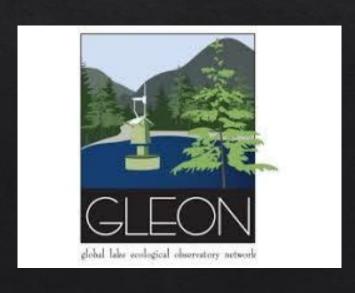
# Introduction to data.table

Fast data handling and concise syntax in R



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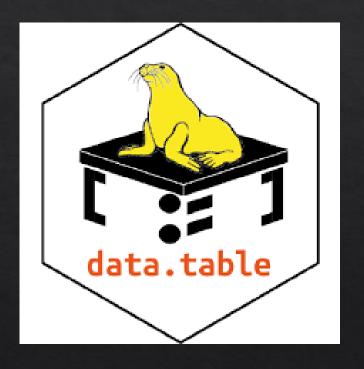


## Workshop Outline

- ♦ Introductory presentation (20 min.)
  - ♦ What is data.table and why should you use it?
- ♦ Questions (5 min.)
- Walkthrough script Part 1 (20 min.)
- ♦ Break (10 min.)
- ♦ Walkthrough script Part 2 (20 min.)
- ♦ Questions & Try it out yourself! (15 min.)

#### What is data.table?

- ♦ CRAN R package
  - Main developer: Matt Dowle
  - ♦ First release: 2006
  - ♦ Active development
- ♦ Extension to data.frame
- ♦ Fast data handling
  - ♦ Main focus of the package
  - ♦ Reading, writing, selecting, joining, etc.
- Consistent and concise syntax



# Why this workshop?

- ♦ One of the main packages in R, but little advertised
- Can help you write clearer code and faster scripts
- Faster data handling is particularly useful for GLEONites!
  - ♦ High-frequency data and long-term datasets
  - ♦ Modelling
  - ♦ Global analyses

- ♦ It is often said that R is slow
- Whether speed is relevant depends on the size of your data and how often you will call your function
  - ♦ Single use:
    - ♦ Small datasets (<10MB), speed is not so important
    - ♦ Large datasets (hundreds of MB), speed matters a lot
  - ♦ Repeated analyses:
    - ♦ Even small improvements in performance can save you hours

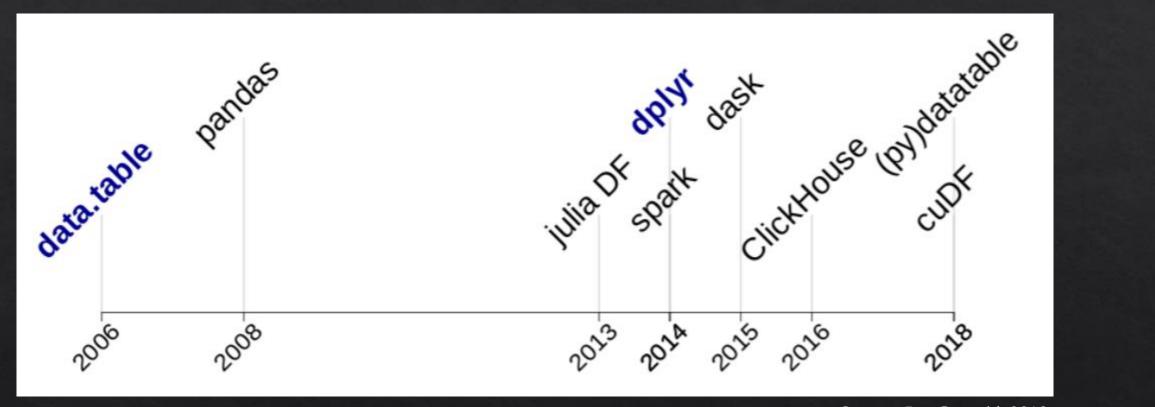
#### <u>Indication of file sizes</u>:

13 years of water temperature data, text format, 12 distinct depths

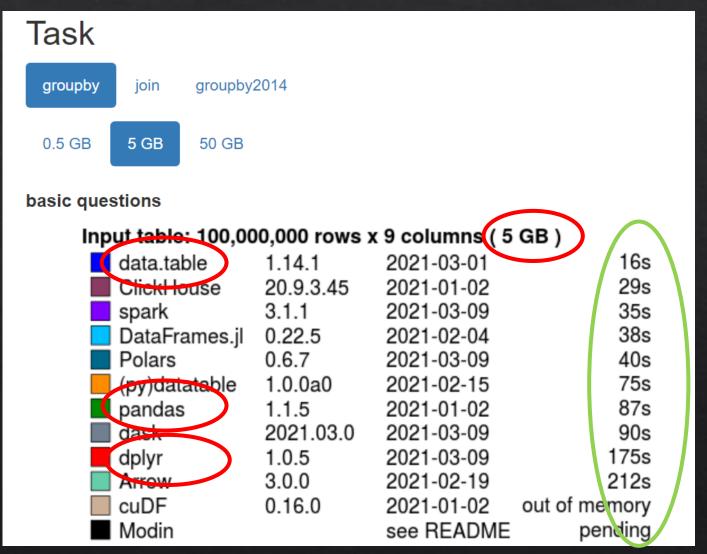
- Hourly: 11 MB

- 10 minutes: 85 MB

- 2 minutes: 387 MB



Source: Jan Gorecki, 2019



Data.table can be multiple factors faster than alternatives, even those in other languages!

