



LABORATOIRE
DE SCIENCES JUDICIAIRES
ET DE MÉDECINE LÉGALE



Using R in forensic toxicology

Your automation Swiss knife

Brigitte Desharnais, Ph.D.

SOFT 2022

Monday, October 31st 2022

From raw ingredients to beautiful dishes



Data:

- Literature
- Case info
- Instruments output
- Toxicology reports



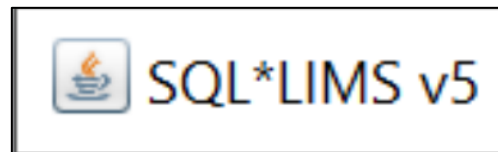
Outcome:

- Summaries to stakeholders
- Informative tools
- Actionable insight

Welcome to your data analysis kitchen!



Dann McKenzie, Flickr



Welcome to your data analysis kitchen!



Peter Salerno, HGTV



solution:



Programming language

<https://www.r-project.org/>



Integrated
development
environment (IDE)

<https://www.rstudio.com/>

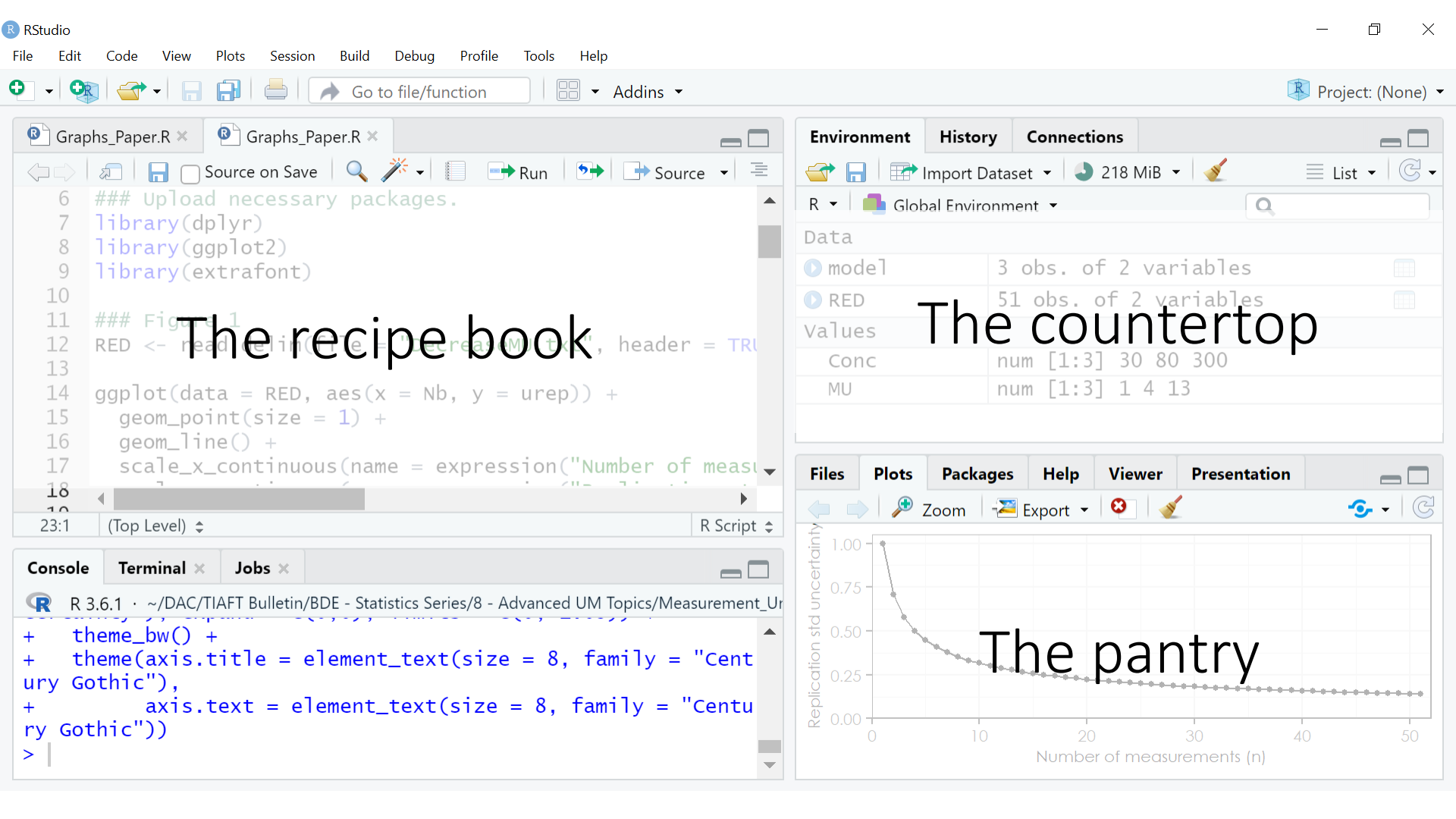


Disclaimer

I am by no means an R absolutist.

I am a pragmatist – I use what does the job quick & easy.

... it just so happens that this is often R!



Packages: fancy appliances in your kitchen



 quarto®

 plotly

Importation

Wrangling (structure, clean)

Visualization

Reporting


A good cook can start from any ingredient...

(Data importation)

Text files

- Available in base R

```
1 Data <- read.delim("C:/Users/DESB02/RDev/Cook.txt",
2                     header = TRUE,
3                     sep = "\t")
```

| Environment | History | Connections | |
|--|-----------------------|-------------|---|
| Data | | | |
| ▼ Data | 8 obs. of 2 variables | |  |
| \$ Ingredients: Factor w/ 8 levels "Bak... | | | |
| \$ Amounts : int 1 1 1 2 2 1 2 1 | | | |

NB. Can also use "clipboard" instead of a file name!

| Cook.txt - Bloc-notes | |
|-----------------------|-----------|
| Fichier | Edition |
| Format | Affichage |
| Aide | |
| Ingredients | Amounts |
| Butter | 1 |
| Sugar | 1 |
| BrownSugar | 1 |
| Eggs | 2 |
| Flour | 2 |
| BakingSoda | 1 |
| Chocolate | 2 |
| Zucchini | 1 |

(Data importation)

Text files

- Of course aim is more complex files!
 - E.g., exported instrumental output (Sciex)

2018-05-18 (E) .txt - Blob notes

| Polaris E8 (E) .txt - Blob notes | | | | | | | | | | | | | | | | | | | | | |
|-----------------------------------|--------------|------------|---------------------|---------------|--------------|---------------------|-------------|-------------------|---------------|--------------|------------------|---------------------|-----------------------|-----------|-----------------|-----------|-----------------|----------------|----------------|-----|--|
| 2018-05-18 Edition: Affixage Aide | | | | | | | | | | | | | | | | | | | | | |
| Rate | Failed Query | % Is Valid | Query other Reasons | Fees Per Cent | % Peak Count | Affix Concentration | % Is Actual | N/A Concentration | Concentration | Expected #T | % Is Expected | #T Integration Type | % Is Integration Type | Area Rate | Attention Time | % Is Name | % Is Area | Corrected Area | % Is | | |
| false | Accuracy | 0 | 1 | 0 | 0.4 | 0.3 | N/A | Baseline | N/A | Loraprom_24 | 295436.920999999 | N/A | 295436.920999999 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.6 | 0.3 | N/A | Baseline | N/A | Loraprom_24 | 295436.920999999 | N/A | 295436.920999999 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.4 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
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| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
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| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
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| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
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| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
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| false | Accuracy | 0 | 1 | 0 | 0.3 | 0.3 | N/A | Baseline | N/A | Amiriprom_24 | 330901.405000000 | N/A | 330901.405000000 | N/A | 99702.738028209 | N/A | 99702.738028209 | N/A | 2.964933767927 | N/A | |
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| | | | | | | | | | | | | | | | | | | | | | |

A good cook can start from any ingredient...

(Data importation)

Spreadsheets

- Available in base R
- Muuuuuuuch better with packages

| File format | Package |
|--------------|---------------|
| .csv | readr |
| .xls(x) | readxl |
| Google Sheet | googlesheets4 |

- E.g.: databases,
exported instrumental output (Agilent)

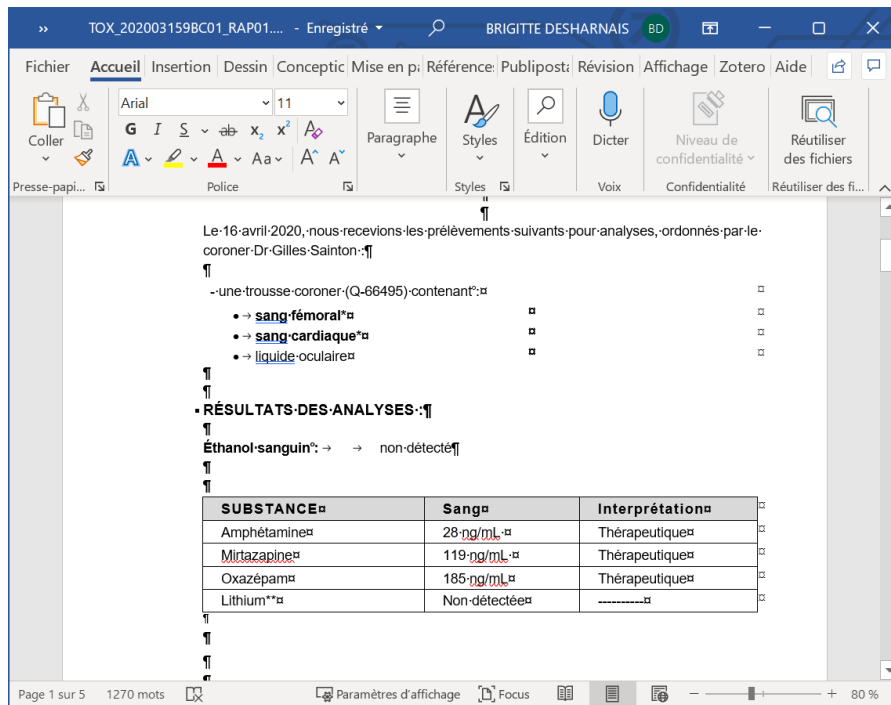


A good cook can start from any ingredient...

(Data importation)

Text based documents

- File type: .doc(x), .pdf, .html
- Packages required
 - readtext
 - docxtractr
- E.g.: toxicology reports, literature, systems output



... transform it however they want it...

