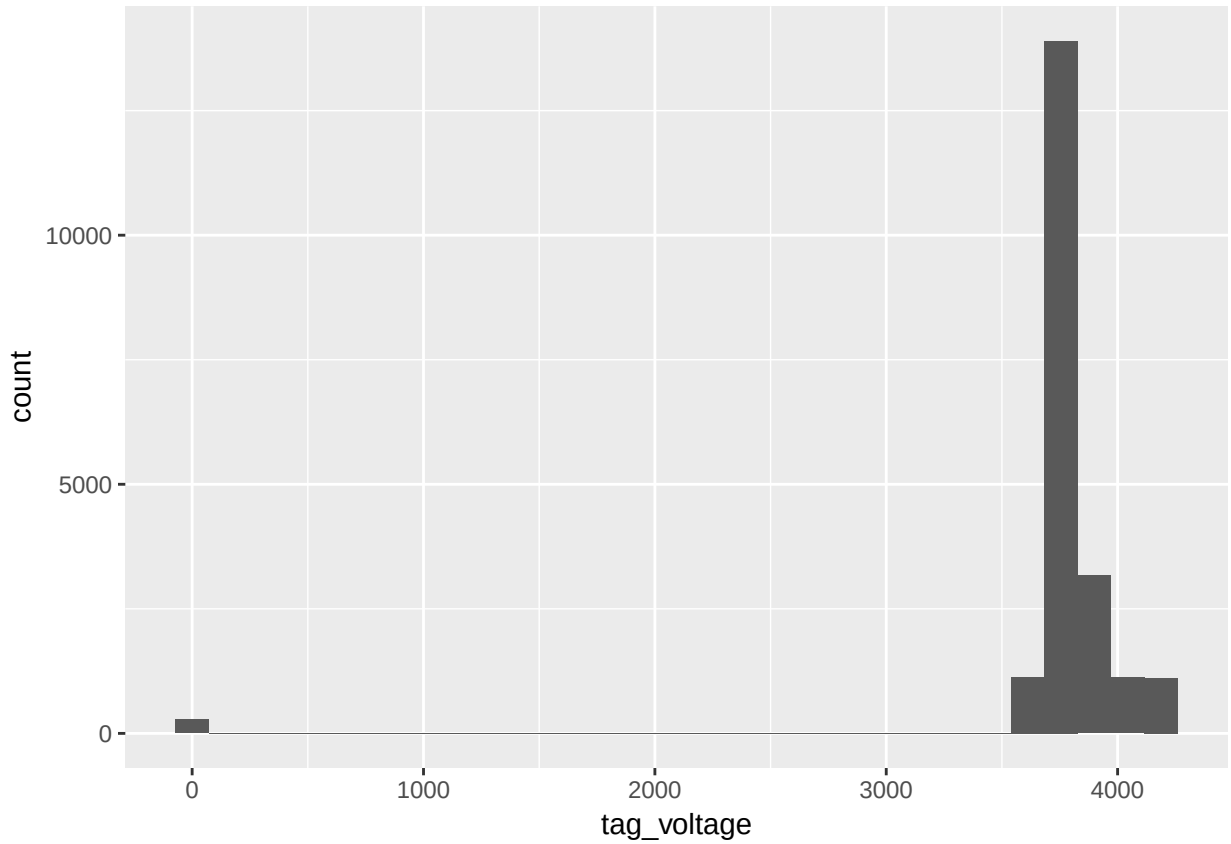

```
qplot(mag_magnetic_field_raw_y, data=data_crane)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## Warning: Removed 432 rows containing non-finite values (stat_bin).
```

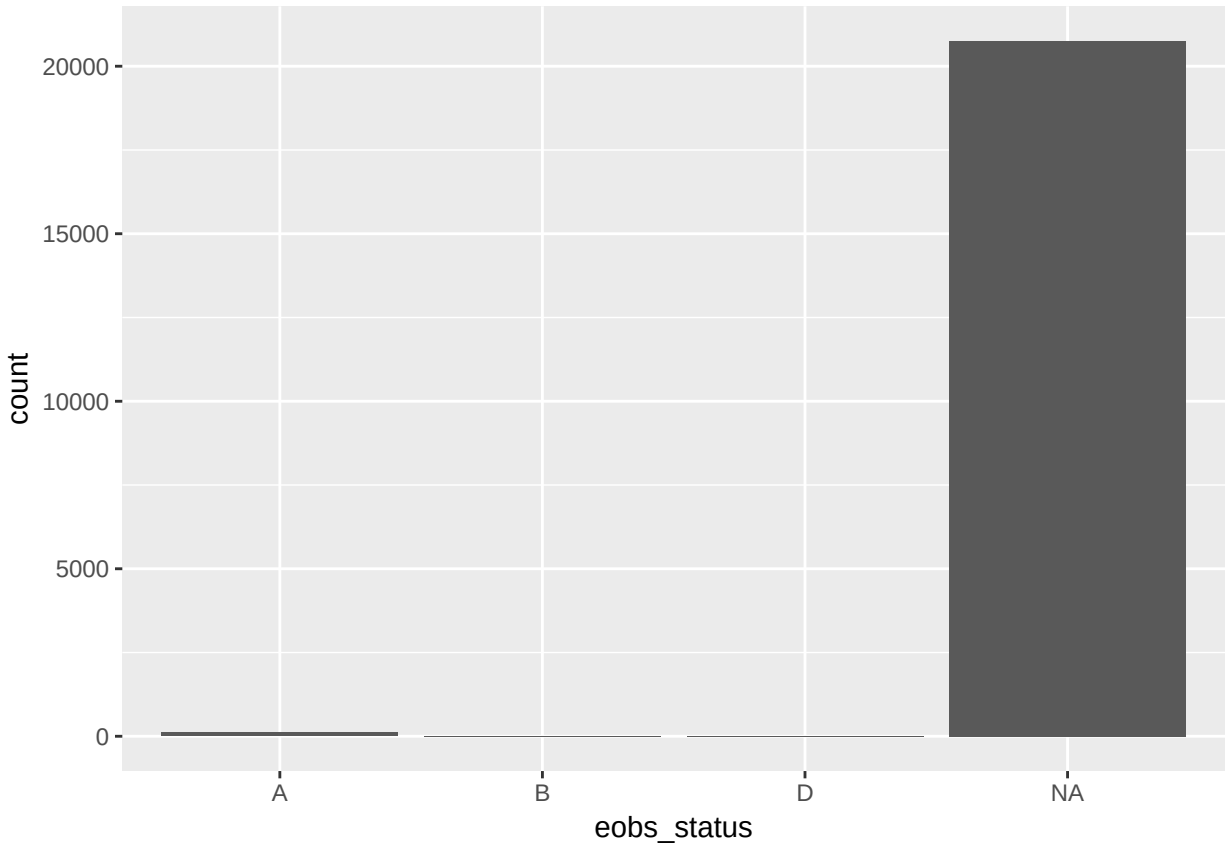



```
qplot(tag_voltage, data=data_crane)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## Warning: Removed 155 rows containing non-finite values (stat_bin).
```



