# Data visualization

for incomplete datasets in R

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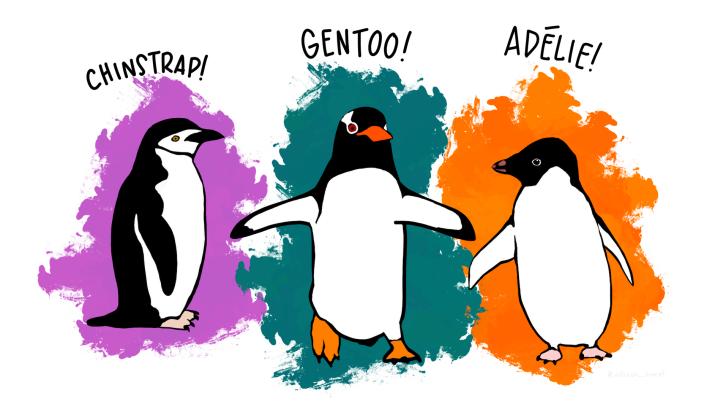
## Missingness





### Case study

```
1 set.seed(123)
2 library(palmerpenguins)
3 library(mice)
4 library(ggmice)
5 library(ggplot2)
```





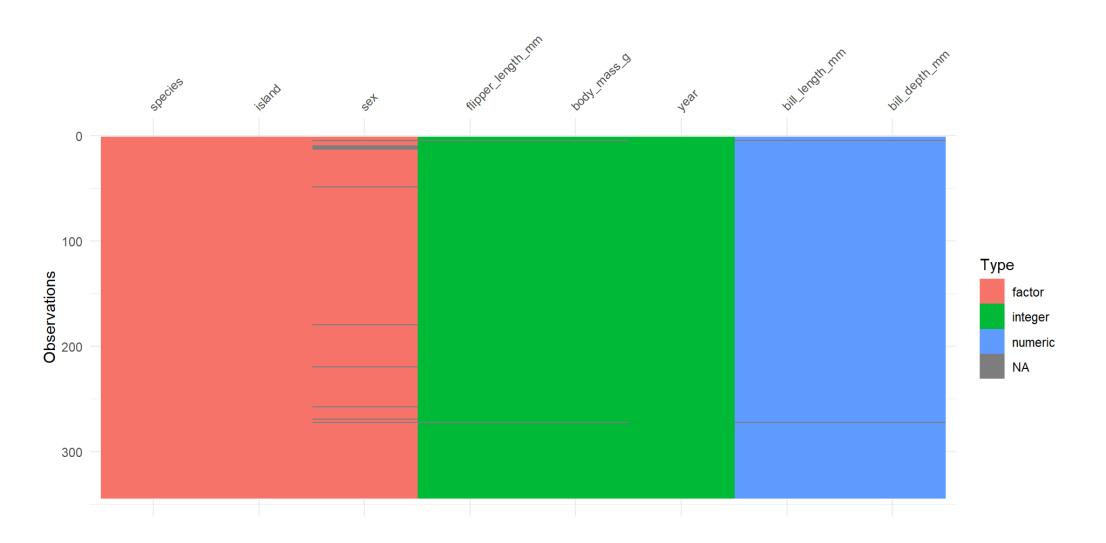
### Incomplete data

```
1 str(penguins)
tibble [344 \times 8] (S3: tbl df/tbl/data.frame)
 $ species
           : Factor w/ 3 levels "Adelie", "Chinstrap", ...: 1 1 1 1 1 1
1 1 1 1 . . .
              : Factor w/ 3 levels "Biscoe", "Dream", ...: 3 3 3 3 3 3 3
 $ island
3 3 . . .
 $ bill length mm : num [1:344] 39.1 39.5 40.3 NA 36.7 39.3 38.9 39.2 34.1
42 . . .
 $ bill depth mm : num [1:344] 18.7 17.4 18 NA 19.3 20.6 17.8 19.6 18.1
20.2 ...
 $ flipper length mm: int [1:344] 181 186 195 NA 193 190 181 195 193 190 ...
                   : int [1:344] 3750 3800 3250 NA 3450 3650 3625 4675 3475
 $ body mass g
4250 ...
                   : Factor w/ 2 levels "female", "male": 2 1 1 NA 1 2 1 2 NA
 $ sex
NA ...
 $ year
                   \bigcirc \bigcirc \bigcirc \bigcirc
```