(Data wrangling)

- Hands down, packages:
 - **dplyr** – select, filter, arrange, group...

Get the
appliance out
of the pantry

```
1 library(dplyr)
2 Data <- as_tibble(Data)
3 Dessert <- Data %>% filter(Ingredients != "Zucchini")
```



| | Ingredients | Amounts |
|---|-------------|---------|
| 1 | Butter | 1 |
| 2 | Sugar | 1 |
| 3 | BrownSugar | 1 |
| 4 | Eggs | 2 |
| 5 | Flour | 2 |
| 6 | BakingSoda | 1 |
| 7 | Chocolate | 2 |
| 8 | Zucchini | 1 |

| | Ingredients | Amounts |
|---|-------------|---------|
| 1 | Butter | 1 |
| 2 | Sugar | 1 |
| 3 | BrownSugar | 1 |
| 4 | Eggs | 2 |
| 5 | Flour | 2 |
| 6 | BakingSoda | 1 |
| 7 | Chocolate | 2 |

... transform it however they want it...

(Data wrangling)

- Hands down, packages:
 - **dplyr** – select, filter, arrange, group...
 - **tidyr** – gather, spread, separate, unite...

```
1 library(tidyr)
2 Cleaned <- Results %>% pivot_wider(names_from = Matrix,
3                                     values_from = Concentration)
```

| | File | Matrix | Analyte | Concentration |
|---|--------|--------|-------------|---------------|
| 1 | 2022-A | CB | Amphetamine | 20 |
| 2 | 2022-A | FB | Amphetamine | 19 |
| 3 | 2022-A | UR | Amphetamine | 39 |
| 4 | 2022-B | CB | Amphetamine | 32 |
| 5 | 2022-B | VH | Amphetamine | 24 |
| 6 | 2022-B | GC | Amphetamine | 5 |
| 7 | 2022-B | UR | Amphetamine | 36 |
| 8 | 2022-C | CB | Ethanol | 22 |
| 9 | 2022-C | FB | Ethanol | 29 |

| | File | Analyte | CB | FB | UR | VH | GC |
|---|--------|-------------|----|----|----|----|----|
| 1 | 2022-A | Amphetamine | 20 | 19 | 39 | NA | NA |
| 2 | 2022-B | Amphetamine | 32 | NA | 36 | 24 | 5 |
| 3 | 2022-C | Ethanol | 22 | 29 | NA | NA | NA |

(And you can use dplyr to filter – on the same line if you want!)

... transform it however they want it...

(Data wrangling)

- Hands down, packages:
 - **dplyr** – select, filter, arrange, group...
 - **tidyr** – gather, spread, separate, unite...
 - **stringr** – search & modify character strings (text)
 - **lubridate** – work with time and dates
 - **janitor** – cleaning names, empty column/lines

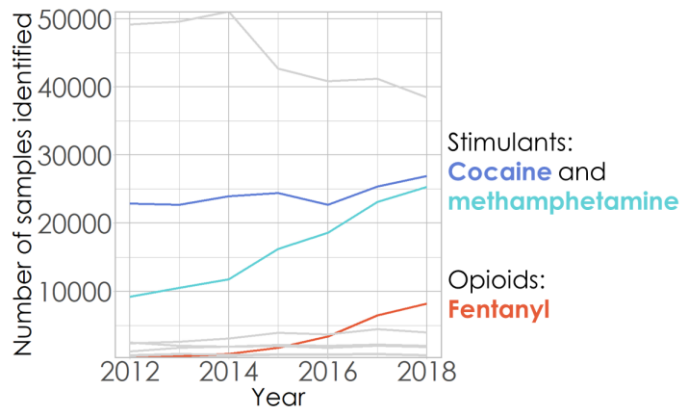


... to create beautiful and delicious plates!

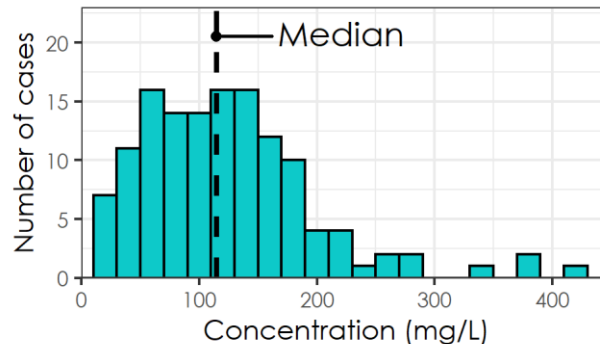
(Visualization & reporting)



- For graphs, ggplot2 & associated packages:
 - extrafont – use alternative fonts for labelling
 - export – send & manipulate in e.g. Power Point (!!!)
 - ganimate – animated plots
 - plotly – graphs that can be manipulated (zoom, save, etc.)



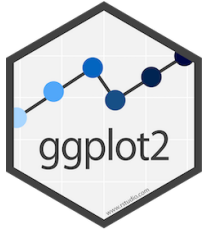
TIAFT Bulletin (2019) 49 (4), 14-26



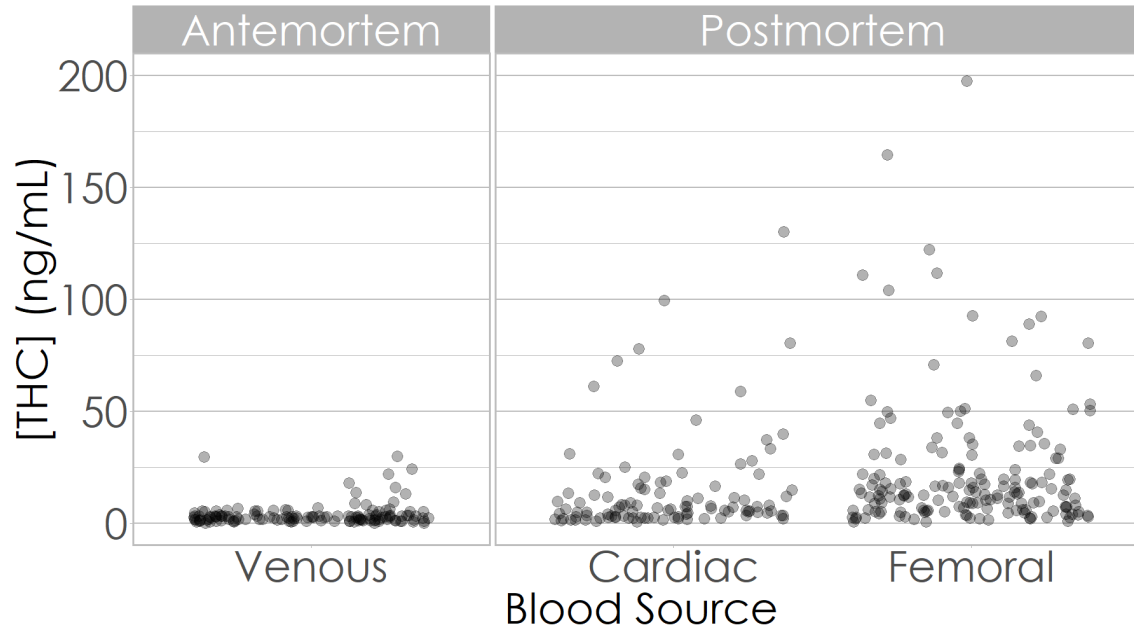
TIAFT Bulletin (2020) 50 (4), 11-19

... to create beautiful and delicious plates!

(Visualization & reporting)



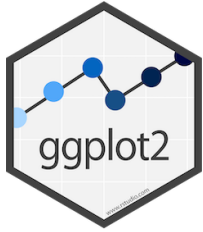
■ Faceted plots



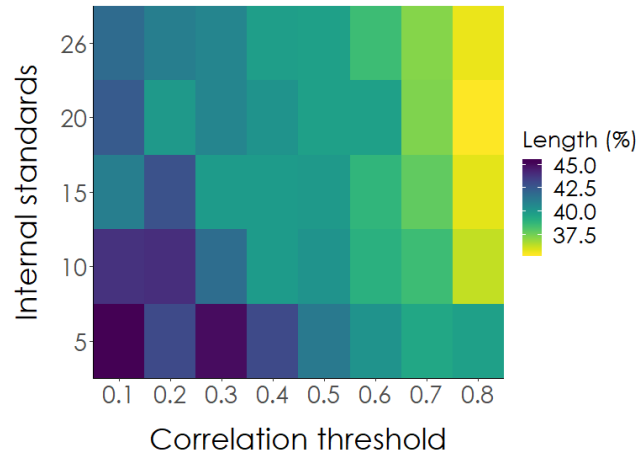
TIAFT Bulletin (2019) 49 (4), 14-26

... to create beautiful and delicious plates!

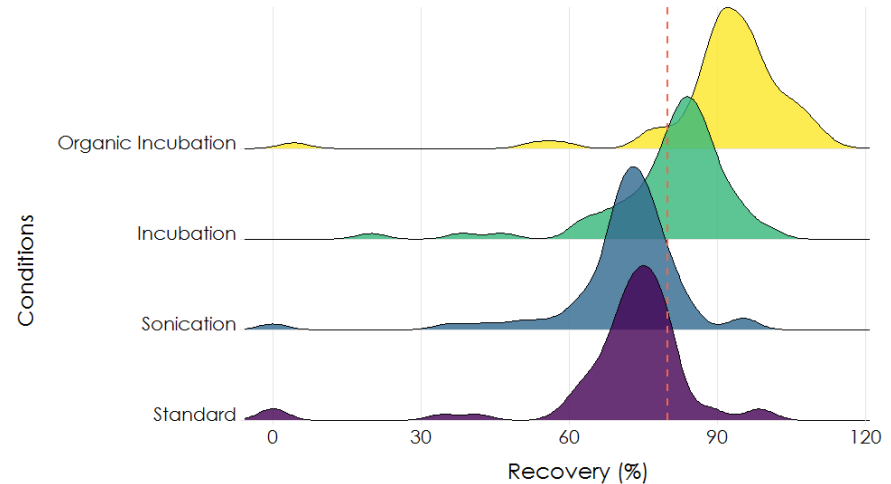
(Visualization & reporting)



