





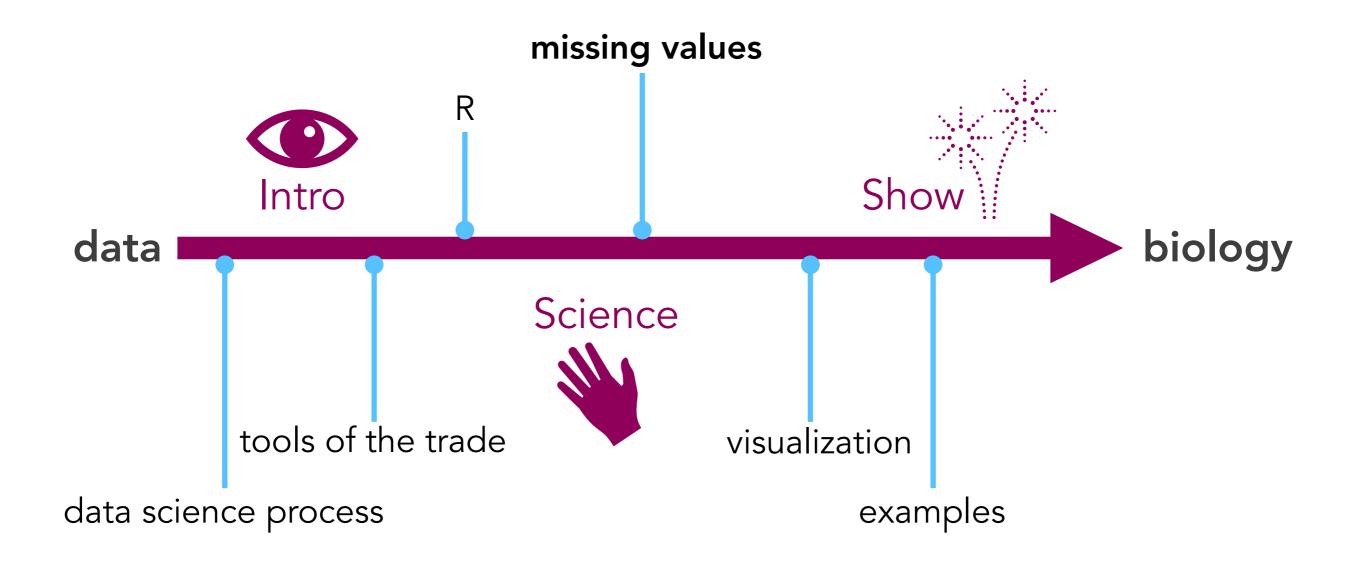
## Missing value imputation

10 / 09 - 2019

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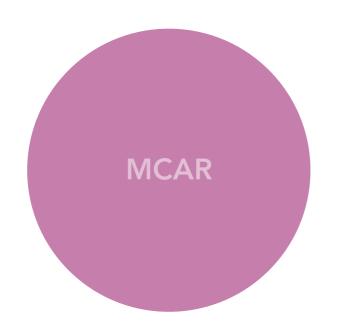
matthias.stahl@ki.se