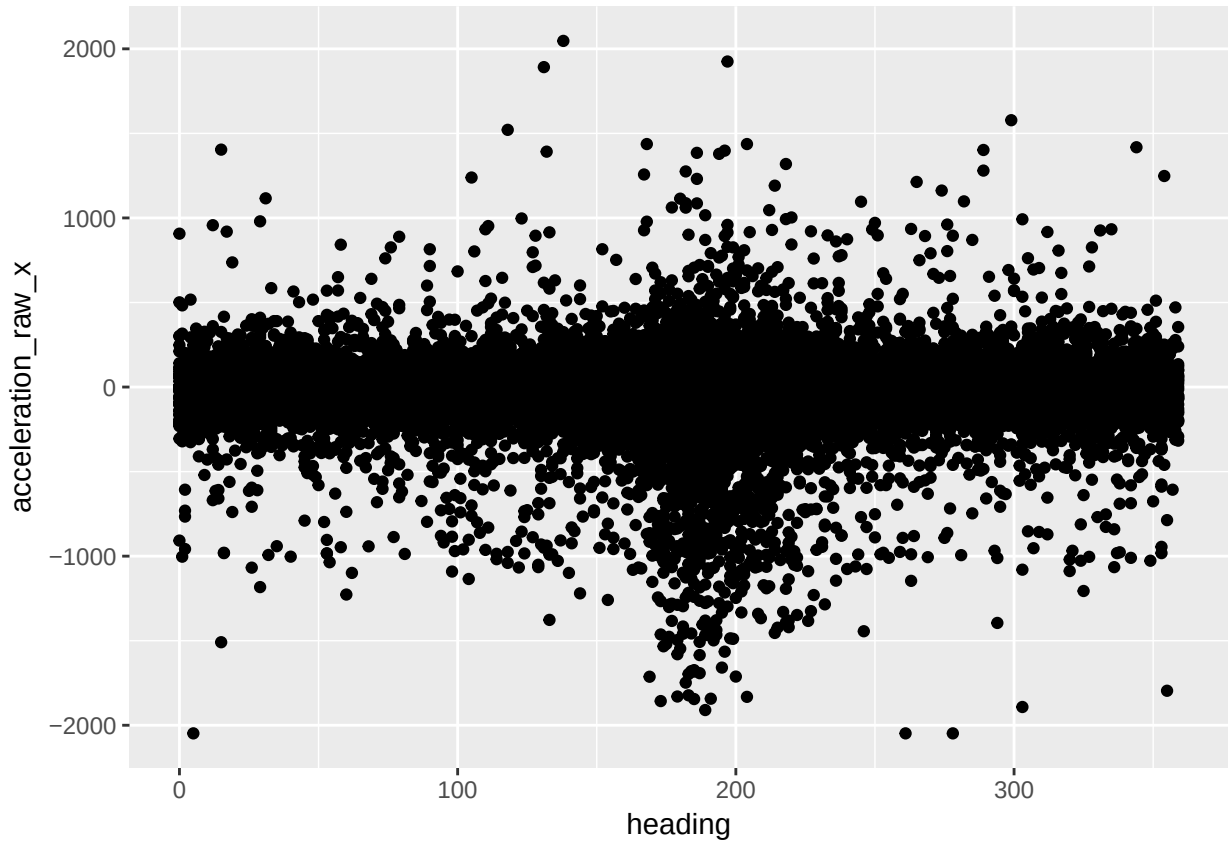

```
qplot(eobs_status, data=data_crane)
```