- Heat maps
- Density
- Ridges



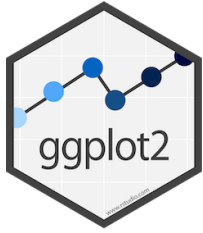
TIAFT 2019 Annual Meeting



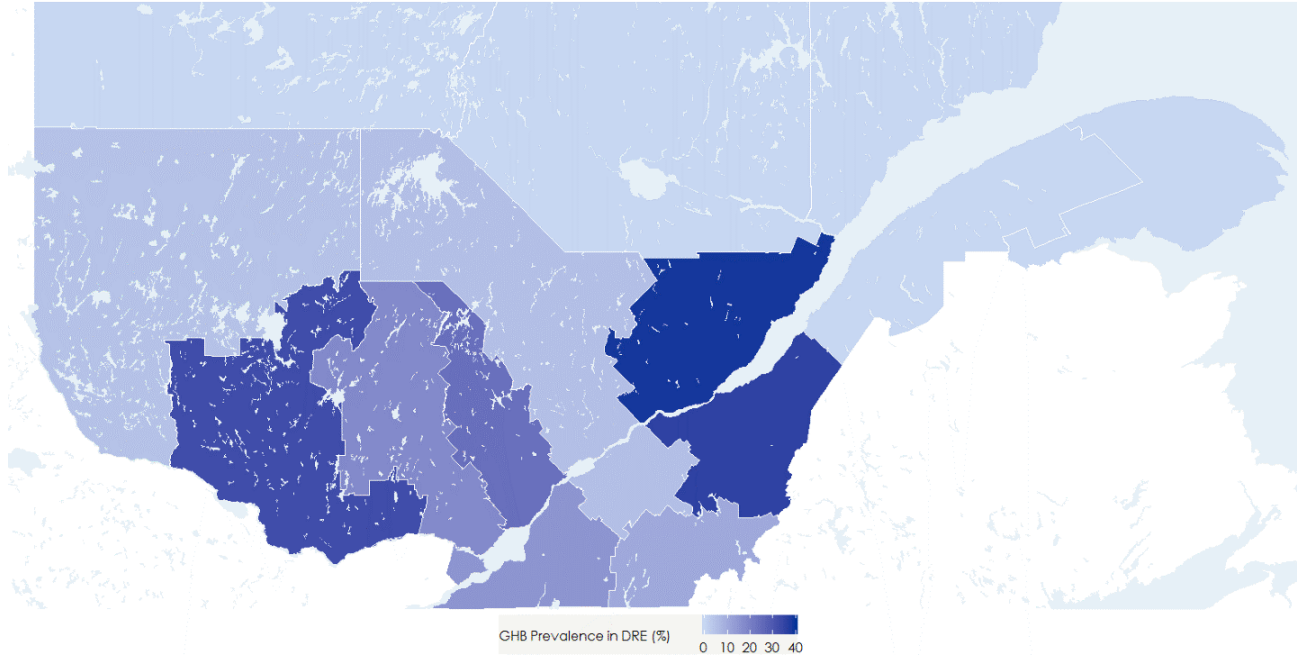
Forensic Sci Int (2020) 317, 110506

... to create beautiful and delicious plates!

(Visualization & reporting)



■ Maps



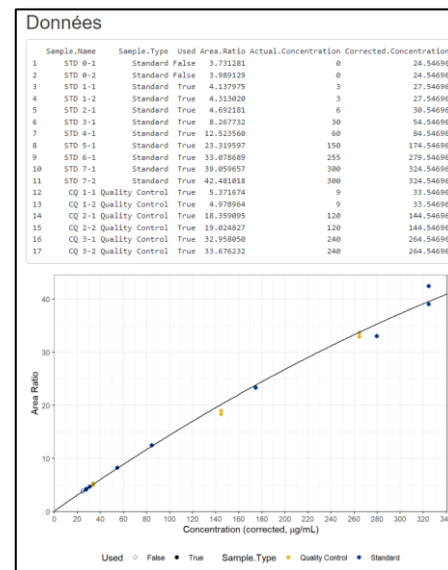
SOFT 2019 Annual Meeting

... to create beautiful and delicious plates!

(Visualization & reporting)



- For standardized reports:
 - rmarkdown – generate .html or .pdf reports
 - quarto – new generation rmarkdown, includes WYSIWYG editor



... to create beautiful and delicious plates!

(Visualization & reporting)



- For interactive reports, dashboards:
 - shiny – the foundation
 - shinydashboard
 - bs4dash
 - flexdashboard
 - ...
- Interactive elements can be integrated in reports!
- Examples at <https://shiny.rstudio.com/gallery/>



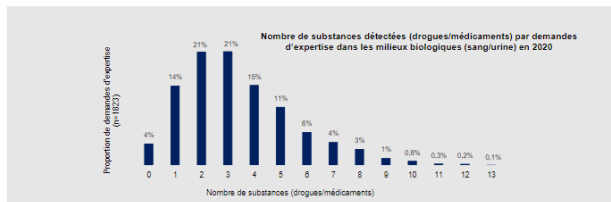
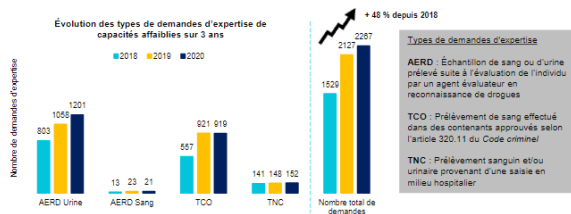
A five course meal

Applications of R in our laboratory

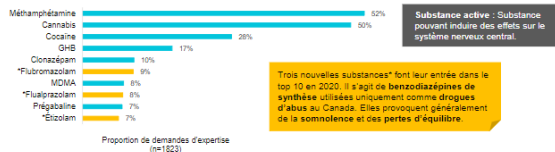
Hors d'oeuvre: annual summary reports



Statistiques sur les demandes d'expertise en matière de capacités affaiblies reçues au LSJML en 2020



Top 10 des substances actives (drogues/médicaments) retrouvées dans les demandes d'expertise de capacités affaiblies en 2020



© LSJML 2022 Les statistiques présentées dans ce document ont été compilées du 1^{er} janvier au 31 décembre de chaque année.



Statistiques sur les dossiers post mortem reçus au LSJML en toxicologie

Les dossiers post mortem et la toxicologie

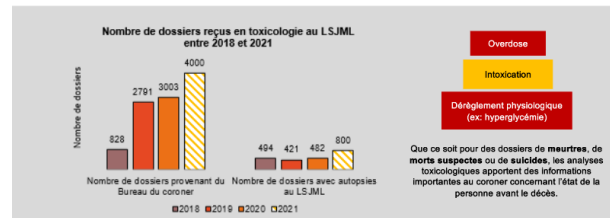
Chaque année, le département de toxicologie du LSJML reçoit environ **6000 dossiers**, dont plus de la moitié (~55%) sont des dossiers post mortem.

Les analyses toxicologiques sont effectuées pour mettre en évidence la présence d'**alcool**, de **médicaments**, de **drogues** et/ou de **poisons**. Ces demandes d'analyses proviennent du Bureau du coronar.

Les résultats toxicologiques peuvent aider à établir ou à éclaircir les causes et les circonstances des décès qui surviennent au Québec.



Environ 15 % des prélèvements reçus en toxicologie proviennent d'autopsies réalisées au LSJML.



Que ce soit pour des dossiers de **meurtres**, de **morts suspectes** ou de **suicides**, les analyses toxicologiques apportent des informations importantes au coronar concernant l'état de la personne avant le décès.

Les types de prélèvements reçus

Dans la majorité des dossiers :

- Sang fémoral
- Sang cardiaque
- Liquide oculaire
- Urine

Moins fréquemment (car nécessite une autopsie complète) :

- Contenu gastrique
- Foie et/ou bile
- Échantillons de poulmon, de cerveau



© LSJML 2022 Les statistiques présentées dans ce document ont été compilées de 2018 à 2021.

Appetizer: population distributions

GHB - Étude de population

AUTHOR

Brigitte Desharnais, Laboratoire de sciences
judiciaires et de médecine légale

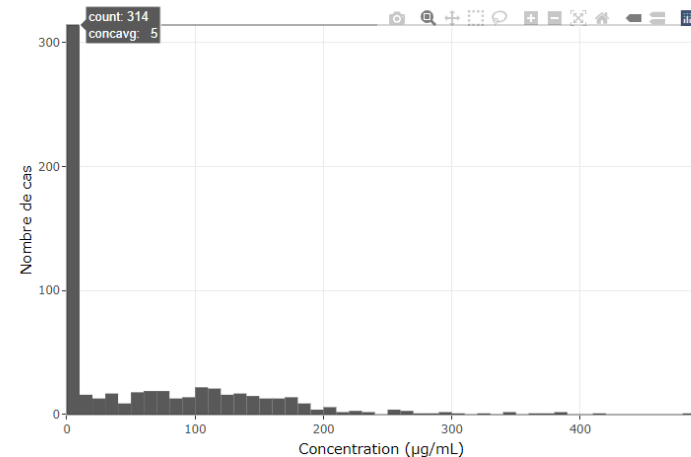
Données entre le 2017-01-22 et le 2022-07-22 (date de batch).

Répartition des données

| Milieu | Ante mortem | Post mortem |
|-----------------------|-------------|-------------|
| Sang veineux (SV) | 630 | 66 |
| Sang fémoral (SF) | 6 | 9788 |
| Sang cardiaque (SC) | 0 | 2379 |
| Plasma/Sérum (PS) | 81 | 21 |
| Urine (UR) | 715 | 1314 |
| Liquide oculaire (LO) | 0 | 458 |