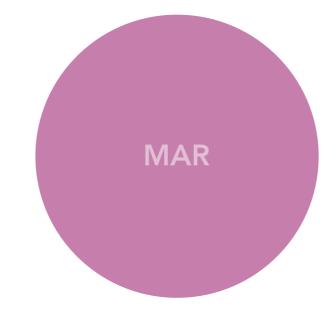


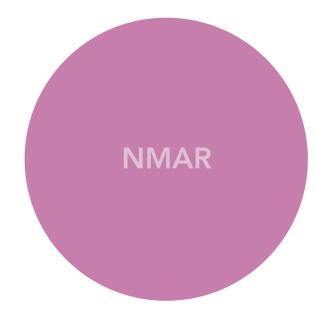












missing completely at random

missing at random

not missing at random

no cause to missingness

missing values can be explained by the available data

missing data can be ignored

missing data cannot be ignored

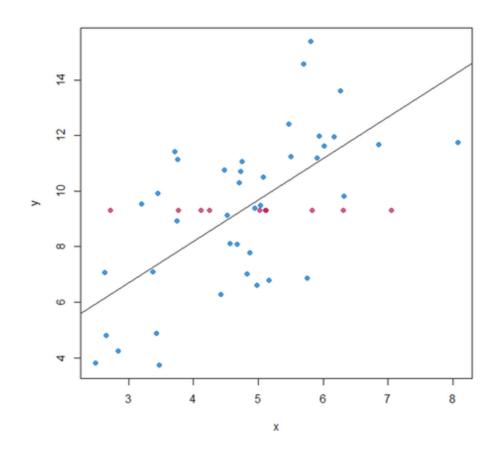


# There are numerous methods to fill missing values.



## Keep the sample statistics

impute with **mean | median | mode** of the data



not optimal!

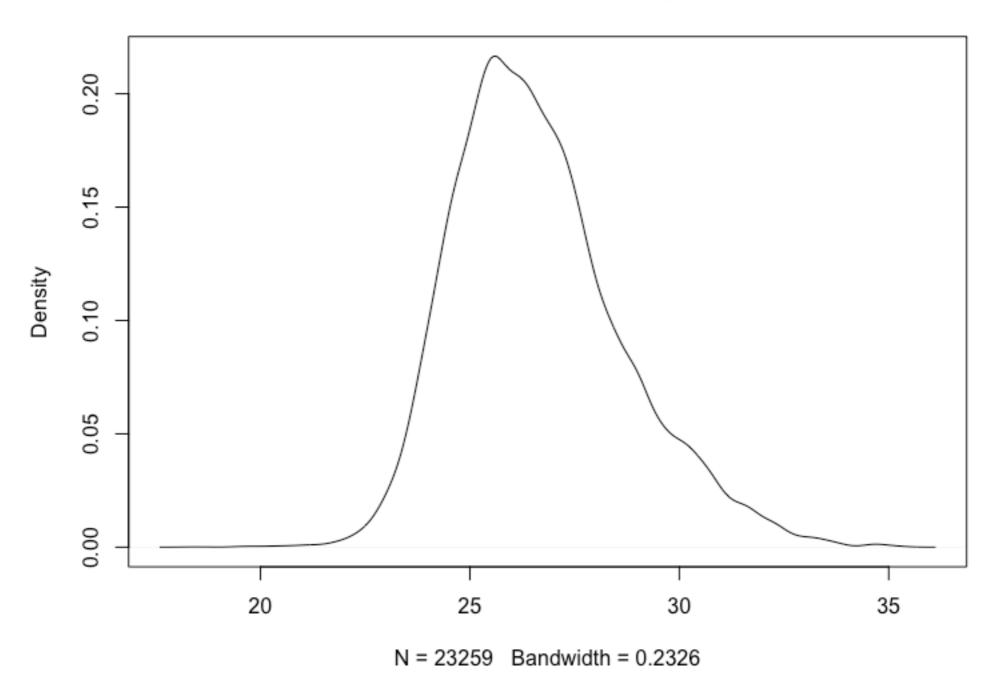
changes variance, introduces bias, ...