groupby	<b>join</b> groupby	2014						
0.5 GB	5 GB 50 GB							
basic questions								
Inp	ut table: 10,000	0,000 rows x	7 columns ( 0	).4 GB )				
	cuDF	0.16.0	2021-01-02	2s				
	Polars	0.6.7	2021-03-09	6s				
	data.table	1.14.1	2021-03-01	8s				
	ClickHouse	20.9.3.45	2021-01-02	22s				
	dplyr	1.0.5	2021-03-09	29s				
	spark	3.1.1	2021-03-09	41s				
	pandas	1.1.5	2021-01-02	56s				
	DataFrames.jl	0.22.5	2021-02-04	129s				
	(py)datatable	1.0.0a0	2021-02-15	195s				
	dask	2021.03.0	2021-03-09	938s				
	Arrow	3.0.0	2021-02-19	not yet implemented				
	Modin		see READM	E pending				

	groupby	join gro	pupby2014					
	0.5 GB	5 GB 50	GB					
basic questions								
Input table: 1,000,000,000 rows x 9 columns ( 50 GB )								
		data.table	1.14.1	2021-03-01	197s			
		Polars	0.6.7	2021-03-09	311s			
		ClickHouse	20.9.3.4	5 2021-01-02	332s			
		spark	3.1.1	2021-03-09	537s			
		DataFrame	s.jl 0.22.5	2021-02-04	708s			
		(py)datatak	ole 1.0.0a0	2021-02-15	727s			
		dplyr	1.0.5	2021-03-09	internal error			
		pandas	1.1.5	2021-01-02	out of memory			
		dask	2021.03	.0 2021-03-09	out of memory			
		cuDF	0.16.0	2021-01-02	out of memory			
		Arrow	3.0.0	2021-02-19	internal error			
		Modin		see README	pending			

```
groupby join groupby2014

0.5 GB 5 GB 50 GB

basic questions

Input table: 1,000,000,000 rows x 9 columns (50 GB)
data.table 1.14.1 2021-03-01 197s
```

for data.frame-like objects, R is not slow – if you're using data.table!

dplyr	1.0.5	2021-03-09	internal error
pandas	1.1.5	2021-01-02	out of memory
dask	2021.03.0	2021-03-09	out of memory
cuDF	0.16.0	2021-01-02	out of memory
Arrow	3.0.0	2021-02-19	internal error
Modin		see README	pending

https://h2oai.github.io/db-benchmark/, 2021-03-15

#### What makes data.table so fast?

- ♦ Speed has been the main focus of data.table during its whole development
- ♦ Efficient algorithms (C code)
- ♦ Internal structure of data.tables
- Less in-memory copying
- ♦ Low-level multi-threading
- When using data.table, most of this will happen internally, so you can code "normally" and still achieve a high efficiency

# Not just fast, a clear syntax as well!

- ♦ Data.table has a syntax that is different from base R
- ♦ "Come for the performance, stay for the syntax" Elio Campitelli, R blogger
- Concise
- ♦ Consistent
- ♦ DT[i, j, by]
  - ♦ i: what rows
  - ♦ j: what to do
  - ♦ by: by what groups
- ♦ More on this in the walkthrough...

# Not just fast, a clear syntax as well!

- ♦ Tidyverse also offers an alternative syntax to base R
  - ♦ Verbose
  - ♦ Verbs as functions
  - ♦ Pipes
- ♦ Data.table
  - ♦ More alike query languages such as SQL
  - ♦ Closer to base-R
- Which one is "better" ultimately comes down to preference, but to get optimal performance, you'll have to use at least some of the data.table syntax
- dtplyr package (<a href="https://cran.r-project.org/web/packages/dtplyr/index.html">https://cran.r-project.org/web/packages/dtplyr/index.html</a>)

### Short comparison syntax of data.table and tidyverse

```
<u>Tidyverse</u>
                                                         data.table
                                                         DT[, sum(y), by = z]
DF %>%
  group_by(z) \% >%
  summarise(sum(y))
ans <- DF %>%
                                                         DT[, y := cumsum(y), by = z]
  group_by(z) \% >%
  mutate(y = cumsum(y))
DF %>%
                                                         DT[x > 2, sum(y), by = z]
  filter(x>2) %>%
  group_by(z) \% >%
  summarise(sum(y))
\overline{\text{ans}} = \overline{\text{C-DF \%}} > \%
                                                         DT[x > 2, y := cumsum(y), by = z]
  group_by(z) \% >%
  mutate(y = replace(y, 
                      which(x > 2),
                      cumsum(y)))
```