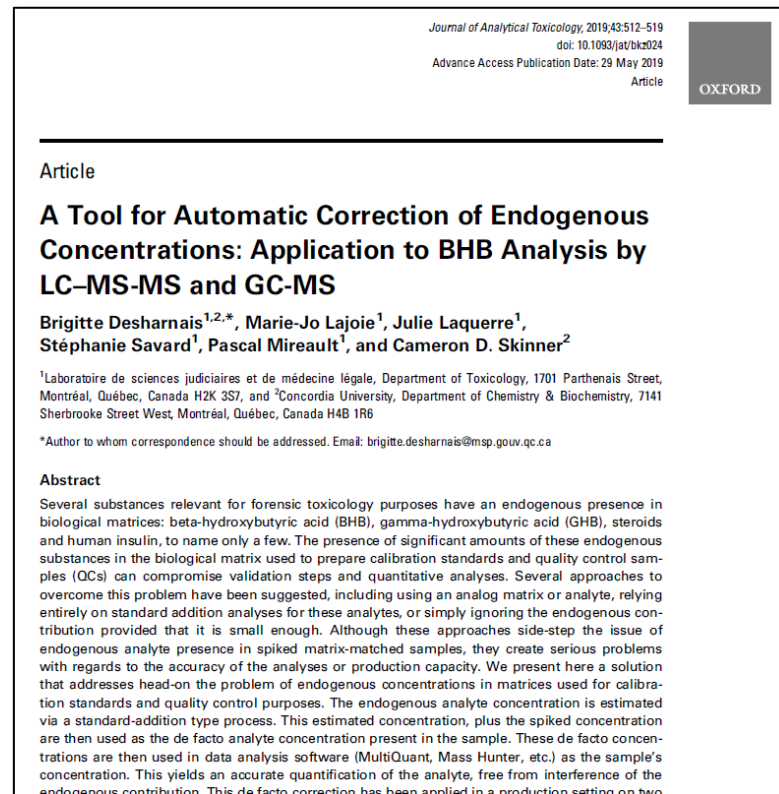
Ante mortem

Sang veineux

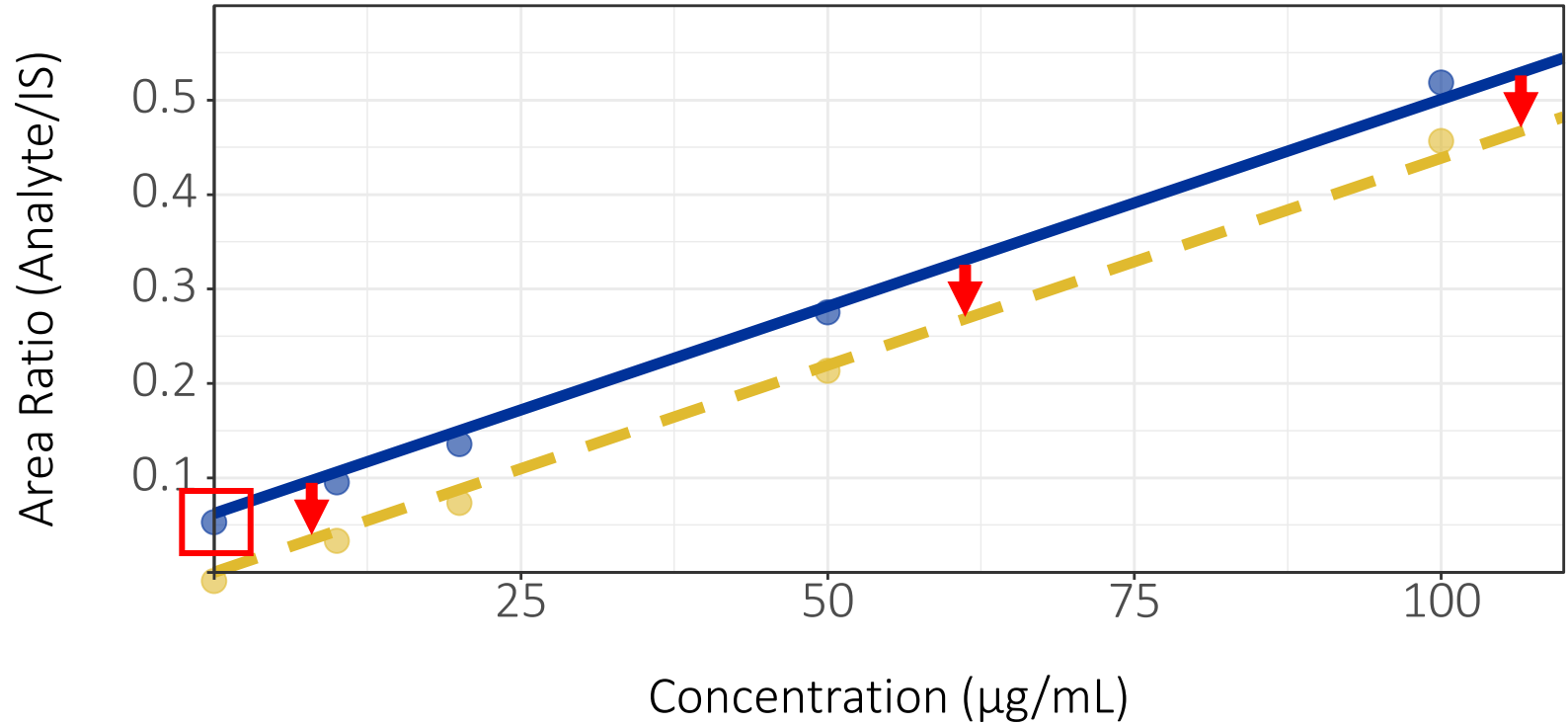


Salad: correction of endogenous content in spiking matrix

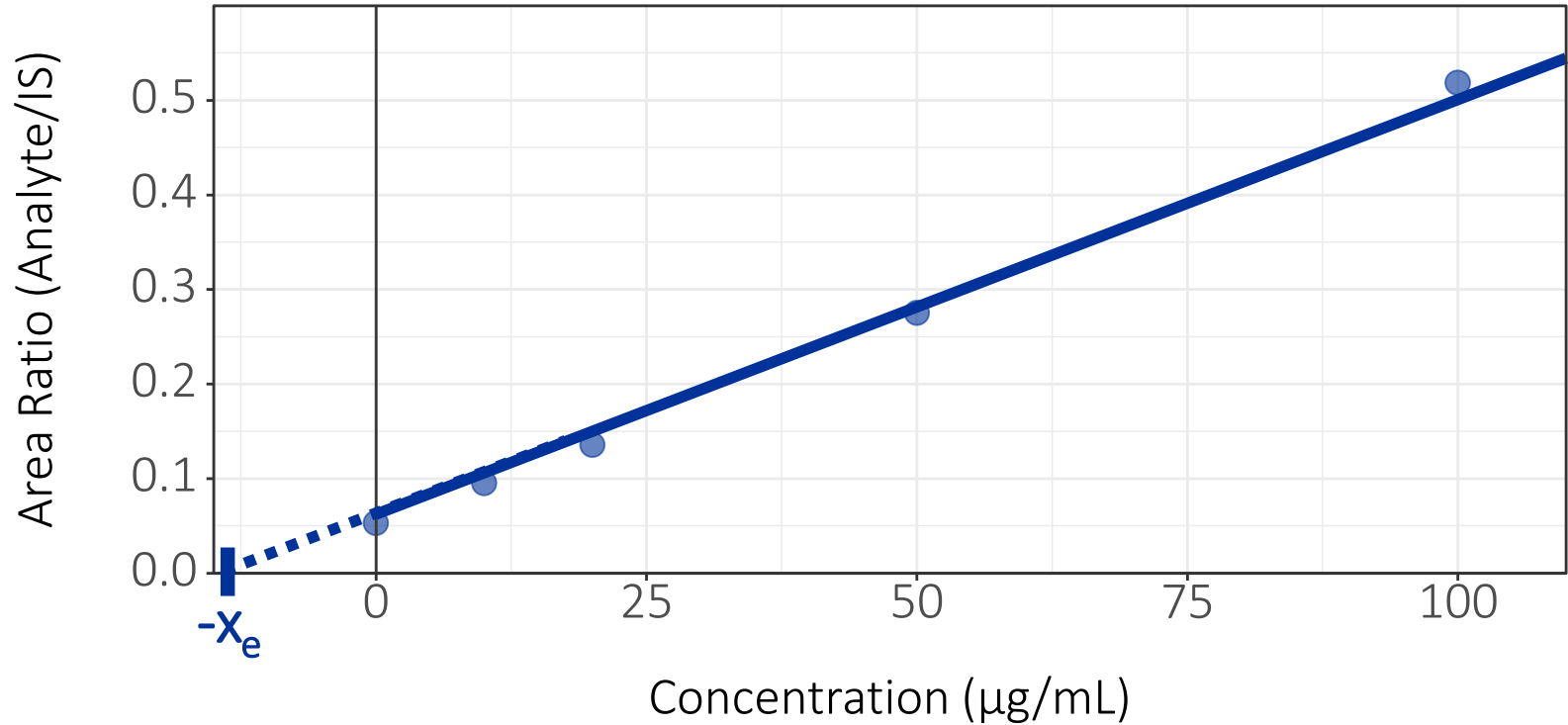
- Endogenous analytes create a validation challenge with regards to spiking matrix
- Possible solutions include
 - Standard addition
 - Surrogate matrix
 - Surrogate analyte
 - Ignoring the problem (!)