## A usual dataset

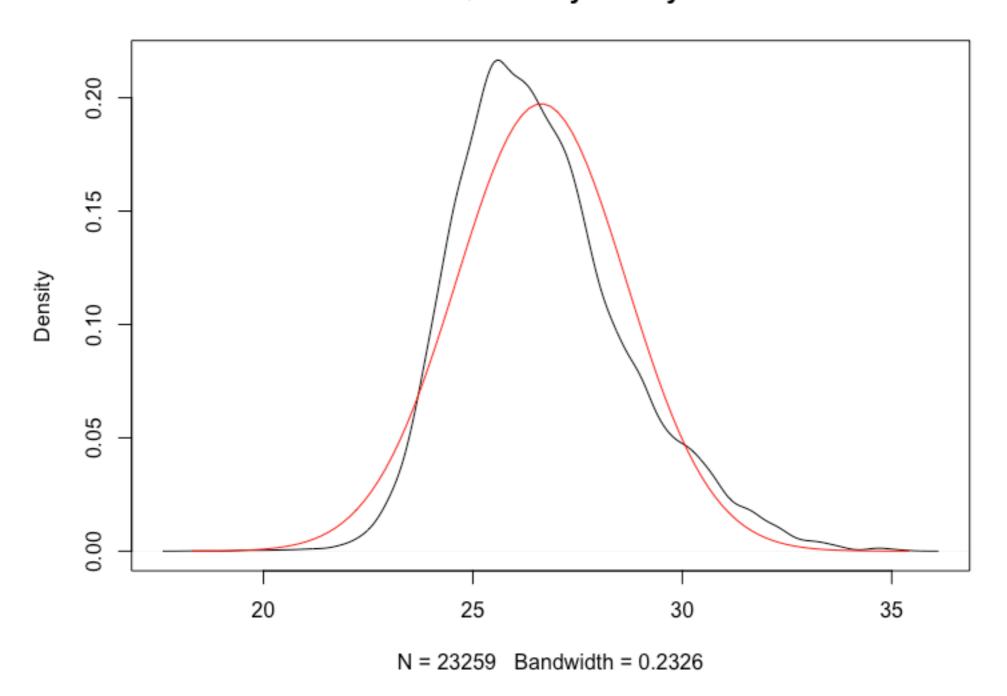
Caro's dataset (HepG2 + **NC1** or DMSO)





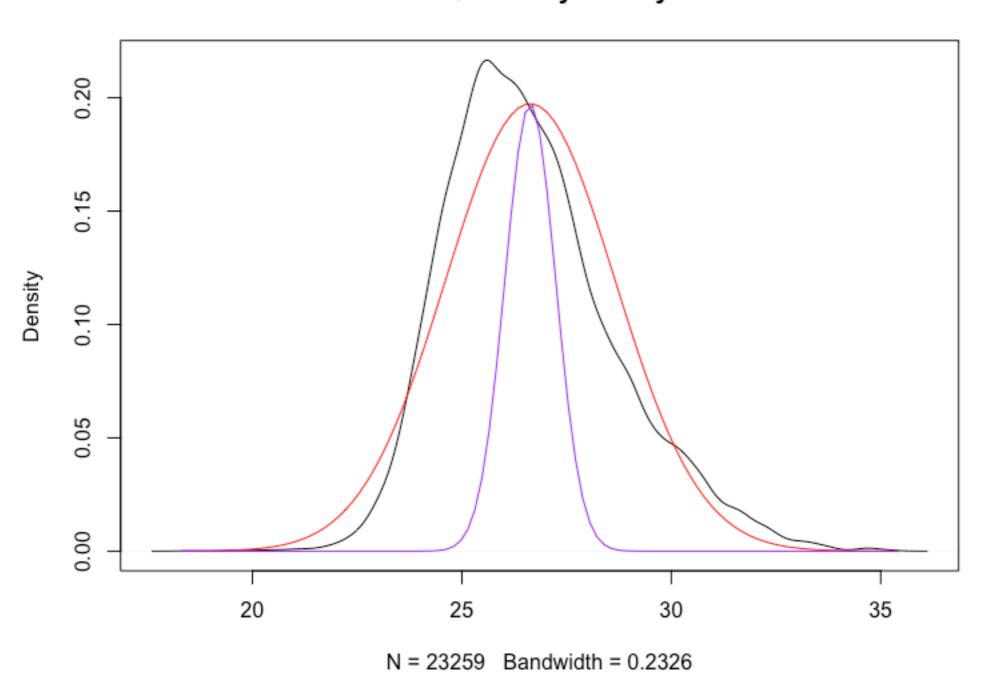


Disclaimer: This is how I understood Perseus does it.;)