### Incomplete data

1 visdat::vis\_dat(penguins)



### Response indicator

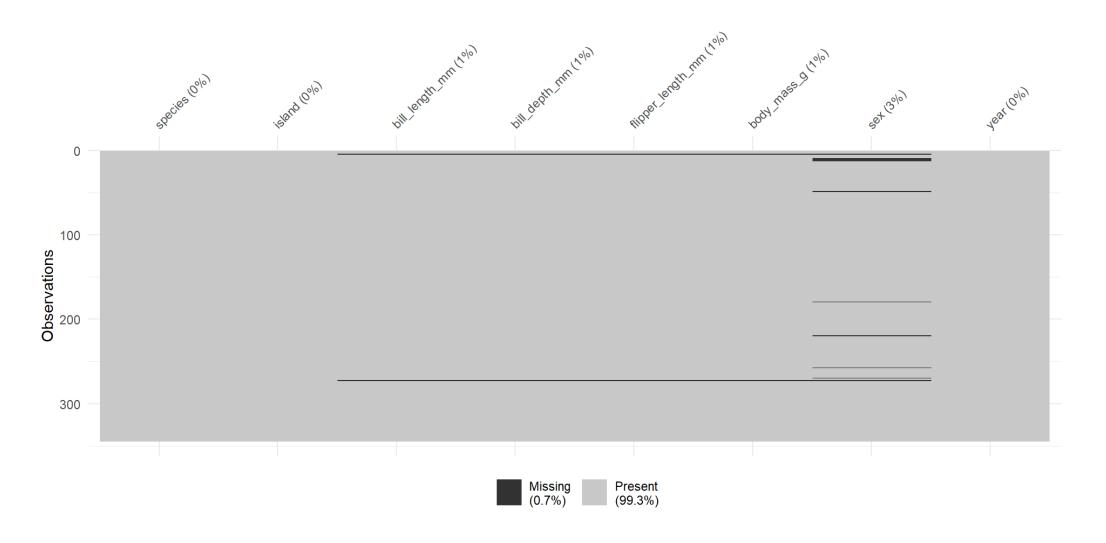
1 is.na(penguins)

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm
[1,]	FALSE	FALSE	FALSE	FALSE	FALSE
[2,]	FALSE	FALSE	FALSE	FALSE	FALSE
[3,]	FALSE	FALSE	FALSE	FALSE	FALSE
[4,]	FALSE	FALSE	TRUE	TRUE	TRUE
[5,]	FALSE	FALSE	FALSE	FALSE	FALSE
[6,]	FALSE	FALSE	FALSE	FALSE	FALSE
[7,]	FALSE	FALSE	FALSE	FALSE	FALSE
[8,]	FALSE	FALSE	FALSE	FALSE	FALSE
[9,]	FALSE	FALSE	FALSE	FALSE	FALSE
[10,]	FALSE	FALSE	FALSE	FALSE	FALSE
[11,]	FALSE	FALSE	FALSE	FALSE	FALSE
[12,]	FALSE	FALSE	FALSE	FALSE	FALSE
[13,]	FALSE	FALSE	FALSE	FALSE	FALSE
[14,]	FALSE	FALSE	FALSE	FALSE	FALSE
гаг п	<b>□</b> 7 7 <b>□</b> □			D 7 T O D	T7 T 7 T



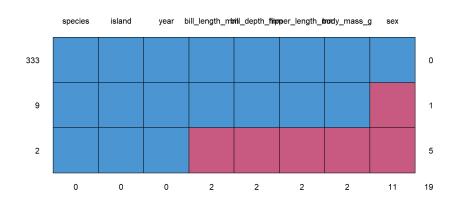
### Response indicator

1 naniar::vis\_miss(penguins)