b) Two column in plots

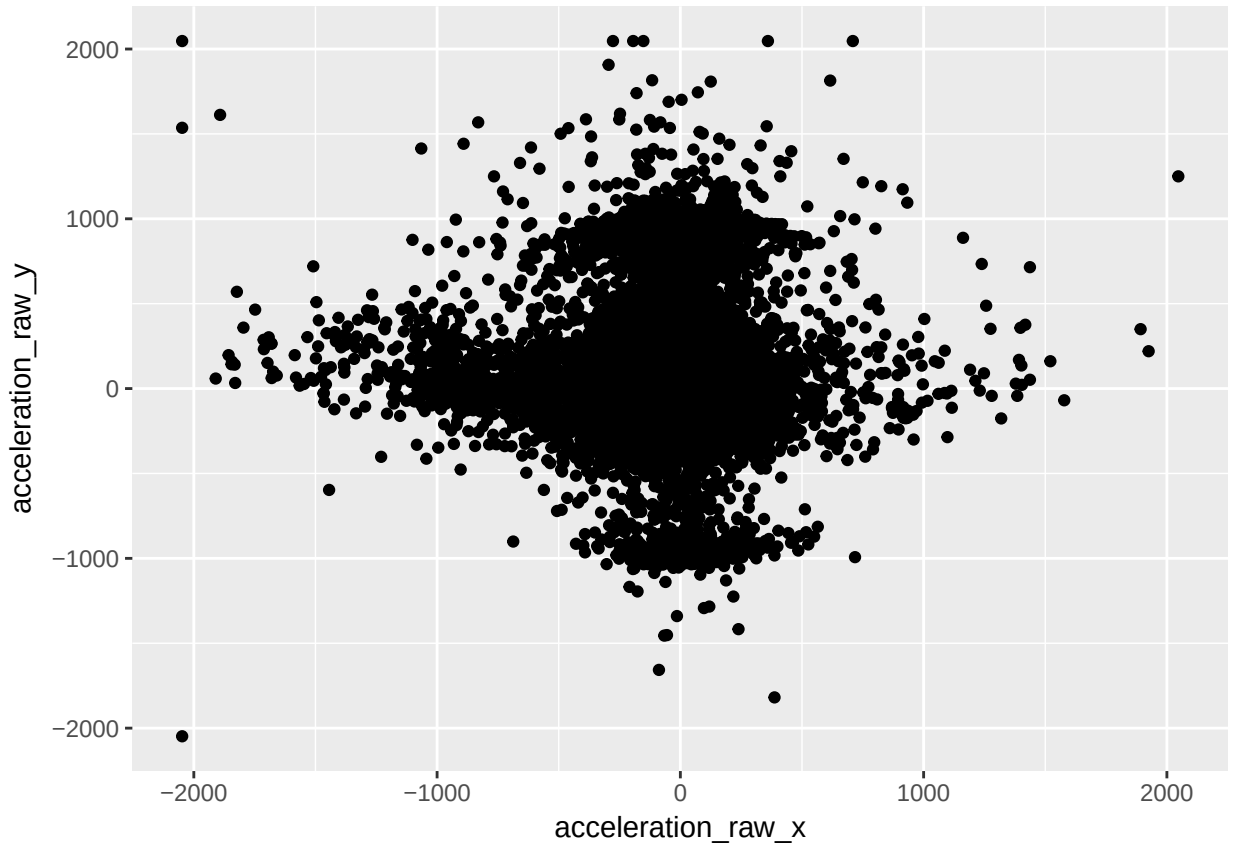
```
qplot(heading, acceleration_raw_x, data=data_crane)
```

```
## Warning: Removed 432 rows containing missing values (geom_point).
```



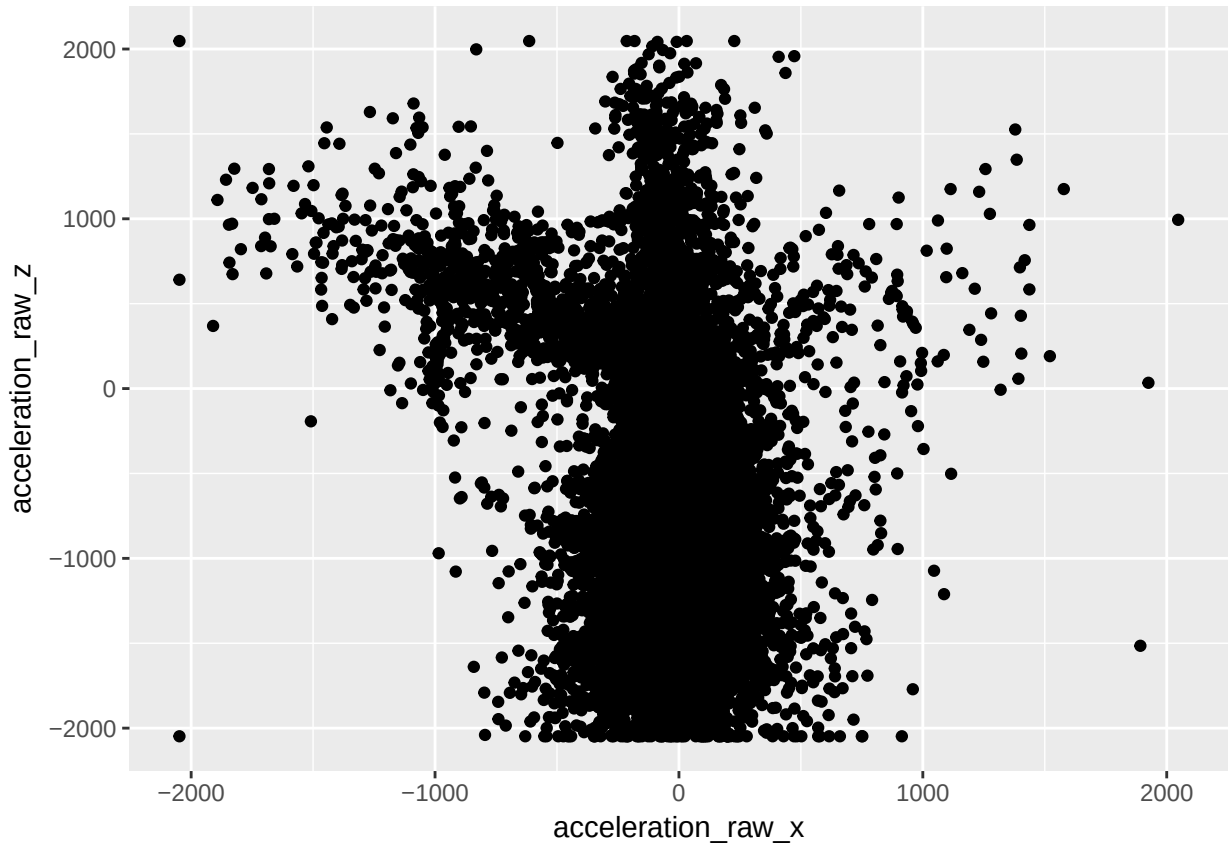

```
qplot(acceleration_raw_x, acceleration_raw_y, data=data_crane)
```

```
## Warning: Removed 432 rows containing missing values (geom_point).
```