Salad: correction of endogenous content in spiking matrix

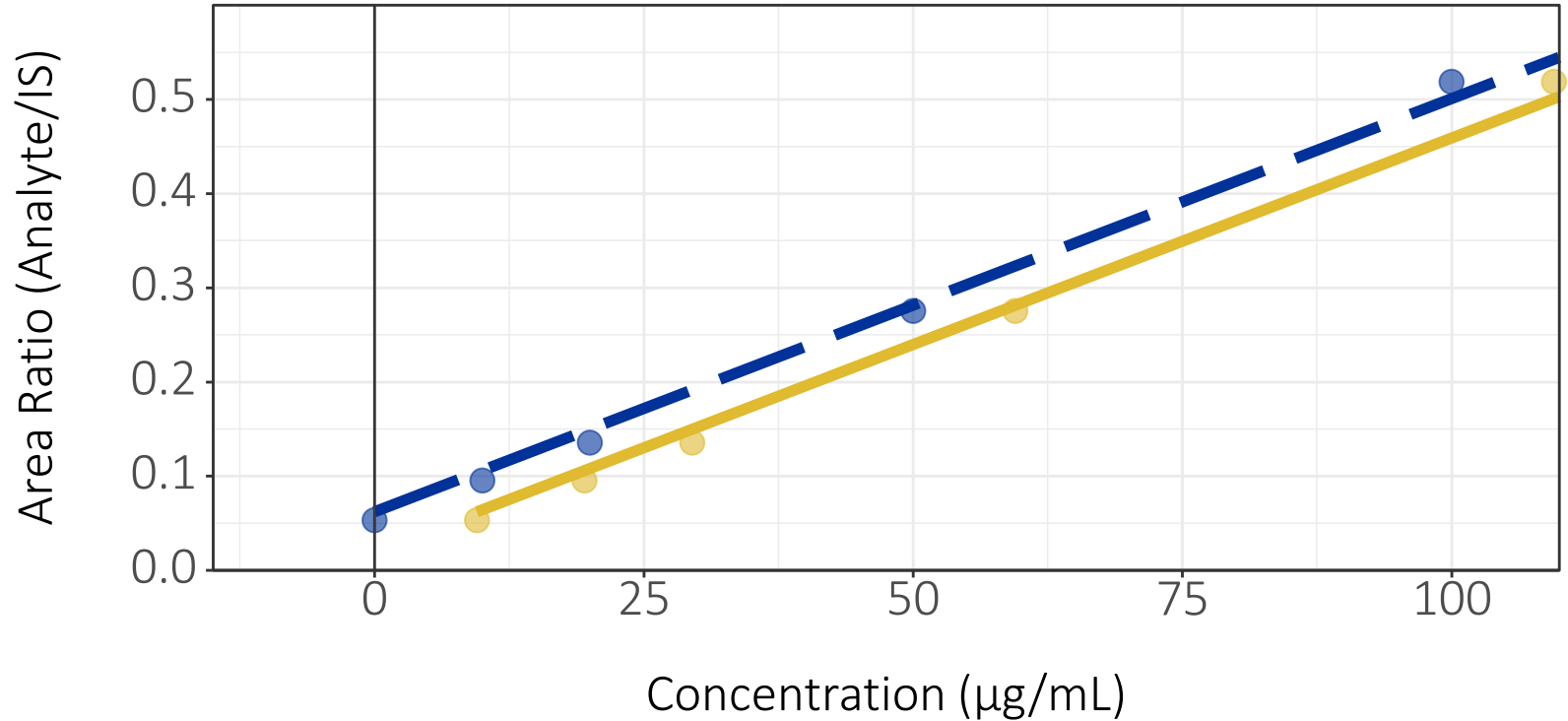


Salad: correction of endogenous content in spiking matrix



Correction: $x_r = x_s + x_e$

Salad: correction of endogenous content in spiking matrix

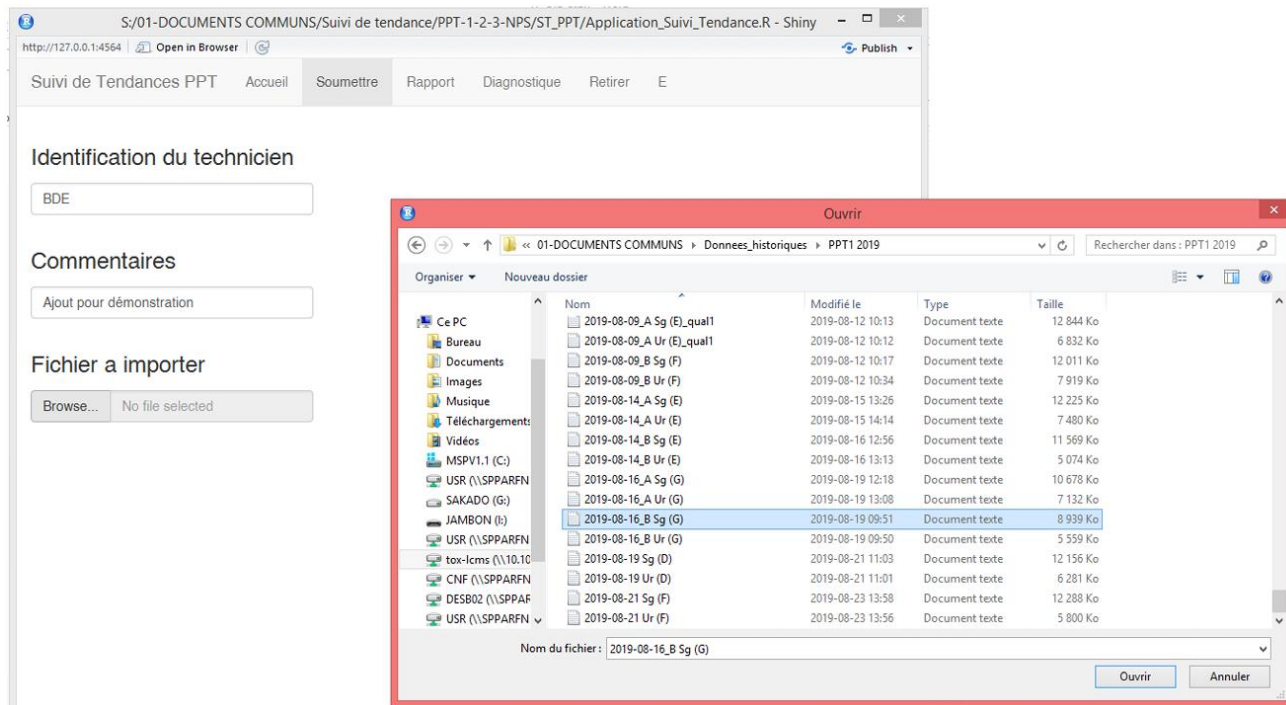


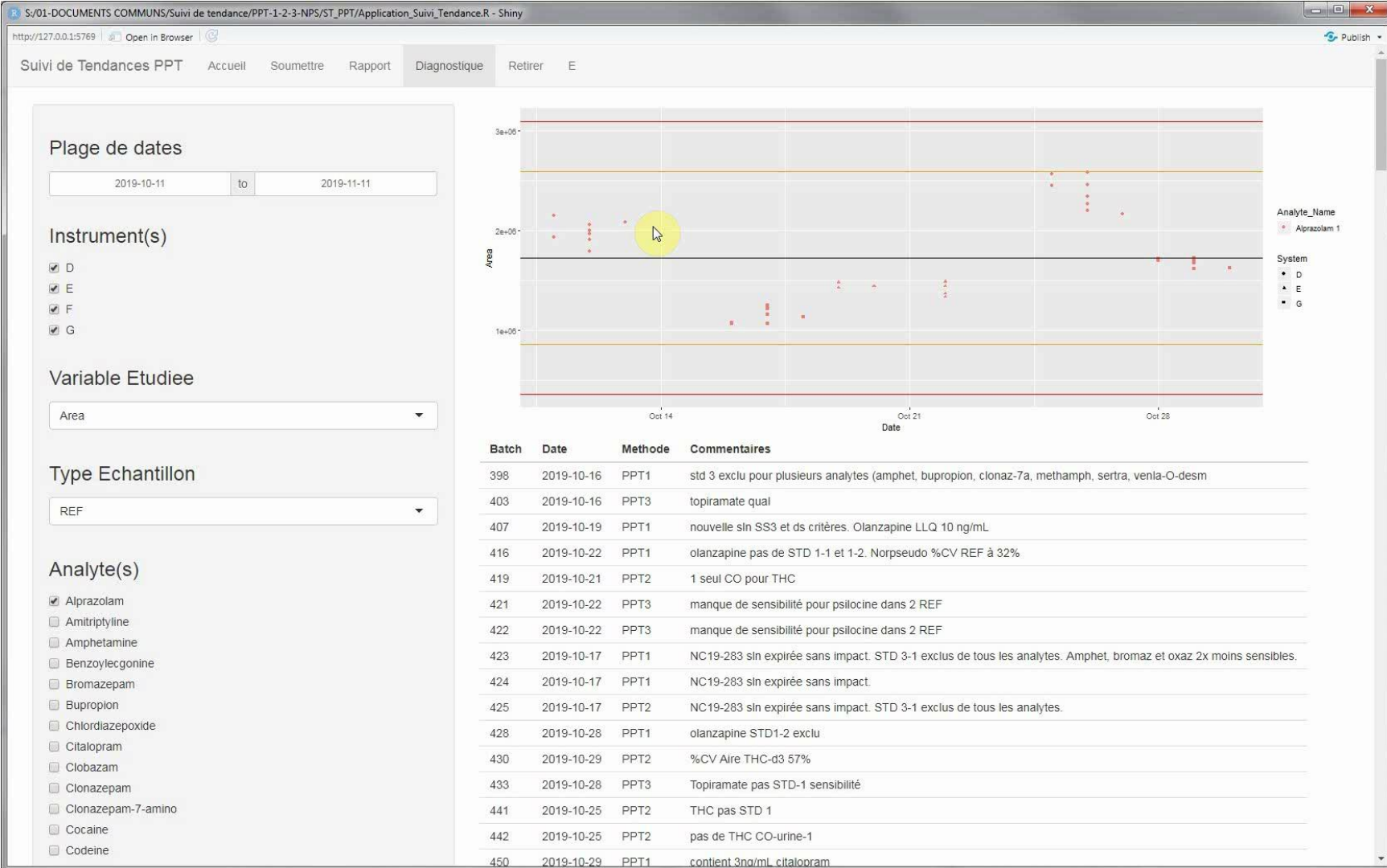
Correction: $X_r = X_s + X_e$

| Index | Sample Name | Sample Type | IS | Component Name | IS Name | Actual Concentration | Area | IS Area | Area Ratio | Retention Time | IS Reten. Time | Width at 50% | Used | Calculated Concentration | Accuracy |
|-------|-------------|---------------|--------------------------|----------------|---------|----------------------|---------|---------|------------|----------------|----------------|--------------|-------------------------------------|--------------------------|----------|
| 486 | STD 0-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 0.000 | 1.769e4 | 3.081e5 | 0.0574 | 0.98 | 0.80 | 0.04 | <input type="checkbox"/> | 0.199 | N/A |
| 559 | STD 0-2 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 0.000 | 1.649e4 | 3.001e5 | 0.0549 | 1.00 | 0.81 | 0.03 | <input type="checkbox"/> | 0.164 | N/A |
| 632 | STD 1-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 3.000 | 7.901e4 | 2.971e5 | 0.2659 | 0.99 | 0.80 | 0.04 | <input checked="" type="checkbox"/> | 3.124 | 104.14 |
| 705 | STD 1-2 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 3.000 | 7.413e4 | 2.899e5 | 0.2557 | 0.99 | 0.80 | 0.04 | <input checked="" type="checkbox"/> | 2.981 | 99.36 |
| 778 | STD 2-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 6.000 | 1.256e5 | 2.865e5 | 0.4386 | 0.99 | 0.81 | 0.04 | <input checked="" type="checkbox"/> | 5.546 | 92.43 |
| 851 | STD 3-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 30.000 | 6.798e5 | 3.081e5 | 2.2063 | 0.99 | 0.80 | 0.04 | <input checked="" type="checkbox"/> | 30.280 | 100.93 |
| 924 | STD 4-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 60.000 | 1.291e6 | 2.859e5 | 4.5144 | 0.99 | 0.81 | 0.04 | <input checked="" type="checkbox"/> | 62.433 | 104.06 |
| 997 | STD 5-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 150.000 | 3.154e6 | 2.936e5 | 10.7398 | 0.99 | 0.81 | 0.04 | <input checked="" type="checkbox"/> | 148.353 | 98.90 |
| 1070 | STD 6-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 255.000 | 5.336e6 | 2.799e5 | 19.0620 | 0.99 | 0.80 | 0.04 | <input checked="" type="checkbox"/> | 261.456 | 102.53 |
| 1143 | STD 7-1 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 300.000 | 6.075e6 | 2.776e5 | 21.8830 | 0.99 | 0.80 | 0.04 | <input checked="" type="checkbox"/> | 299.354 | 99.78 |
| 1216 | STD 7-2 | Standard | <input type="checkbox"/> | BHB 1 | GHB_d6 | 300.000 | 5.600e6 | 2.611e5 | 21.4504 | 0.99 | 0.81 | 0.04 | <input checked="" type="checkbox"/> | 293.556 | 97.85 |
| 1289 | CQ 1-1 | Quality Co... | <input type="checkbox"/> | BHB 1 | GHB_d6 | 9.000 | 1.894e5 | 2.868e5 | 0.6604 | 0.99 | 0.81 | 0.03 | <input checked="" type="checkbox"/> | 8.654 | 96.16 |