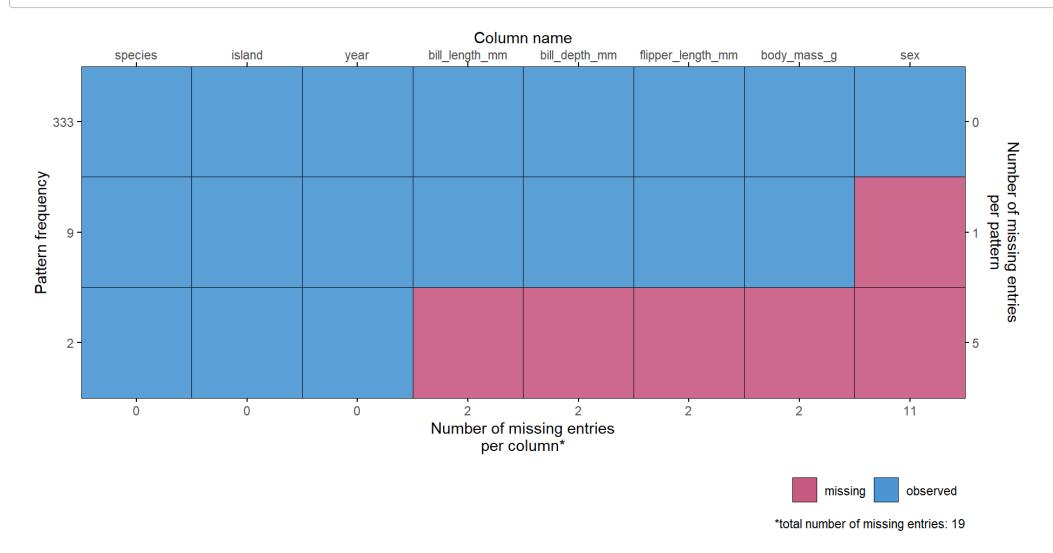
### Missing data pattern





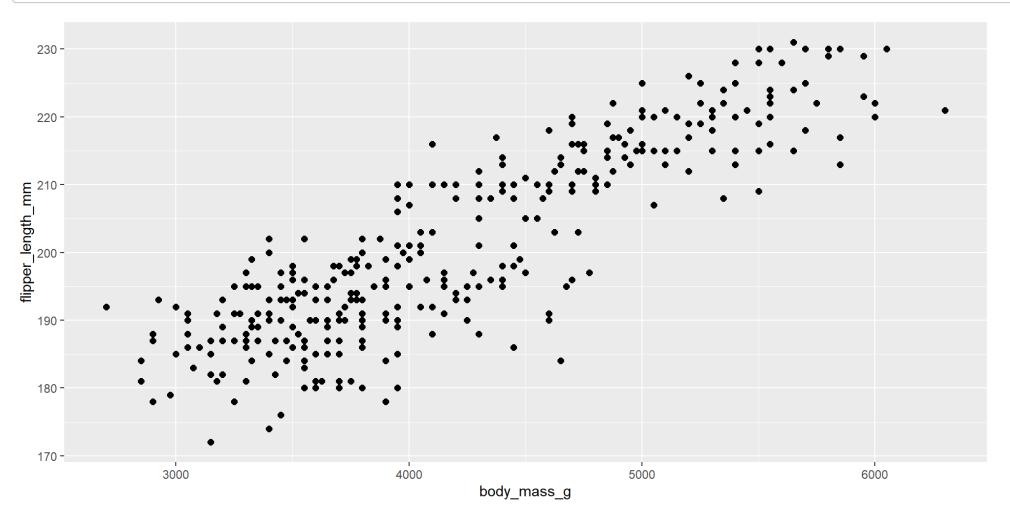
### Missing data pattern

1 plot\_pattern(penguins)



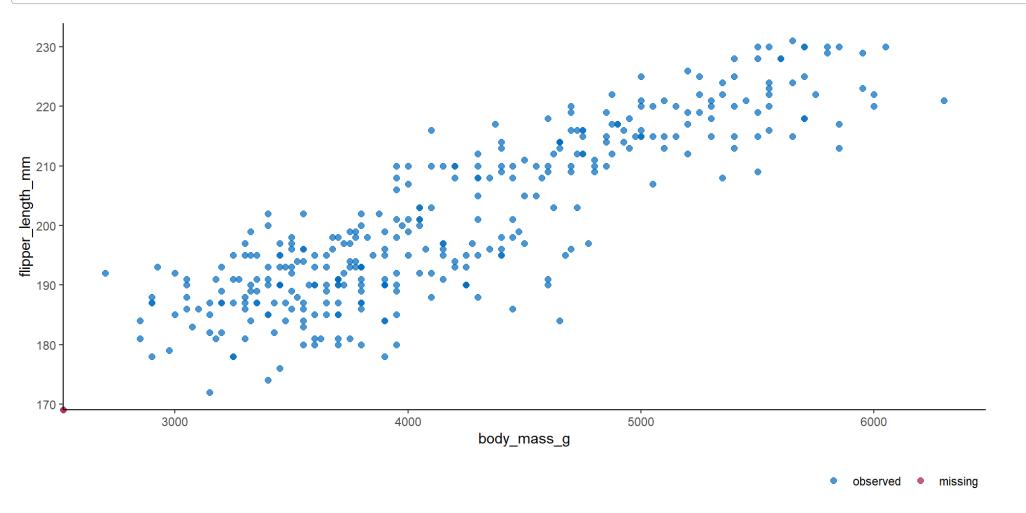


```
1 ggplot(penguins, aes(body_mass_g, flipper_length_mm)) +
2 geom_point(size = 2)
```