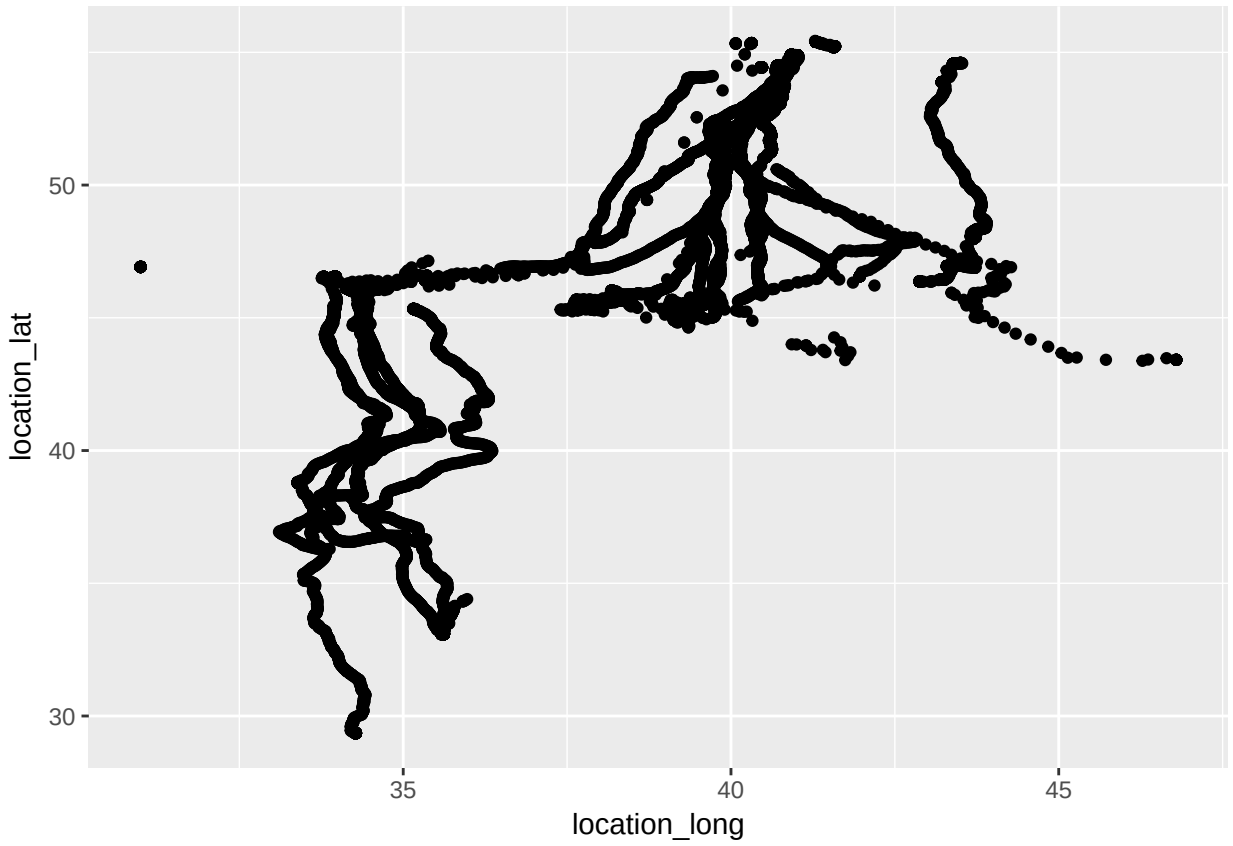

```
qplot(acceleration_raw_x, acceleration_raw_z, data=data_crane)
```

```
## Warning: Removed 432 rows containing missing values (geom_point).
```




```
qplot(location_long, location_lat, data=data_crane)
```

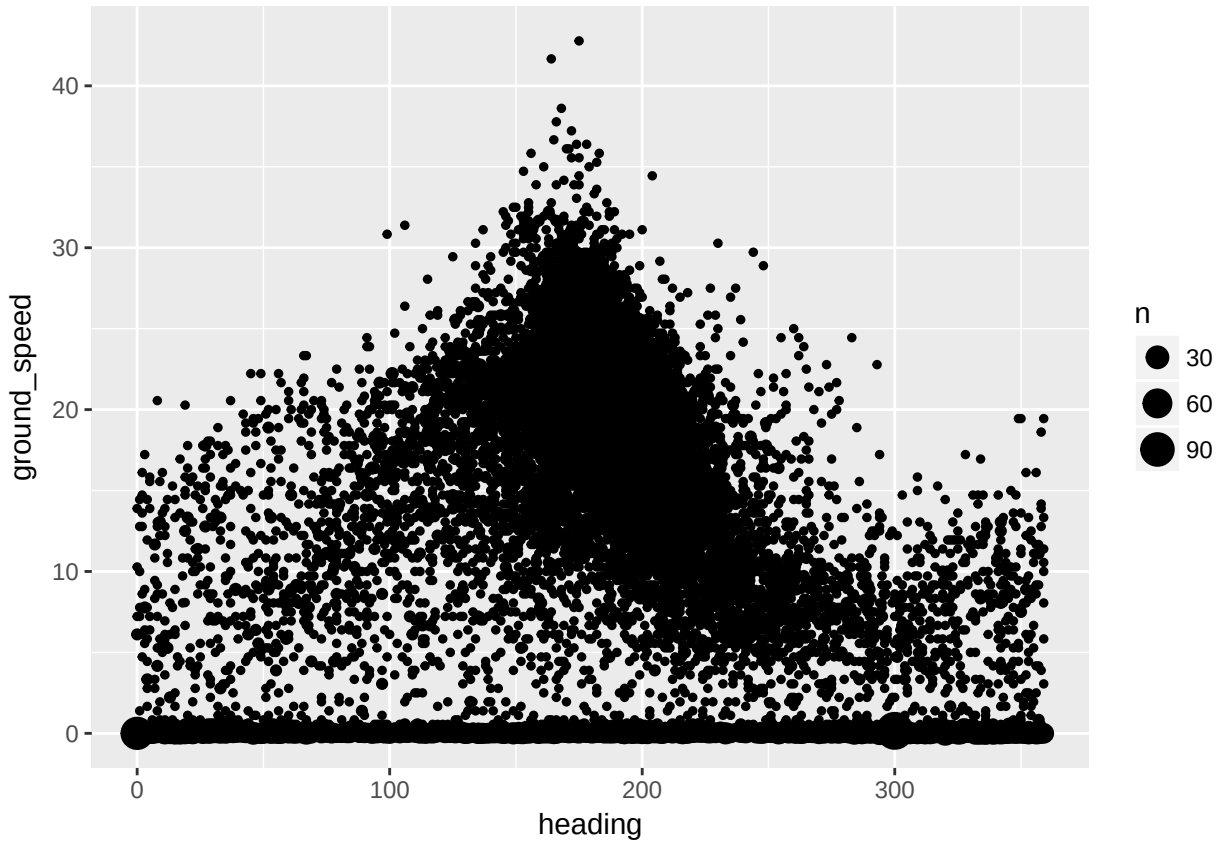
```
## Warning: Removed 200 rows containing missing values (geom_point).
```