### Short comparison syntax of data.table and tidyverse

```
Tidyverse

diamondsDF %>%
filter(cut != "Fair") %>%
group_by(cut) %>%
summarize(
   AvgPrice = mean(price),
   MedianPrice = as.numeric(median(price)),
   Count = n()
) %>%
arrange(desc(Count))
```

```
data.table

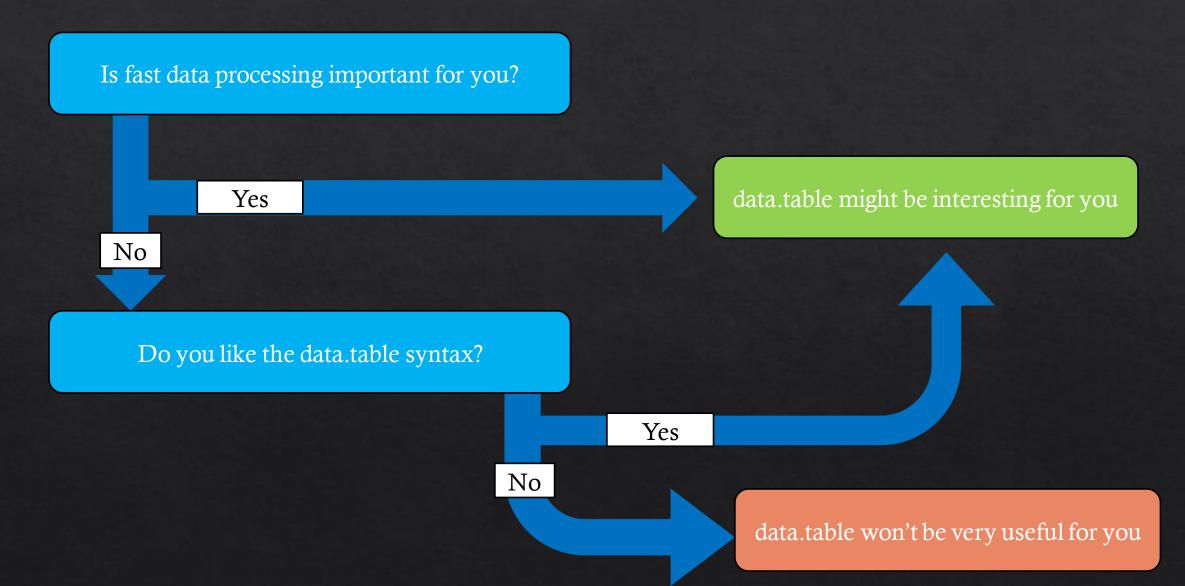
diamondsDT[
    cut != "Fair",
    .(AvgPrice = mean(price),
        MedianPrice = as.numeric(median(price)),
        Count = .N
    ),
    by = cut

[[
    order(-Count)
]
```

## But combinations are possible!

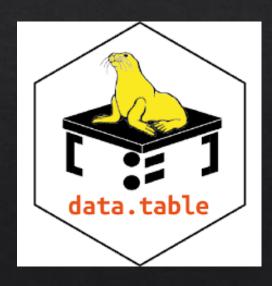
```
data.table with pipes
library(magrittr)
library(ggplot2)
temperature %>%
 [level == 1000] \% > \%
 [, mean(air), by = [(lat, lon)] %>%
 [1at > 0] \% > \%
 ggplot(aes(lon, lat)) +
 geom_raster(aes(fill = V1), interpolate = TRUE)
```

# Should you use data.table?



### In summary

- data.table is very fast on especially large datasets
  - ♦ Significantly faster than many alternatives
  - ♦ File size needs to be large to notice difference
- Consistent and concise syntax
  - ♦ We'll learn how to work with this during the walkthrough
- ♦ To the hands-on part!





#### Outline

- What is data.table (extension to data.frame) Package creator Matt Dowle. Website link. Already in R for a long time.
- ♦ Why this presentation: I like data.table a lot, helped me write better code, not advertised that much. But especially the speed argument may be very relevant for GLEON!
- ♦ First presentation (15 min.), then walkthrough R script, then questions/exercises
- ♦ Speed! Internal optimisation, update by reference. R still has the reputation of being a slow language, but actually...
- ♦ Show benchmark website (faster than base R, Python, tidyverse...) Now the speed argument is important, but realise that you need to have data >100MB to actually notice anything, or many repeated operations.
- ♦ Syntax (come for speed, stay for syntax) (quick comparison with tidyverse; concise and consistent vs. Verbose and accessible to non-coders) -> Comes down to preference

#### Outline

- ♦ Syntax is consistent, as it is always i, j, by (some additional arguments possible)
- ♦ Some like it, some prefer base R or tidyverse, and in that case, the syntax is a burden when you would like to use the fast computation of data.table. I hope that my walkthrough would in that case help (also consider dtplyr if you like the tidyverse syntax; not as fast as data.table, but it comes a long way)
- ♦ Additional upsides: stable, rigorous testing, large community, updates
- ♦ I like it a lot, but this is not a sales pitch; essentially, if you need the speed, check it out. If you like the syntax, also check it out.
- ♦ I hope that my walkthrough script can simultaneously act as a "cheat sheet" when using data.table for the first time.

Speed is good, but you should focus on the bottlenecks