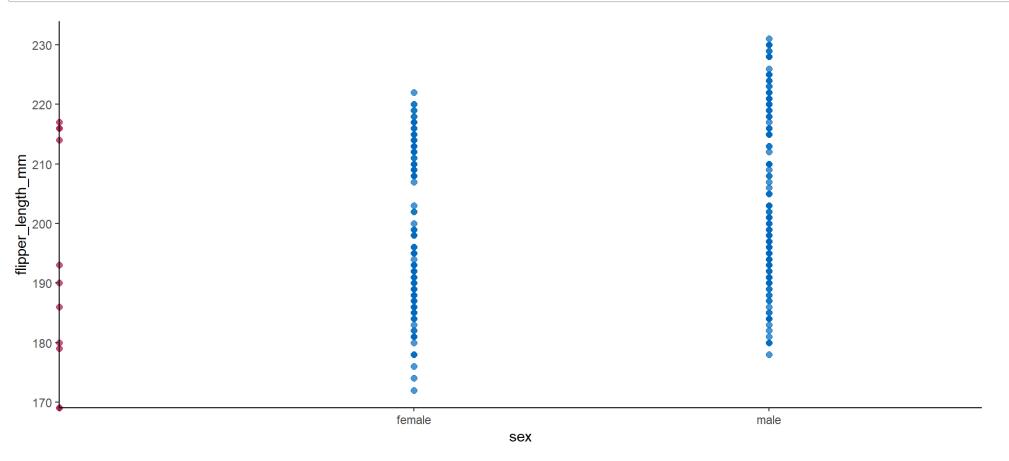


```
1 ggmice(penguins, aes(body_mass_g, flipper_length_mm)) +
2 geom_point(size = 2)
```





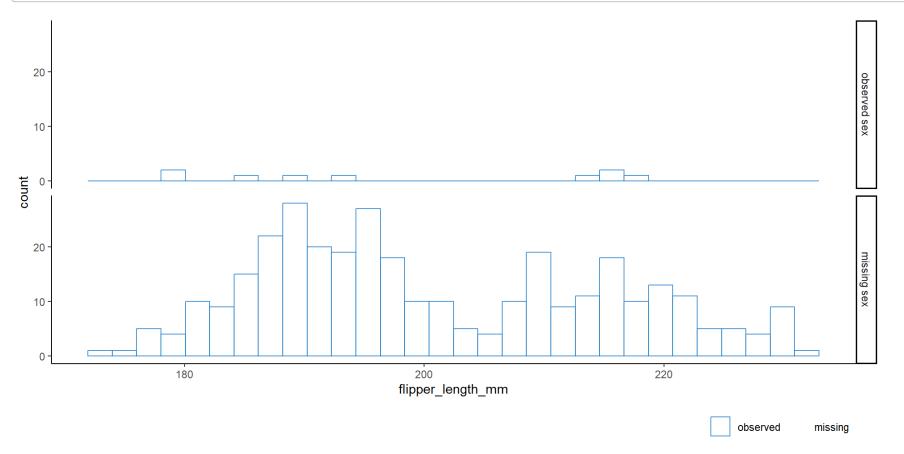
```
1 ggmice(penguins, aes(sex, flipper_length_mm)) +
2 geom_point(size = 2)
```