Basic exercise II - Using ggplot for plotting

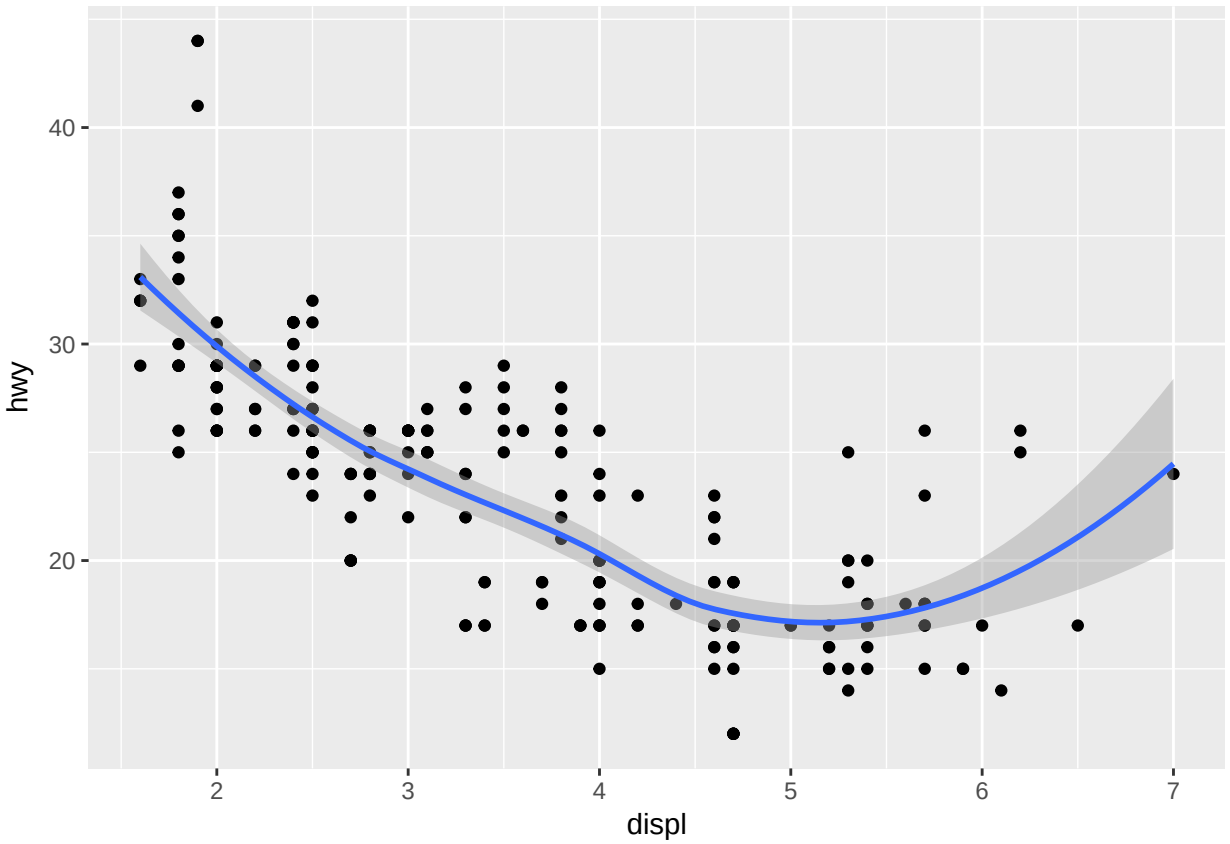
```
ggplot(data_crane, aes(heading, ground_speed)) +
  geom_count()
```

```
## Warning: Removed 40 rows containing non-finite values (stat_sum).
```



Reading exercise - Statistical layers for graphs.

Statistical layers reveal a strong power of ggplot. The following graph illustrates this:



Read Chapter 3.6 of R for Data Science and experiment with the function `geom_smooth()` and `stat_smooth()`. What is the difference between both functions?