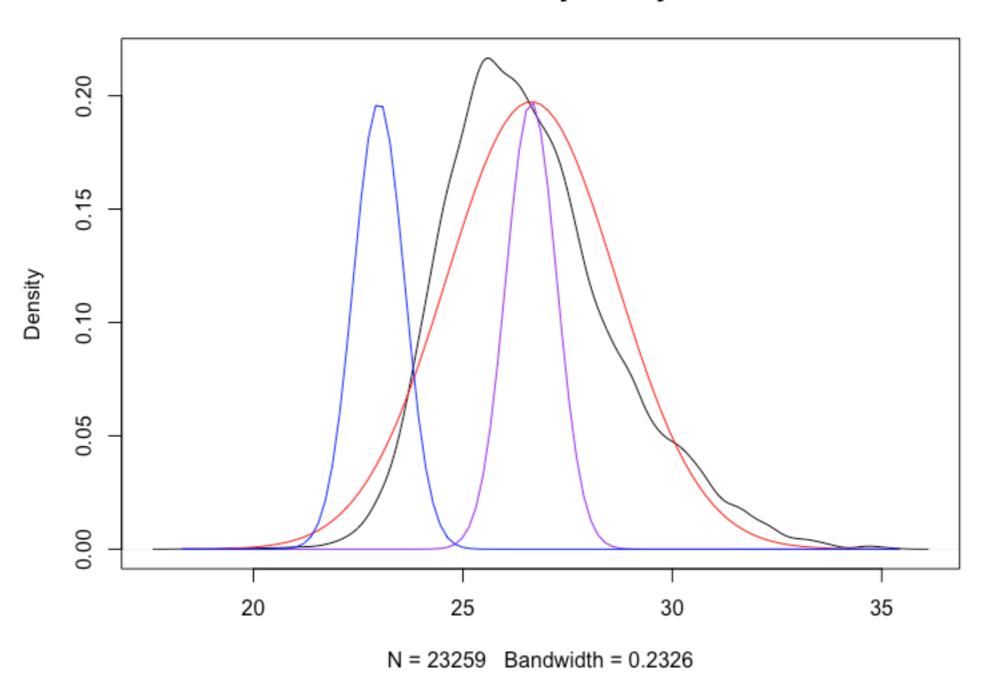


### Make the distribution narrow





## And shift it downwards

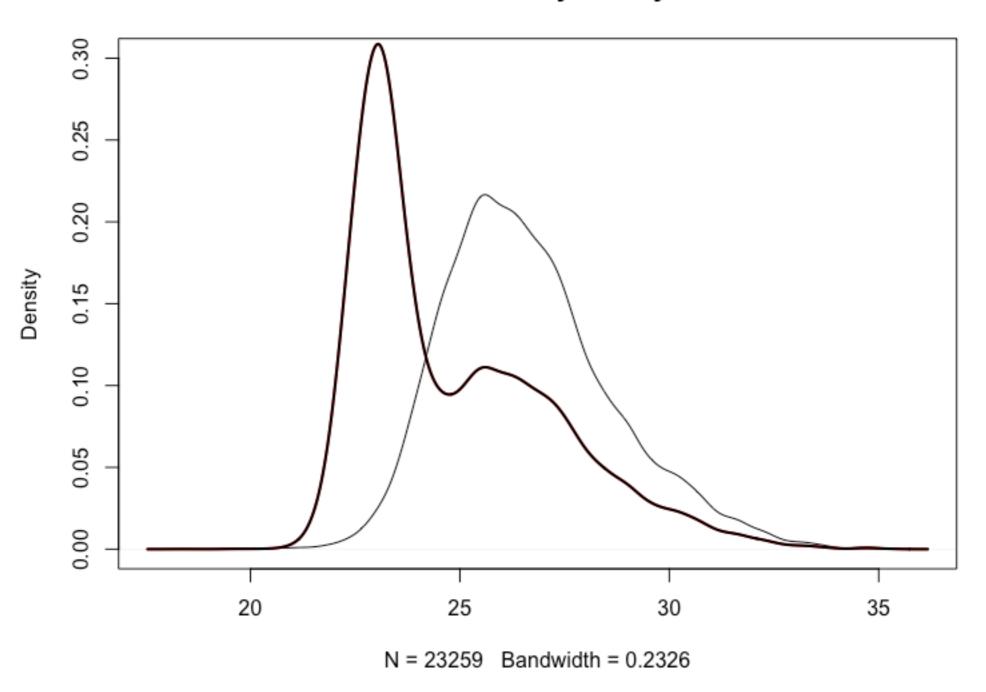




# The new distribution models low intensity values

## Take values from the new distribution



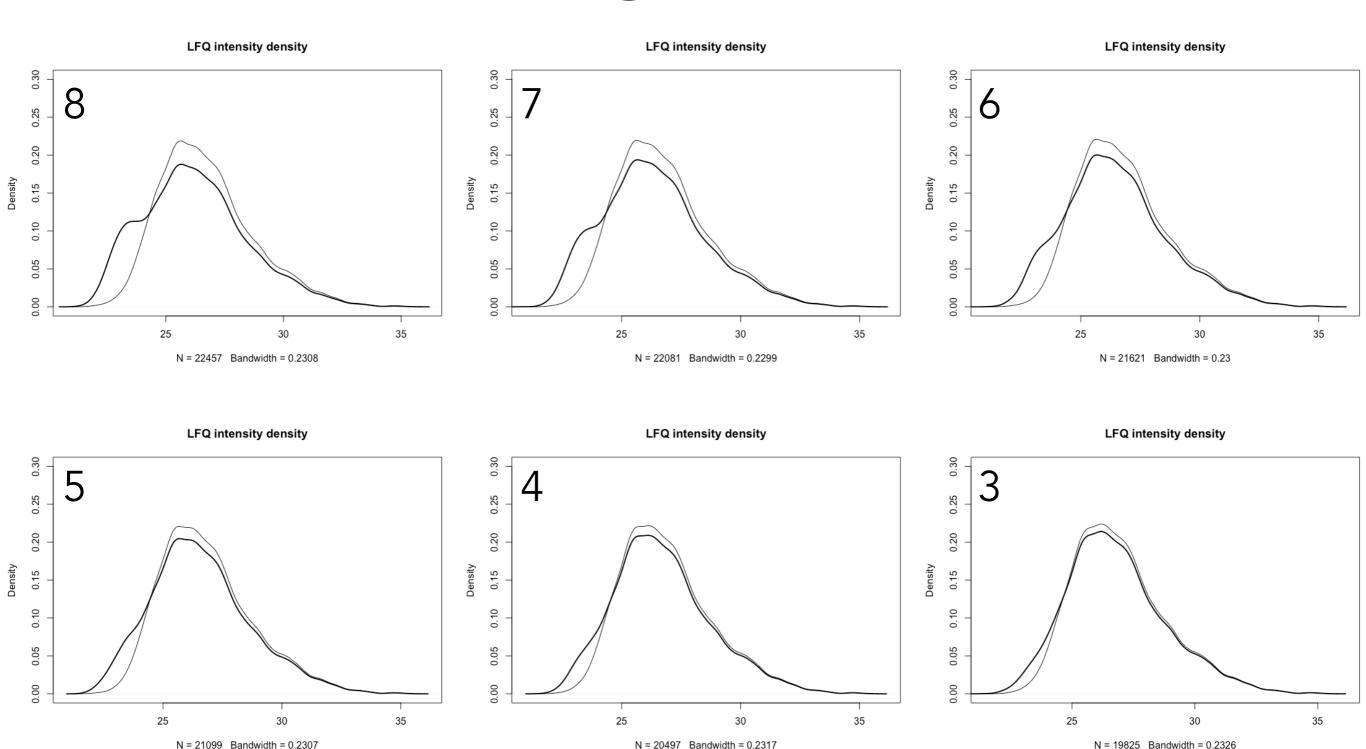




## Prefilter your data!









## Stay in touch with your raw data!



## Imputation guidelines

...according to my experiences

- 1 Try different parameters (filters, fittings).
- Always keep the raw data in mind.
  Especially when interpreting your data.
- Plot, plot, plot.



## Now let's do it in R!

## Your tasks



Split in groups of two

Walk through the code for missing value imputation (impute.R)

Try to understand the lines

modify the filtering for number NAs allowed modify the width of the new distribution

https://github.com/higsch/bioinfo-workshop



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