#### **Faceted distribution**

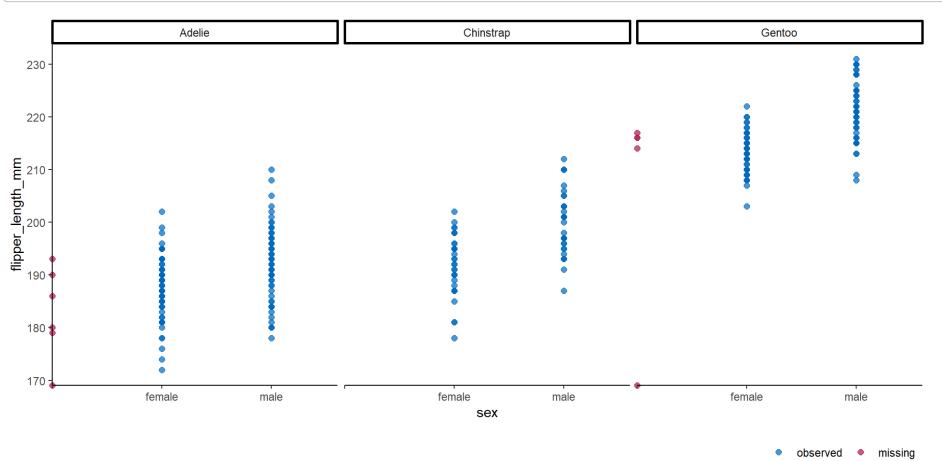
```
ggmice(penguins, aes(flipper_length_mm)) +
geom_histogram(fill = "white") +
facet_grid(
factor(is.na(sex) == 0, labels = c("observed sex", "missing sex")) ~ .)
```





### Faceted scatter plot

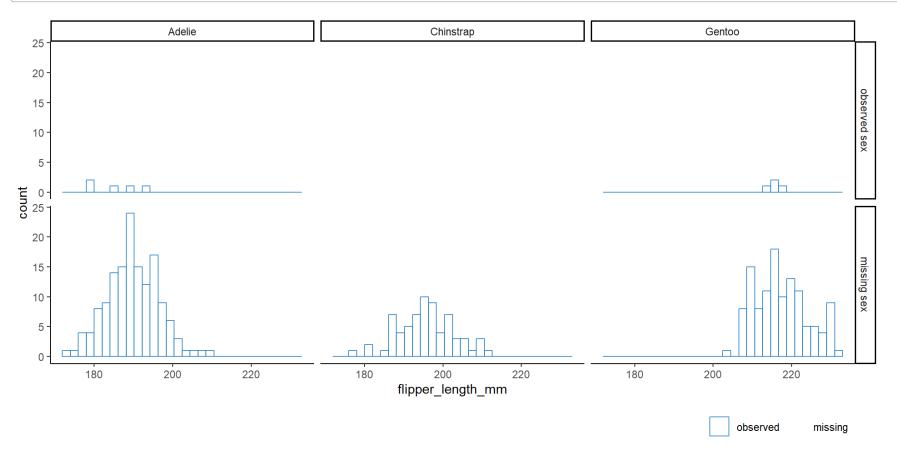
```
1 ggmice(penguins, aes(sex, flipper_length_mm)) +
2 geom_point(size = 2) +
3 facet_wrap(~species)
```





#### **Faceted distribution**

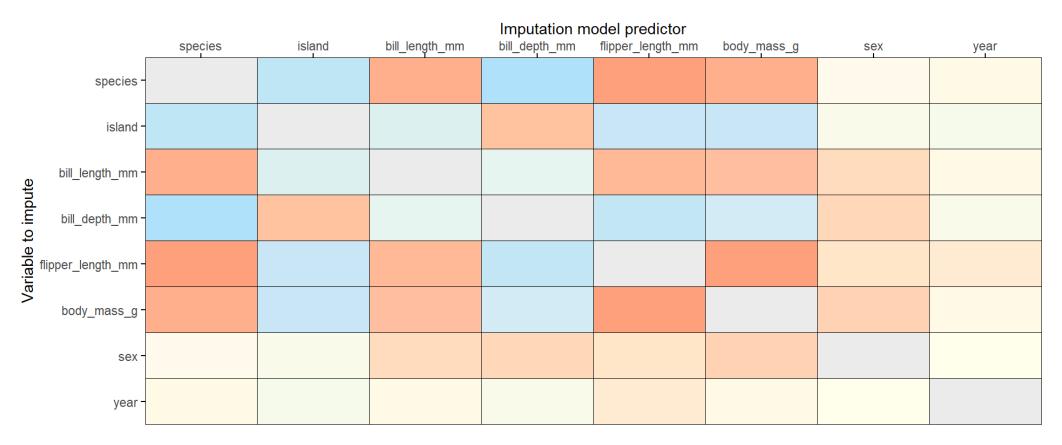
```
1 ggmice(penguins, aes(flipper_length_mm)) +
2   geom_histogram(fill = "white") +
3   facet_grid(
4   factor(is.na(sex) == 0, labels = c("observed sex", "missing sex")) ~ sp
```

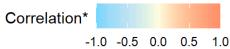




### Correlation

1 plot\_corr(penguins, square = FALSE)



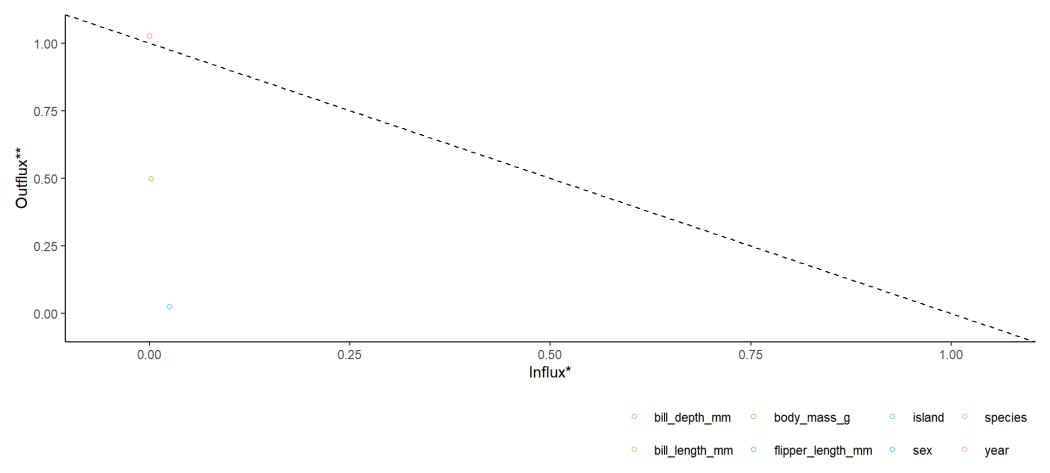


\*pairwise complete observations



### Flux plot

```
1 plot flux(penguins, label = FALSE)
```



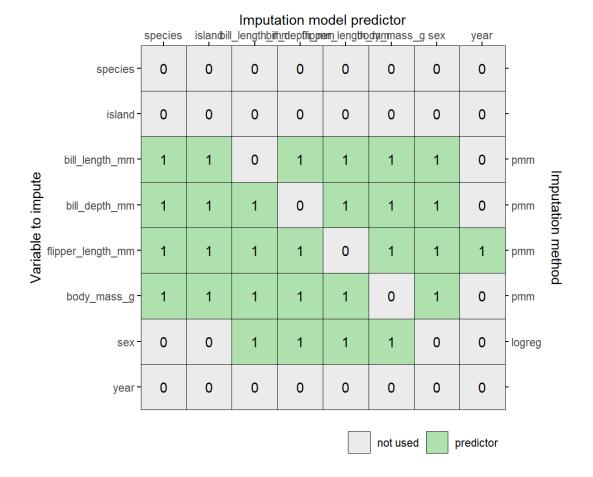
\*connection of a variable's missingness indicator with observed data on other variables

\*\*connection of a variable's observed data with missing data on other variables



### Imputation models

```
1 pred <- quickpred(penguins)
2 meth <- make.method(penguins)
3 plot_pred(pred, method = meth)</pre>
```





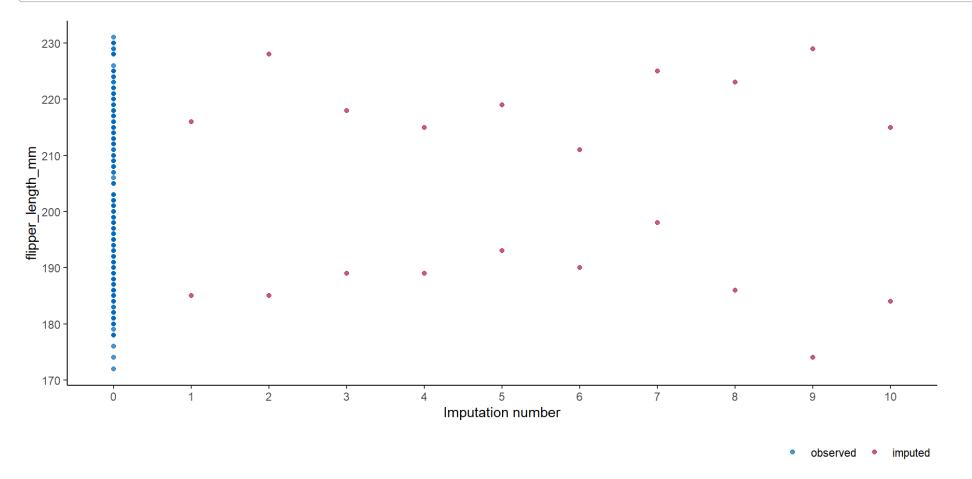
### **Impute**

```
1 imp <- mice(penguins, pred = pred, method = meth, m = 10, print = FALSE)
2 plot_trace(imp)</pre>
```



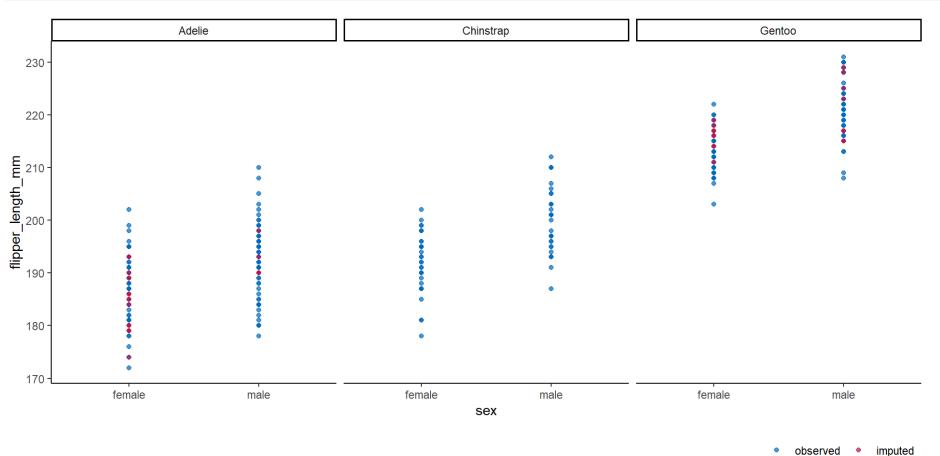
### Boxplot

```
1 ggmice(imp, aes(x = .imp, y = flipper_length_mm)) +
2 geom_point() +
3 labs(x = "Imputation number")
```





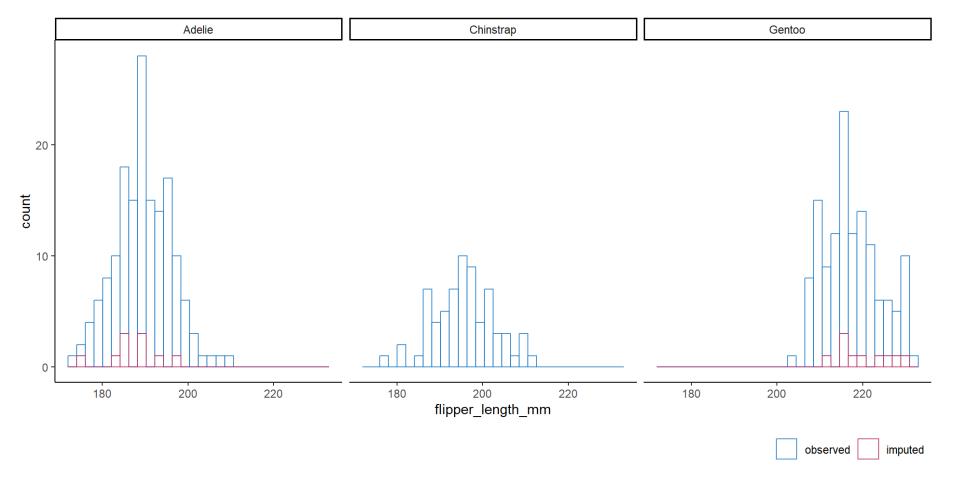
```
1 ggmice(imp, aes(sex, flipper_length_mm)) +
2 geom_point() +
3 facet_grid(~species)
```





#### **Faceted distribution**

```
1 ggmice(imp, aes(flipper_length_mm)) +
2 geom_histogram(fill = "white") +
3 facet_grid(~ species)
```





# Thank you!