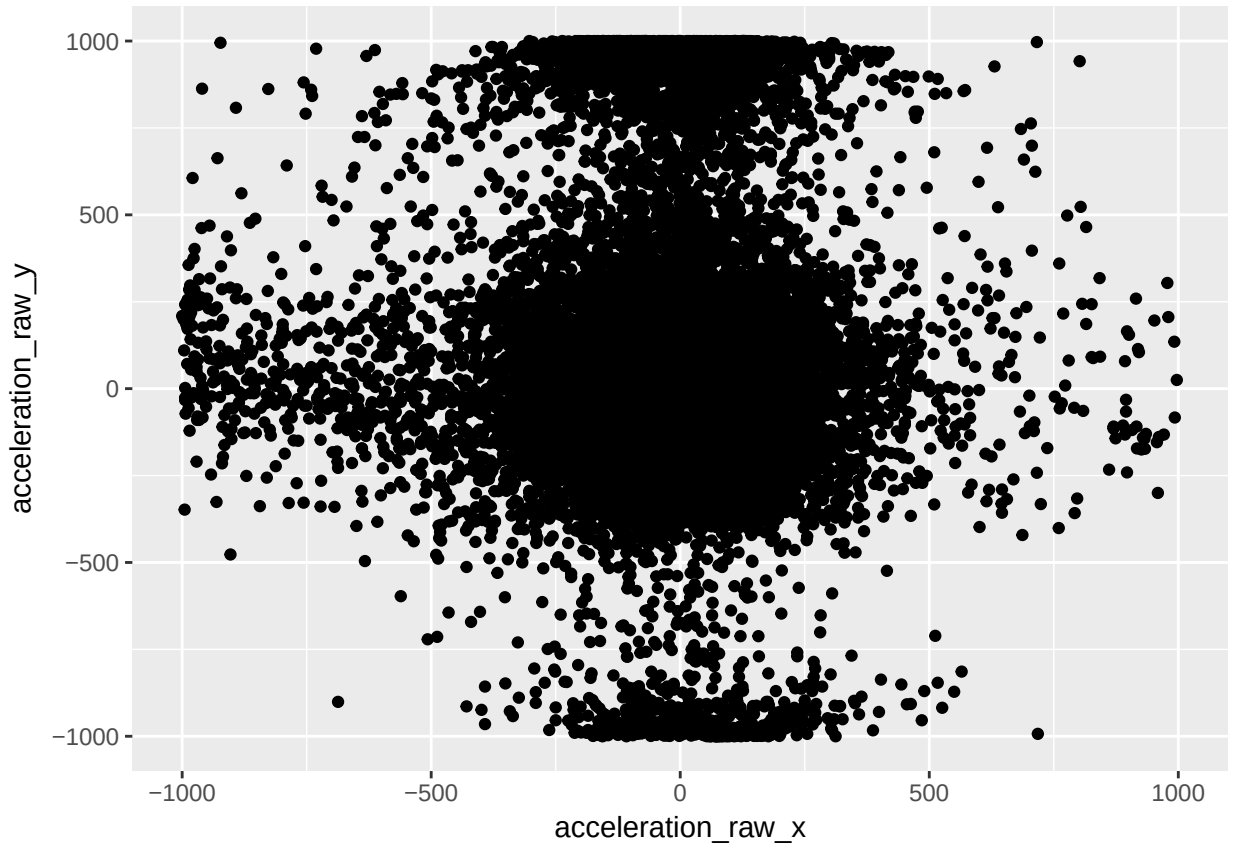
Optional exercise (+) - Scale axes

Scaling of the axes with a ggplot is easy. Take a look at the *data-visualization* cheat sheet. In the bottom right corner of the cheat sheet, you will find the code needed to scale the axes.

Zoom in on the acceleration between -1000 and 1000 for both the x-axis and y-axis.

```
ggplot(data_crane, aes(acceleration_raw_x, acceleration_raw_y)) +
  geom_point()+
  scale_x_continuous(limits = c(-1000, 1000)) +
  scale_y_continuous(limits = c(-1000, 1000))
```

```
## Warning: Removed 2436 rows containing missing values (geom_point).
```



Optional exercise (++) - Plot the crane positions on a map

a) Install package the `maps`

```
# install.packages('maps')
library(maps)
```

```
## Error in library(maps): there is no package called 'maps'
```

b) Plot the crane data on a map.

```
world_map_polygon <- map_data("world2")
```

```
## Error: Package `maps` required for `map_data`.
## Please install and try again.
```

```
ggplot(data_crane) +
  geom_map(data = world_map_polygon, map= world_map_polygon, aes(x=long, y = lat, map_id = region)) +
  scale_x_continuous(limits = c(0, 60)) +
  scale_y_continuous(limits = c(25, 60)) +
  geom_point(data = data_crane, aes(x = location_long, y = location_lat))
```

```
## Error in is.data.frame(map): object 'world_map_polygon' not found
```

c) Use an individual identifier to colour the different cranes.

```
world_map_polygon <- map_data("world2")
```

```
## Error: Package `maps` required for `map_data`.
## Please install and try again.
```

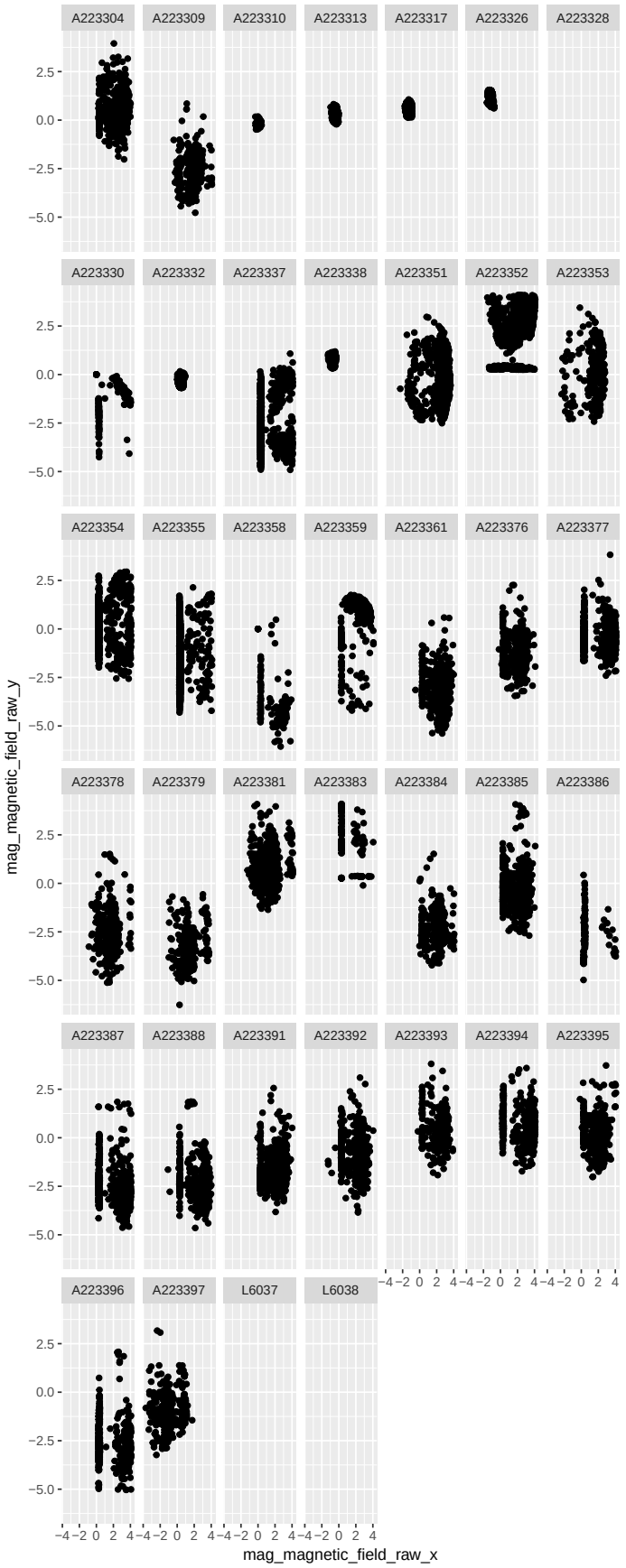
```
ggplot(data_crane) +
  geom_map(data = world_map_polygon, map= world_map_polygon, aes(long, lat, map_id = region)) +
  scale_x_continuous(limits = c(0, 60)) +
  scale_y_continuous(limits = c(25, 60)) +
  geom_point(data = data_crane, aes(x = location_long, y = location_lat, colour=individual_local_identi
```

```
## Error in is.data.frame(map): object 'world_map_polygon' not found
```