| 1362 | CQ 1-2 | Quality Co... | <input type="checkbox"/> | BHB 1 | GHB_d6 | 9.000 | 1.785e5 | 2.795e5 | 0.6384 | 1.00 | 0.81 | 0.04 | <input checked="" type="checkbox"/> | 8.347 | 92.74 |
| 1435 | CQ 2-1 | Quality Co... | <input type="checkbox"/> | BHB 1 | GHB_d6 | 120.000 | 2.810e6 | 2.948e5 | 9.5323 | 0.99 | 0.80 | 0.04 | <input checked="" type="checkbox"/> | 131.777 | 109.81 |
| 1508 | CQ 2-2 | Quality Co... | <input type="checkbox"/> | BHB 1 | GHB_d6 | 120.000 | 2.862e6 | 2.960e5 | 9.6700 | 0.99 | 0.81 | 0.04 | <input checked="" type="checkbox"/> | 133.670 | 111.39 |
| 1581 | CQ 3-1 | Quality Co... | <input type="checkbox"/> | BHB 1 | GHB_d6 | 240.000 | 5.450e6 | 3.014e5 | 18.0816 | 0.99 | 0.80 | 0.04 | <input checked="" type="checkbox"/> | 248.233 | 103.43 |
| 1654 | CQ 3-2 | Quality Co... | <input type="checkbox"/> | BHB 1 | GHB_d6 | 240.000 | 5.113e6 | 2.734e5 | 18.7021 | 0.99 | 0.81 | 0.04 | <input checked="" type="checkbox"/> | 256.604 | 106.92 |

Main course: trend monitoring

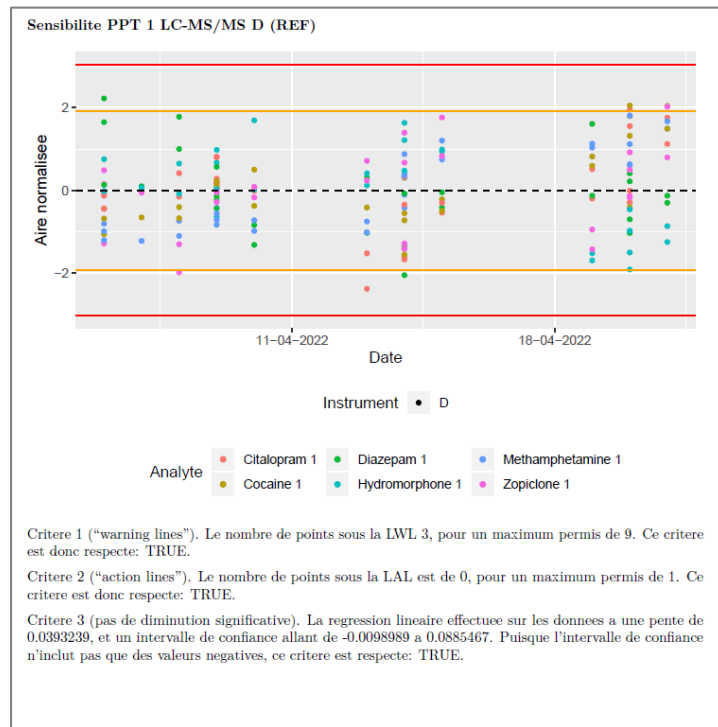
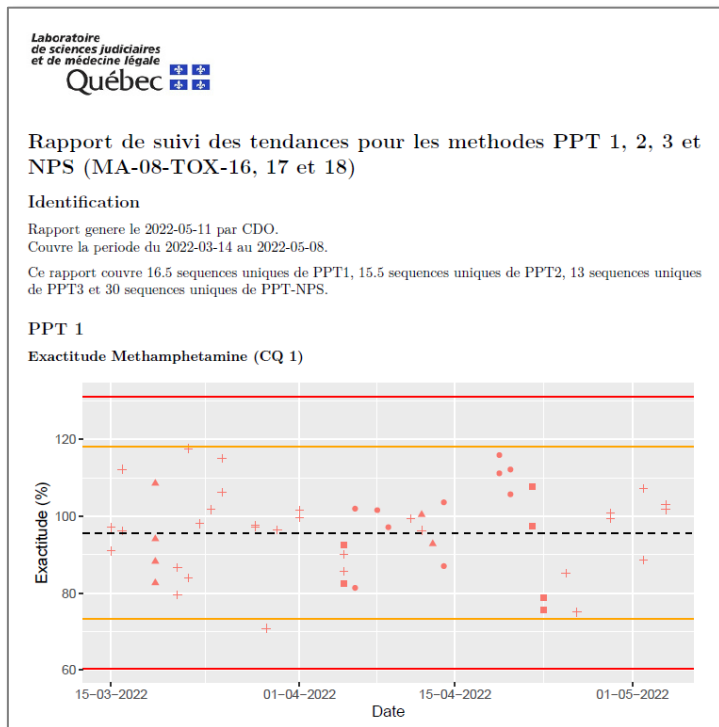
LC-MS/MS method, 3 injections, 125 analytes, 7 STDs, 3x3 QCs...





Main course: trend monitoring

Periodic PDF report for accreditation requirements



Dessert: DUID interactive dashboard

Enregistrement automatique BD_CAPAF_v2-201sm Rechercher BRIGITTE DESHARNAIS

Fichier Accueil Insertion Dessin Mise en page Formules Données Révision Affichage Aide

Calibri 11 A A

G I S A A

Presse-papiers Police Alignement Nombre Styles Cellules Édition Analyse Confidentialité

C3

| Type dossier | Alcoolémie (mg/100ml) | minutes |
|---|---|--|
| Dossier annulé <input type="checkbox"/> Oui <input checked="" type="checkbox"/> Non | | |
| Ville/Région | 1. Substance(s) détectée(s) Conc. (pas d'unité) | 11. Substance(s) détectée(s) Conc. (pas d'unité) |
| Événement Date | 2. | 12. |
| Heure | 3. | 13. |
| Arrestation Date | 4. | 14. |
| Heure | 5. | 15. |
| Genre <input checked="" type="checkbox"/> Homme <input type="checkbox"/> Femme <input type="checkbox"/> Indéterminé Veuillez cocher | 6. | 16. |
| Date de naissance | 7. | 17. |
| Opinion(s) de l'agent 1. | 8. | 18. |
| 2. | 9. | 19. |
| 3. | 10. | 20. |
| 4. | | |
| Aveux/saisie <input type="checkbox"/> Oui <input type="checkbox"/> Non Veuillez cocher | | |
| Commentaires | | |

Entrer les données Entrer un autre prélèvement du même dossier Annuler la saisie

Formulaire BD Consultation Menu_deroulant Modifi_version

Prêt Paramètres d'affichage 80%

Dessert: DLD interactive dashboard

Enregistrement automatique BD_CAPAF_v2-2024.xlsx Rechercher BRIGITTE DESHARNAIS

Fichier Accueil Insertion Dessin Mise en page Formules Données Révision Affichage Aide Conception de la table

Calibri 11 Renvoyer à la ligne automatiquement Mise en forme conditionnelle Mettre sous forme de tableau Styles de cellules Insérer Supprimer Format Recopier Effacer Trieur et sélectionner Analyse de données Niveau de confidentialité