Optional exercise (+++) - Create facets.

```
ggplot(data_crane, aes(mag_magnetic_field_raw_x, mag_magnetic_field_raw_y)) +
  geom_point() +
  facet_wrap(~individual_local_identifier)
```

```
## Warning: Removed 432 rows containing missing values (geom_point).
```


A223304
A223309
A223310
A223313
A223317
A223326
A223328
A223330
A223332
A223337
A223338
A223351
A223352
A223353
A223354
A223355
A223358
A223359
A223361
A223376
A223377
A223378
A223379
A223381
A223383
A223384
A223385
A223386
A223387
A223388
A223391
A223392
A223393
A223394
A223395
A223396
A223397
L6037
L6038
mag_magnetic_field_raw_y
mag_magnetic_field_raw_x
2.5
0.0
−2.5
−5.0
−4 −2 0 2 4

3. Data transformation

Basic exercise I - Subset data (single crane)

a) Make a selection of all observations of crane 'L6037'.

```
filter(data_crane, individual_local_identifier=='L6037')
```

```
## # A tibble: 61 x 43
##    event_id visible timestamp           location_long location_lat
##       <dbl> <chr>   <dttm>                      <dbl>        <dbl>
##  1   3.79e9 true    2017-10-15 00:00:16          40.3         55.3
##  2   3.80e9 true    2017-10-15 00:00:16          40.3         55.3
##  3   3.79e9 true    2017-10-15 02:00:23          40.3         55.3
##  4   3.80e9 true    2017-10-15 02:00:23          40.3         55.3
##  5   3.79e9 true    2017-10-15 04:00:23          40.3         55.3
##  6   3.80e9 true    2017-10-15 04:00:23          40.3         55.3
##  7   3.79e9 true    2017-10-15 05:00:14          40.1         55.3
##  8   3.80e9 true    2017-10-15 05:00:14          40.1         55.3
##  9   3.80e9 true    2017-10-15 06:00:23          40.1         55.3
## 10   3.80e9 true    2017-10-15 08:00:18          40.1         55.3
## # ... with 51 more rows, and 38 more variables: acceleration_raw_x <dbl>,
## #   acceleration_raw_y <dbl>, acceleration_raw_z <dbl>,
## #   bar_barometric_height <dbl>, battery_charge_percent <int>,
## #   battery_charging_current <dbl>, eobs_activity <chr>,
## #   eobs_activity_samples <chr>, eobs_battery_voltage <int>,
## #   eobs_fix_battery_voltage <int>,
## #   eobs_horizontal_accuracy_estimate <dbl>, eobs_key_bin_checksum <dbl>,
## #   eobs_speed_accuracy_estimate <dbl>, eobs_start_timestamp <chr>,
## #   eobs_status <chr>, eobs_temperature <int>, eobs_type_of_fix <int>,
## #   eobs_used_time_to_get_fix <int>, external_temperature <dbl>,
## #   gps_fix_type <chr>, gps_hdop <dbl>, gps_satellite_count <int>,
## #   gps_time_to_fix <dbl>, ground_speed <dbl>, heading <dbl>,
## #   height_above_ellipsoid <dbl>, height_above_msl <dbl>,
## #   import_marked_outlier <chr>, gls_light_level <dbl>,
## #   mag_magnetic_field_raw_x <dbl>, mag_magnetic_field_raw_y <dbl>,
## #   mag_magnetic_field_raw_z <dbl>, orn_transmission_protocol <chr>,
## #   tag_voltage <dbl>, sensor_type <chr>,
## #   individual_taxon_canonical_name <chr>,
## #   individual_local_identifier <chr>, study_name <chr>
```

b) Make a selection of all observations of crane 'L6037' where the variable `eobs_status` isn't missing.

```
filter(data_crane, individual_local_identifier=='L6037', !is.na(eobs_status))
```

```
## # A tibble: 42 x 43
##    event_id visible timestamp           location_long location_lat
##       <dbl> <chr>   <dttm>                      <dbl>        <dbl>
##  1   3.80e9 true    2017-10-15 00:00:16          40.3         55.3
##  2   3.80e9 true    2017-10-15 02:00:23          40.3         55.3
##  3   3.80e9 true    2017-10-15 04:00:23          40.3         55.3
##  4   3.80e9 true    2017-10-15 05:00:14          40.1         55.3
##  5   3.80e9 true    2017-10-15 06:00:23          40.1         55.3
```

```
##  6   3.80e9 true    2017-10-15 08:00:18          40.1         55.3
##  7   3.80e9 true    2017-10-15 10:00:23          40.1         55.3
##  8   3.80e9 true    2017-10-15 12:00:14          40.1         55.3
##  9   3.80e9 true    2017-10-15 14:00:23          40.3         55.3
## 10   3.80e9 true    2017-10-15 16:00:23          40.3         55.3
## # ... with 32 more rows, and 38 more variables: acceleration_raw_x <dbl>,
## #   acceleration_raw_y <dbl>, acceleration_raw_z <dbl>,
## #   bar_barometric_height <dbl>, battery_charge_percent <int>,
## #   battery_charging_current <dbl>, eobs_activity <chr>,
## #   eobs_activity_samples <chr>, eobs_battery_voltage <int>,
## #   eobs_fix_battery_voltage <int>,
## #   eobs_horizontal_accuracy_estimate <dbl>, eobs_key_bin_checksum <dbl>,
## #   eobs_speed_accuracy_estimate <dbl>, eobs_start_timestamp <chr>,
## #   eobs_status <chr>, eobs_temperature <int>, eobs_type_of_fix <int>,
## #   eobs_used_time_to_get_fix <int>, external_temperature <dbl>,
## #   gps_fix_type <chr>, gps_hdop <dbl>, gps_satellite_count <int>,
## #   gps_time_to_fix <dbl>, ground_speed <dbl>, heading <dbl>,
## #   height_above_ellipsoid <dbl>, height_above_msl <dbl>,
## #   import_marked_outlier <chr>, gls_light_level <dbl>,
## #   mag_magnetic_field_raw_x <dbl>, mag_magnetic_field_raw_y <dbl>,
## #   mag_magnetic_field_raw_z <dbl>, orn_transmission_protocol <chr>,
## #   tag_voltage <dbl>, sensor_type <chr>,
## #   individual_taxon_canonical_name <chr>,
## #   individual_local_identifier <chr>, study_name <chr>
```