Police Alignement Styles Cellules Édition

A11838 202204634CM01

| | A | X | Y | Z | AA | AB | AC | AD | AE | AF | AG | AH | AI | AJ | |
|------|----------------|-----------------|----------------------------|---------------------|-----------------------|----------------------------|---------------------|--------------------------|----------------------------|---------------------|---------------------------|------------------------------|---------------------|--------------------------|------------|
| | Dossier | Alcoolémie | Substance 1 | Concentrat ion 1 | Type de drogue 1 | Substance 2 | Concentrat ion 2 | Type de drogue 2 | Substance 3 | Concentrat ion 3 | Type de drogue 3 | Substance 4 | Concentrat ion 4 | Type de drogue 4 | Substan |
| 1790 | 202204320CM01 | Non-analyse | Néglig | Détece | Non actif | Diphendramine | Détece | Dépresseur | Nitazépan / métabolite | Détece | Dépresseur | Cyclobenzaprine (métabolite) | Détece | Non actif | Trazodone |
| 1791 | 202203247CM01 | Non-analyse | Diazepam | Détece | Analogue narcotique | Méthamphétamine | Détece | Stimulant | MDMA | Détece | Stimulant | MDA | Détece | Stimulant | Pseudo |
| 1792 | 2022032675CM01 | Non-analyse | Cocaine | Détece | Stimulant | Cocaine | Détece | Stimulant | Méthamphétamine | Détece | Stimulant | Morphine (métabolite) | Détece | Non actif | |
| 1793 | 202204340MT01 | Non-analyse | Fluorazepam | Détece | Stimulant | Fluorazepam | Détece | Stimulant | Acétaminophène | Détece | Non actif | Fluorazepam | Détece | Dépresseur | |
| 1794 | 202205019MT01 | Non-analyse | Cocaine | Détece | Stimulant | Méthamphétamine | Détece | Stimulant | MDMA | Détece | Stimulant | MDA | Détece | Stimulant | |
| 1795 | 202204123CM01 | Néglig | Cocaine | 81 | Stimulant | Méthamphétamine | 311 | Stimulant | MDMA | 6.2 | Dépresseur | MDA | Détece | Dépresseur | |
| 1796 | 2022050195CM01 | Non-analyse | | | | | | | | | | | | | |
| 1797 | 2022057335CM01 | 108 | | | | | | | | | | | | | |
| 1798 | 2022057305CM01 | 48 | | | | | | | | | | | | | |
| 1799 | 2022055685CM01 | 23 | | | | | | | | | | | | | |
| 1790 | 2022055685CM01 | 54 | | | | | | | | | | | | | |
| 1771 | 2022055685CM01 | 60 | | | | | | | | | | | | | |
| 1772 | 202205226CM01 | GHB | Détece | Dépresseur | Diazepam (métabolite) | Détece | Dépresseur | Méthamphétamine | Détece | Stimulant | | Roloxylidine | Détece | Anesthésique dissociatif | |
| 1773 | 202205186CM01 | THC | Détece | Cannabis | | | | | | | | | | | |
| 1774 | 2022052045CM01 | GHB | 77 | Dépresseur | Méthamphétamine | 98 | Stimulant | Cocaine (métabolite) | Détece | Non actif | | | | | |
| 1775 | 202204173MT01 | Fluorazepam | Détece | Dépresseur | Chlorphéniramine | Détece | Dépresseur | MDMA | Détece | Analogue narcotique | Cocaine | Détece | Stimulant | Pseudo | |
| 1776 | 2022041729CM01 | Cocaine | Détece | Stimulant | Méthamphétamine | Détece | Stimulant | MDMA | Détece | Stimulant | THC (métabolite THC-COOH) | Détece | Stimulant | Cannabis | |
| 1777 | 202203825CM01 | Méthamphétamine | Détece | Stimulant | Cocaine (métabolite) | Détece | Non actif | | Détece | Non actif | | | | | |
| 1778 | 2022048635CM01 | Non-analyse | Cocaine | Détece | Stimulant | Méthamphétamine | Détece | Stimulant | | | | | | | |
| 1779 | 2022037955CM01 | Non-analyse | | | | | | | | | | | | | |
| 1780 | 202203825CM01 | Non-analyse | | | | | | | | | | | | | |
| 1781 | 202205252CM01 | Non-analyse | THC (métabolite THC-COOH) | Détece | Cannabis | Cocaine | Détece | Stimulant | Bupropion | Détece | Dépresseur | Citalopram / Escitalopram | Détece | Dépresseur | Duloxetine |
| 1782 | 202203825CM01 | 62 | Midazolam | 26 | Dépresseur | Cocaine (métabolite) | Détece | Non actif | | | | | | | |
| 1783 | 2022051865CM01 | Non-analyse | THC (Tétrahydrocannabinol) | 2.8 | Cannabis | MDMA | 22 | Stimulant | Lamotrigine | 8.1 | Dépresseur | Citalopram / Escitalopram | Détece | Dépresseur | |
| 1784 | 20220330MT01 | Non-analyse | GHB | 224 | Dépresseur | MDMA | 22 | Stimulant | THC (Tétrahydrocannabinol) | 1.5 | Cannabis | Citalopram / Escitalopram | 111 | Dépresseur | Cocaine |
| 1785 | 202205104CM01 | Non-analyse | Cocaine | 108 | Stimulant | Méthamphétamine | 20 | Stimulant | | | | | | | |
| 1786 | 2022051155CM01 | Néglig | | | | | | | | | | | | | |
| 1787 | 2022050195CM01 | Non-analyse | | | | | | | | | | | | | |
| 1788 | 2022055685CM01 | 251 | | | | | | | | | | | | | |
| 1789 | 2022055685CM01 | 10 | | | | | | | | | | | | | |
| 1790 | 2022055685CM01 | 223 | | | | | | | | | | | | | |
| 1791 | 2022048395CM01 | Non-analyse | Méhadone | 253 | Analogue narcotique | THC (Tétrahydrocannabinol) | 3 | Cannabis | Méhadone | Détece | Analogue narcotique | Méthamphétamine | Détece | Stimulant | Cocaine |
| 1792 | 202203825CM01 | Non-analyse | Citalopram / métabolite | Détece | Dépresseur | Kétamine | Détece | Anesthésique dissociatif | Détece | Dépresseur | Méthamphétamine | Détece | Stimulant | Non actif | |
| 1793 | 2022033035CM01 | Non-analyse | Cocaine | Détece | Stimulant | Méthamphétamine | Détece | Stimulant | Diphendramine | Détece | Dépresseur | Bupropion | Détece | Non actif | |
| 1794 | 202204386CM01 | Non-analyse | Cocaine | 19 | Stimulant | Méthamphétamine | Détece | Stimulant | Trazodone | Détece | Dépresseur | Veratrine | Détece | Dépresseur | Clonazép |
| 1795 | 2022043395CM01 | Non-analyse | THC (métabolite THC-COOH) | Détece | Cannabis | Méthamphétamine | Détece | Stimulant | | | | | | | |
| 1796 | 2022050195CM01 | Non-analyse | Citalopram / métabolite | Détece | Dépresseur | Trazodone | Détece | Dépresseur | Cocaine | Détece | Stimulant | | | | |
| 1797 | 202204123CM01 | Non-analyse | Méthamphétamine | 306 | Stimulant | THC (Tétrahydrocannabinol) | 1.1 | Cannabis | | | | | | | |
| 1798 | 202204708CM01 | Néglig | Quétapine | 104 | Dépresseur | Topiramate | 6 | Dépresseur | Acide valproïque | 41 | Dépresseur | Alprazolam | Détece | Non actif | |
| 1799 | 2022048635CM01 | Non-analyse | GHB | 278 | Dépresseur | Cocaine (métabolite) | Détece | Non actif | | | | | | | |
| 1800 | 202204386CM01 | Non-analyse | Cocaine (métabolite) | Détece | Non actif | Cocaine | Détece | Stimulant | Quétapine (métabolite) | Détece | Non actif | Midazolam (métabolite) | Détece | Non actif | |
| 1801 | 202204386CM01 | Non-analyse | THC (Tétrahydrocannabinol) | 12 | Cannabis | Cocaine (métabolite) | Détece | Non actif | | | | | | | |
| 1802 | 202204386CM01 | Non-analyse | Diazepam | 250 | Dépresseur | Cocaine (métabolite) | Détece | Non actif | | | | | | | |
| 1803 | 202204735CM01 | Non-analyse | THC (Tétrahydrocannabinol) | 14 | Cannabis | Trazodone | Détece | Dépresseur | Lorazépam | 19 | Dépresseur | | | | |
| 1804 | 2022033035CM01 | Non-analyse | Cocaine | 5.4 | Cannabis | Cocaine (métabolite) | Détece | Non actif | | | | | | | |
| 1805 | 2022033035CM01 | Non-analyse | THC (Tétrahydrocannabinol) | 13 | Cannabis | Lorazépam | Détece | Dépresseur | MDMA | Détece | Stimulant | MDA | Détece | Stimulant | |
| 1806 | 2022033035CM01 | Non-analyse | THC (Tétrahydrocannabinol) | 11 | Cannabis | Lorazépam | Détece | Dépresseur | MDMA | Détece | Stimulant | MDA | Détece | Stimulant | |
| 1807 | 2022033035CM01 | Non-analyse | Cocaine | Détece | Stimulant | GHB | Détece | Dépresseur | Zopiclone | Détece | Dépresseur | Cocaine (métabolite) | Détece | Dépresseur | Clonazép |
| 1808 | 2022037955CM01 | Non-analyse | Méthamphétamine | Détece | Stimulant | GHB | Détece | Dépresseur | THC (métabolite THC-COOH) | Détece | Cannabis | Cocaine (métabolite) | Détece | Non actif | |
| 1809 | 2022047885CM01 | Non-analyse | Alprazolam | 47 | Dépresseur | Amisulpride | 18 | Dépresseur | Codine | 34 | Analogue narcotique | Quétapine | Détece | Dépresseur | Veratrine |
| 1810 | 2022053025CM01 | Non-analyse | Cocaine | Détece | Stimulant | THC (métabolite THC-COOH) | Détece | Cannabis | | | | | | | |
| 1811 | 2022033035CM01 | Non-analyse | | | | | | | | | | | | | |

Formulaire BD Consultation Menu déroulant Modifier version

Moyenne : 25290251.73 Nb (non vides) : 87 Somme : 203222013.8 Paramètres d'affichage 70%



A teaser of our upcoming menus

- Just a sample of what we are doing with R!
- **Take home message: if you have an idea, there's a way to do it in R.**
- But we want to explore more...
- Docking with other tools
 - Excel with VBA
 - Power BI (Microsoft)
- Version control using GitHub (would help with accreditation requirements!)

Want to get cooking?

Do you need to learn it early to be good with it?



Want to get cooking? Several free resources online!

R Studio Cheat Sheets – super dense, but it's all there...

> <https://www.rstudio.com/resources/cheatsheets/>

Data transformation with dplyr : CHEAT SHEET

dplyr functions work with pipes and expect tidy data. In tidy data:

- Each variable is in its own column
- Each observation, or case, is in its own row
- a %>% f() becomes f(x, y)

Summarise Cases

Apply summary functions to columns to create a new table of summary statistics. Summary functions take vectors as input and return one value (see back).

summary function

- `summarise(data, ...)` Compute table of summaries. `summarise(mtcars, avg = mean(mpg))`
- `count(data, ..., wt = NULL, sort = FALSE, name = NULL)` Count number of rows in each group defined by the variables in ... Also tally! `count(mtcars, cyl)`

Group Cases

Use `group_by(data, ...)`, `add = FALSE`, `drop = TRUE` to create a "grouped" copy of a table grouped by columns in ... `dplyr` functions will manipulate each "group" separately and combine the results.

Use `rowwise(data, ...)` to group data into individual rows. `dplyr` functions will compute results for each row. Also apply functions to list-columns. See tidy cheat sheet for list-column workflow.

`ungroup(x, ...)` Returns ungrouped copy of table. `ungroup(mtcars)`

Manipulate Cases

Row functions return a subset of rows as a new table.