c) Make a selection of all observations of crane 'L6037' where the variable `eobs_status` isn't missing. Return only the `event_id` and `timestamp` variables.

```
data_crane_filter <- filter(data_crane, individual_local_identifier=='L6037')
select(data_crane_filter, event_id, timestamp)
```

```
## # A tibble: 61 x 2
##      event_id timestamp
##         <dbl> <dttm>
##  1 3794510958 2017-10-15 00:00:16
##  2 3796034354 2017-10-15 00:00:16
##  3 3794510962 2017-10-15 02:00:23
##  4 3796034355 2017-10-15 02:00:23
##  5 3794510961 2017-10-15 04:00:23
##  6 3796034356 2017-10-15 04:00:23
##  7 3794510960 2017-10-15 05:00:14
##  8 3796034357 2017-10-15 05:00:14
##  9 3796034358 2017-10-15 06:00:23
## 10 3796034359 2017-10-15 08:00:18
## # ... with 51 more rows
```

Basic exercise II - Compute the magnitude of the magnetic field

```
data_crane_with_mag_magnetic <- mutate(
  data_crane,
  magnitude = sqrt(mag_magnetic_field_raw_x ^ 2 +
                     mag_magnetic_field_raw_y ^ 2 +
                     mag_magnetic_field_raw_y ^ 2)
```

```
)

View(data_crane_with_mag_magnetic)
```

Reading exercise - Pipe operator

The pipe operator %>% is used by many tidyverse users. Read Chapter 18 of R for Data Science. Make a selection columns and rows of the data and make use of the pipe syntax.

Optional exercise (+) - Exclude variables

```
select(data_crane, -starts_with('location'))
```

```
## # A tibble: 20,873 x 41
##    event_id visible timestamp           acceleration_ra~ acceleration_ra~
##       <dbl> <chr>   <dttm>                         <dbl>            <dbl>
##  1   3.79e9 true    2017-10-15 00:00:16               NA               NA
##  2   3.80e9 true    2017-10-15 00:00:16               NA               NA
##  3   3.79e9 true    2017-10-15 02:00:23               NA               NA
##  4   3.80e9 true    2017-10-15 02:00:23               NA               NA
##  5   3.79e9 true    2017-10-15 04:00:23               NA               NA
##  6   3.80e9 true    2017-10-15 04:00:23               NA               NA
##  7   3.79e9 true    2017-10-15 05:00:14               NA               NA
##  8   3.80e9 true    2017-10-15 05:00:14               NA               NA
##  9   3.80e9 true    2017-10-15 06:00:23               NA               NA
## 10   3.80e9 true    2017-10-15 08:00:18               NA               NA
## # ... with 20,863 more rows, and 36 more variables:
## #   acceleration_raw_z <dbl>, bar_barometric_height <dbl>,
## #   battery_charge_percent <int>, battery_charging_current <dbl>,
## #   eobs_activity <chr>, eobs_activity_samples <chr>,
## #   eobs_battery_voltage <int>, eobs_fix_battery_voltage <int>,
## #   eobs_horizontal_accuracy_estimate <dbl>, eobs_key_bin_checksum <dbl>,
## #   eobs_speed_accuracy_estimate <dbl>, eobs_start_timestamp <chr>,
## #   eobs_status <chr>, eobs_temperature <int>, eobs_type_of_fix <int>,
## #   eobs_used_time_to_get_fix <int>, external_temperature <dbl>,
## #   gps_fix_type <chr>, gps_hdop <dbl>, gps_satellite_count <int>,
## #   gps_time_to_fix <dbl>, ground_speed <dbl>, heading <dbl>,
## #   height_above_ellipsoid <dbl>, height_above_msl <dbl>,
## #   import_marked_outlier <chr>, gls_light_level <dbl>,
## #   mag_magnetic_field_raw_x <dbl>, mag_magnetic_field_raw_y <dbl>,
## #   mag_magnetic_field_raw_z <dbl>, orn_transmission_protocol <chr>,
## #   tag_voltage <dbl>, sensor_type <chr>,
## #   individual_taxon_canonical_name <chr>,
## #   individual_local_identifier <chr>, study_name <chr>
```

Optional exercise (++) - Summarise results

```
summarise(
  data_crane,
  min_latitude = max(location_lat, na.rm=T),
```

```
  first_observation = min(timestamp),
  magnitude_acceleration = mean(sqrt(acceleration_raw_x^2+acceleration_raw_y^2+acceleration_raw_x^2), na
)
```

```
## # A tibble: 1 x 3
##   min_latitude first_observation   magnitude_acceleration
##          <dbl> <dttm>                               <dbl>
## 1         55.4 2017-10-15 00:00:05                   516.
```

Optional exercise (+++) - Join datasets

```
data_crane_with_measures <- left_join(data_crane, data_crane_additional, by=`event_id`)
```

```
## Error in common_by(by, x, y): object 'event_id' not found
```