- `filter(data, ...)` `preserve = FALSE` Extract rows that meet logical criteria. `filter(mtcars, mpg > 20)`
- `distinct(data, ..., keep_all = FALSE)` Remove rows with duplicate values. `distinct(mtcars, gear)`
- `slice(data, ..., preserve = FALSE)` Select rows by position. `slice(mtcars, 10:15)`
- `slice_sample(data, ..., n, prop, weight, by = NULL, replace = FALSE)` Randomly select rows. Use `n` to select a number of rows and `prop` to select a fraction of rows. `slice_sample(mtcars, n = 5, replace = TRUE)`
- `slice_min(data, order_by, ..., n, prop, with_ties = TRUE)` and `slice_max()` Select rows with the lowest and highest values. `slice_min(mtcars, mpg, prop = 0.25)`
- `slice_head(data, ..., n, prop)` and `slice_tail()` Select the first or last rows. `slice_head(mtcars, n = 5)`

Logical and boolean operators to use with filter()

| | | | | | | | | | |
|--------------------|--------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| <code>=</code> | <code><</code> | <code><=</code> | <code>is.na()</code> | <code>is.na()</code> | <code>is.na()</code> | <code>is.na()</code> | <code>is.na()</code> | <code>is.na()</code> | <code>is.na()</code> |
| <code>></code> | <code>>=</code> | <code>!is.na()</code> | <code>!is.na()</code> | <code>!is.na()</code> | <code>!is.na()</code> | <code>!is.na()</code> | <code>!is.na()</code> | <code>!is.na()</code> | <code>!is.na()</code> |
| <code>&</code> | <code>&</code> | <code>&</code> | <code>&</code> | <code>&</code> | <code>&</code> | <code>&</code> | <code>&</code> | <code>&</code> | <code>&</code> |

See ?base::Logic and ?Comparison for help.

ARRANGE CASES

- `arrange(data, ...)` `by_group = FALSE` Order rows by values of a column or columns (low to high), use with `desc()` to order from high to low. `arrange(mtcars, mpg)`
- `arrange(desc(mpg))`

ADD CASES

- `add_row(data, ..., before = NULL, after = NULL)` Add one or more rows to a table. `add_row(cars, speed = 1, dist = 1)`

Manipulate Variables

Column functions return a set of columns as a new vector or table.

- `pull(data, var = -1, name = NULL, ...)` Extract column values as a vector, by name or index. `pull(mtcars, wt)`
- `select(data, ...)` Extract columns as a table. `select(mtcars, mpg, wt)`
- `relocate(data, ..., before = NULL, after = NULL)` Move columns to new position. `relocate(mtcars, mpg, cyl, after = last_col())`

Use these helpers with select() and across()

- `contains(match)` `num_range(prefix, range)` `ends_with(match)` `all_of(x)(any_of(x, ... vars))` `starts_with(match)` `matches(match)` `everything()`

MANIPULATE MULTIPLE VARIABLES AT ONCE

- `across(cols, funs, ..., names = NULL)` Summarise or mutate multiple columns in the same way. `summarise(mtcars, across(everything(), mean))`
- `c_across(cols)` Compute across columns in row-wise data. `transmute(bowlers(Lk(gas), total = sum(c_across(1:2)))`

MAKE NEW VARIABLES

Apply vectorized functions to columns. Vectorized functions take vectors as input and return vectors of the same length as output (see back).

vectorized function

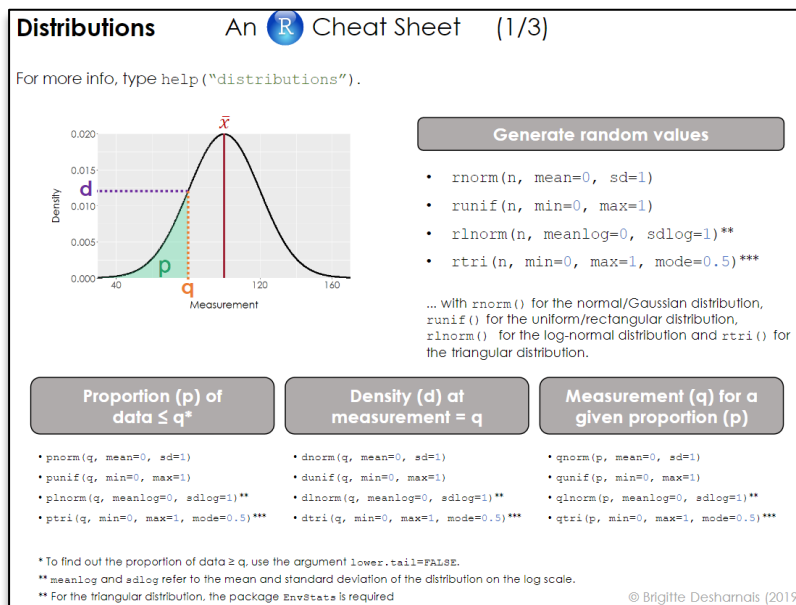
- `mutate(data, ..., keep = "all", before = NULL, after = NULL)` Compute new column(s). Also `add_column()`, `add_count()`, and `add_tally()`. `mutate(mtcars, gpm = 1 / mpg)`
- `transmute(data, ...)` Compute new column(s), drop others. `transmute(mtcars, gpm = 1 / mpg)`
- `rename(data, ...)` Rename columns. Use `rename_with()` to rename with a function. `rename(mtcars, distance = dist)`

Want to get cooking? Several free resources online!

TIAFT Bulletin Statistics Series, including R Cheat Sheets – less dense

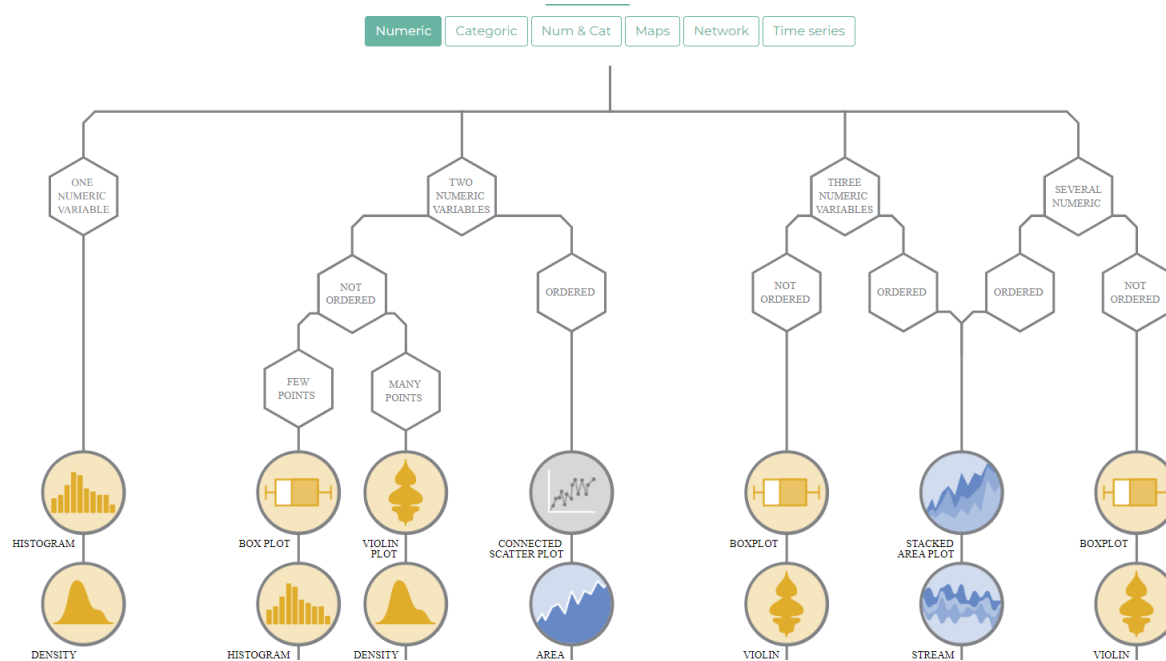
(but more stats oriented)

> <https://1drv.ms/u/s!At1r-QnPAVZ0jjVG1QA3ex84qiRI?e=UJKKj5>



Want to get cooking? Several free resources online!

From Data to Viz
> data-to-viz.com



A density plot shows the distribution of a numeric variable. It takes only numeric variables as input and is very close from an [histogram](#). It can be use in the same exact condition.

Common Mistakes

- Play with the [bandwidth](#) of your density function.
- Don't show the distribution of [more than ~5 variables](#). Use Violin or Ridge line plot instead.
- Avoid filling with [color palettes](#).

Code

[R graph gallery](#) [Python gallery](#) [D3.js gallery](#)

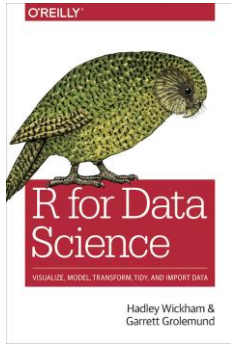
Read More

See the [dedicated page](#)

Want to get cooking? Several free resources online!

R Bootcamp – gentle introduction by Oregon Health & Science University

> <https://r-bootcamp.netlify.app/>



R for Data Science (H. Wickham & G. Grolemund)

> <https://r4ds.had.co.nz/index.html>

More free books!

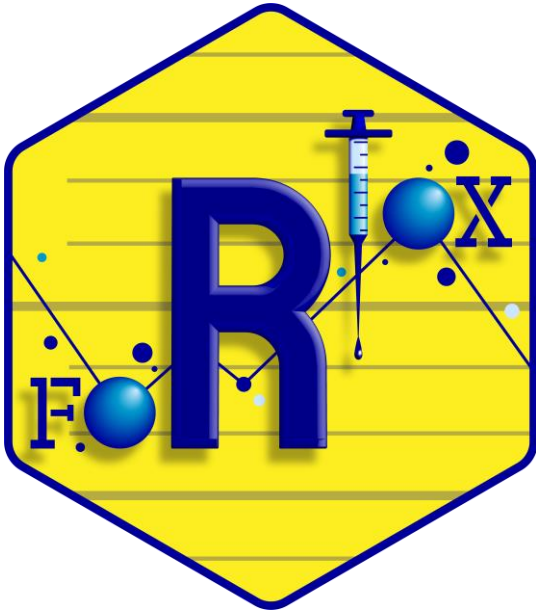
> <https://www.bigbookofr.com/>

Shiny tutorial by R Studio

> <https://shiny.rstudio.com/tutorial/written-tutorial/lesson1/>

Want to get cooking? Join FoRTox!

Googling “how to (...) in R” will get you far, but still, it helps to have a community!



> Forensic Toxicology R Users Group

<https://forms.gle/zejoxSemZGEHM7ak8>

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● [GitHub: This is where you can find all my R scripts!](#)

Find this presentation's material on GitHub

> <https://github.com/ToxBrigitte/soft-2022-automation-workshop>

- For those not familiar with GitHub, download everything on your computer like this:

