# Biostat 203B Homework 2

Due Feb 7, 2025 @ 11:59PM

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Display machine information for reproducibility:

#### sessionInfo()

```
R version 4.4.2 (2024-10-31)
Platform: aarch64-apple-darwin20
Running under: macOS Sequoia 15.3
Matrix products: default
BLAS:
        /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib;
locale:
[1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
time zone: America/Los_Angeles
tzcode source: internal
attached base packages:
[1] stats
              graphics grDevices utils
                                            datasets methods
                                                                 base
loaded via a namespace (and not attached):
 [1] compiler_4.4.2
                       fastmap_1.2.0
                                         cli_3.6.3
                                                            tools_4.4.2
 [5] htmltools_0.5.8.1 rstudioapi_0.17.1 yaml_2.3.10
                                                            rmarkdown_2.29
 [9] knitr_1.49
                       jsonlite_1.8.9
                                         xfun_0.50
                                                            digest_0.6.37
[13] rlang_1.1.4
                       evaluate_1.0.1
```

Load necessary libraries (you can add more as needed).

```
library(arrow)
Attaching package: 'arrow'
The following object is masked from 'package:utils':
    timestamp
library(data.table)
library(duckdb)
Loading required package: DBI
library(memuse)
library(pryr)
Attaching package: 'pryr'
The following object is masked from 'package:data.table':
    address
library(R.utils)
Loading required package: R.oo
Loading required package: R.methodsS3
R.methodsS3 v1.8.2 (2022-06-13 22:00:14 UTC) successfully loaded. See ?R.methodsS3 for help.
R.oo v1.27.0 (2024-11-01 18:00:02 UTC) successfully loaded. See ?R.oo for help.
Attaching package: 'R.oo'
```

```
The following object is masked from 'package:R.methodsS3':
    throw
The following objects are masked from 'package:methods':
   getClasses, getMethods
The following objects are masked from 'package:base':
    attach, detach, load, save
R.utils v2.12.3 (2023-11-18 01:00:02 UTC) successfully loaded. See ?R.utils for help.
Attaching package: 'R.utils'
The following object is masked from 'package:arrow':
    timestamp
The following object is masked from 'package:utils':
    timestamp
The following objects are masked from 'package:base':
    cat, commandArgs, getOption, isOpen, nullfile, parse, use, warnings
library(tidyverse)
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr
          1.1.4
                    v readr
                                 2.1.5
v forcats 1.0.0
                                 1.5.1
                     v stringr
v ggplot2
          3.5.1
                     v tibble
                                 3.2.1
v lubridate 1.9.4
                     v tidyr
                                 1.3.1
          1.0.2
v purrr
```

```
----- tidyverse_conflicts() --
-- Conflicts -----
x dplyr::between()
                         masks data.table::between()
x purrr::compose()
                         masks pryr::compose()
x lubridate::duration() masks arrow::duration()
x tidyr::extract()
                         masks R.utils::extract()
x dplyr::filter()
                         masks stats::filter()
x dplyr::first()
                        masks data.table::first()
x lubridate::hour()
                        masks data.table::hour()
x lubridate::isoweek()
                        masks data.table::isoweek()
x dplyr::lag()
                        masks stats::lag()
x dplyr::last()
                         masks data.table::last()
x lubridate::mday()
                         masks data.table::mday()
x lubridate::minute()
                         masks data.table::minute()
x lubridate::month()
                         masks data.table::month()
x purrr::partial()
                         masks pryr::partial()
x lubridate::quarter()
                         masks data.table::quarter()
x lubridate::second()
                         masks data.table::second()
                         masks data.table::transpose()
x purrr::transpose()
x lubridate::wday()
                         masks data.table::wday()
x lubridate::week()
                         masks data.table::week()
                         masks pryr::where()
x dplyr::where()
                        masks data.table::yday()
x lubridate::yday()
x lubridate::year()
                         masks data.table::year()
i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become
```

Display memory information of your computer

```
memuse::Sys.meminfo()
```

Totalram: 16.000 GiB Freeram: 1.742 GiB

In this exercise, we explore various tools for ingesting the MIMIC-IV data introduced in homework 1.

Display the contents of MIMIC hosp and icu data folders:

```
ls -l ~/mimic/hosp/

total 12306256
-rw-rw-r--@ 1 ningkezhang staff 19928140 Jun 24 2024 admissions.csv.gz
```

```
427554 Apr 12 2024 d_hcpcs.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
-rw-rw-r--@ 1 ningkezhang
                          staff
                                     876360 Apr 12 2024 d_icd_diagnoses.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
                                     589186 Apr 12 2024 d_icd_procedures.csv.gz
-rw-rw-r--@ 1 ningkezhang
                                      13169 Oct 3 06:07 d_labitems.csv.gz
                          staff
-rw-rw-r--@ 1 ningkezhang
                          staff
                                   33564802 Oct 3 06:07 diagnoses icd.csv.gz
-rw-rw-r--@ 1 ningkezhang staff
                                    9743908 Oct 3 06:07 drgcodes.csv.gz
-rw-rw-r--@ 1 ningkezhang staff
                                  811305629 Apr 12 2024 emar.csv.gz
-rw-rw-r--@ 1 ningkezhang staff
                                  748158322 Apr 12 2024 emar_detail.csv.gz
                                    2162335 Apr 12 2024 hcpcsevents.csv.gz
-rw-rw-r--@ 1 ningkezhang staff
-rw-rw-r--@ 1 ningkezhang staff
                                       2907 Dec 28 18:04 index.html
-rw-rw-r--0 1 ningkezhang staff
                                 2592909134 Oct 3 06:08 labevents.csv.gz
                                  117644075 Oct 3 06:08 microbiologyevents.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
-rw-rw-r--@ 1 ningkezhang staff
                                   44069351 Oct 3 06:08 omr.csv.gz
-rw-rw-r--0 1 ningkezhang staff
                                    2835586 Apr 12
                                                    2024 patients.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
                                  525708076 Apr 12
                                                    2024 pharmacy.csv.gz
-rw-rw-r--@ 1 ningkezhang
                         staff
                                  666594177 Apr 12
                                                    2024 poe.csv.gz
-rw-rw-r--@ 1 ningkezhang
                         staff
                                   55267894 Apr 12 2024 poe_detail.csv.gz
                                  606298611 Apr 12
                                                    2024 prescriptions.csv.gz
-rw-rw-r--@ 1 ningkezhang staff
-rw-rw-r--@ 1 ningkezhang
                                    7777324 Apr 12 2024 procedures_icd.csv.gz
                          staff
-rw-rw-r--0 1 ningkezhang staff
                                     127330 Apr 12 2024 provider.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
                                    8569241 Apr 12
                                                    2024 services.csv.gz
                                   46185771 Oct 3 06:08 transfers.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
```

#### ls -l ~/mimic/icu/

```
total 8506792
```

```
-rw-rw-r--@ 1 ningkezhang
                          staff
                                      41566 Apr 12 2024 caregiver.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
                                 3502392765 Apr 12
                                                    2024 chartevents.csv.gz
-rw-rw-r--@ 1 ningkezhang
                          staff
                                      58741 Apr 12 2024 d_items.csv.gz
                                   63481196 Apr 12 2024 datetimeevents.csv.gz
-rw-rw-r--@ 1 ningkezhang
                         staff
-rw-rw-r--@ 1 ningkezhang
                          staff
                                    3342355 Oct 3 04:36 icustays.csv.gz
-rw-rw-r--@ 1 ningkezhang
                                       1336 Dec 28 18:04 index.html
                          staff
-rw-rw-r--@ 1 ningkezhang staff
                                  311642048 Apr 12
                                                    2024 ingredientevents.csv.gz
-rw-rw-r--@ 1 ningkezhang staff
                                                    2024 inputevents.csv.gz
                                  401088206 Apr 12
-rw-rw-r--@ 1 ningkezhang staff
                                   49307639 Apr 12
                                                    2024 outputevents.csv.gz
-rw-rw-r--@ 1 ningkezhang staff
                                   24096834 Apr 12 2024 procedureevents.csv.gz
```

# Q1. read.csv (base R) vs read\_csv (tidyverse) vs fread (data.table)

#### Q1.1 Speed, memory, and data types

There are quite a few utilities in R for reading plain text data files. Let us test the speed of reading a moderate sized compressed csv file, admissions.csv.gz, by three functions: read.csv in base R, read\_csv in tidyverse, and fread in the data.table package.

Which function is fastest? Is there difference in the (default) parsed data types? How much memory does each resultant dataframe or tibble use? (Hint: system.time measures run times; pryr::object\_size measures memory usage; all these readers can take gz file as input without explicit decompression.)

Solusion: fread is the fastest. The default parsed data types are different. read.csv uses factor for Character Columns, num or int for Numeric Columns, and Chr for Logical Columns, and automatically converted from character for factors. read\_csv uses chr for character columns, dbl or int for numeric columns, parsed as date or datetime for date columns,lgl for logical columns. fread uses chr for character columns, num or int for numeric columns, lgl for logical columns. The memory usage of fread is the smallest.

```
system.time(read.csv("~/mimic/hosp/admissions.csv.gz"))
   user system elapsed
  4.905
         0.045
                  4.950
system.time(read csv("~/mimic/hosp/admissions.csv.gz"))
Rows: 546028 Columns: 16
-- Column specification -----
Delimiter: ","
chr (8): admission_type, admit_provider_id, admission_location, discharge_1...
     (3): subject_id, hadm_id, hospital_expire_flag
dttm (5): admittime, dischtime, deathtime, edregtime, edouttime
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
   user system elapsed
  0.838
         0.077
                 0.473
```

```
user system elapsed
  0.492
        0.046
                 0.241
str(read.csv("~/mimic/hosp/admissions.csv.gz"))
'data.frame':
               546028 obs. of 16 variables:
                : int 10000032 10000032 10000032 10000038 10000084 10000084
 $ subject_id
                     : int 22595853 22841357 25742920 29079034 25022803 23052089 29888819
$ hadm_id
 $ admittime
                     : chr "2180-05-06 22:23:00" "2180-06-26 18:27:00" "2180-08-05 23:44:
 $ dischtime
                            "2180-05-07 17:15:00" "2180-06-27 18:49:00" "2180-08-07 17:50:
                     : chr
                            ...
 $ deathtime
                     : chr
 $ admission_type : chr
                            "URGENT" "EW EMER." "EW EMER." "EW EMER." ...
 $ admit_provider_id : chr "P49AFC" "P784FA" "P19UTS" "P060TX" ...
                            "TRANSFER FROM HOSPITAL" "EMERGENCY ROOM" "EMERGENCY ROOM" "EM
 $ admission_location : chr
 $ discharge_location : chr "HOME" "HOME" "HOSPICE" "HOME" ...
                     : chr "Medicaid" "Medicaid" "Medicaid" ...
 $ insurance
                            "English" "English" "English" "...
 $ language
                     : chr
                            "WIDOWED" "WIDOWED" "WIDOWED" ...
 $ marital_status
                    : chr
                            "WHITE" "WHITE" "WHITE" ...
 $ race
                     : chr
                     : chr "2180-05-06 19:17:00" "2180-06-26 15:54:00" "2180-08-05 20:58:
 $ edregtime
                      : chr
                            "2180-05-06 23:30:00" "2180-06-26 21:31:00" "2180-08-06 01:44:
 $ edouttime
 $ hospital_expire_flag: int    0 0 0 0 0 0 0 0 0 ...
glimpse(read_csv("~/mimic/hosp/admissions.csv.gz"))
Rows: 546028 Columns: 16
-- Column specification ----
Delimiter: ","
chr (8): admission_type, admit_provider_id, admission_location, discharge_1...
dbl (3): subject_id, hadm_id, hospital_expire_flag
dttm (5): admittime, dischtime, deathtime, edregtime, edouttime
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
Rows: 546,028
```

system.time(fread("~/mimic/hosp/admissions.csv.gz"))

Columns: 16

```
$ subject_id
                                                            <dbl> 10000032, 10000032, 10000032, 10000032, 10000068,~
                                                            <dbl> 22595853, 22841357, 25742920, 29079034, 25022803,~
$ hadm_id
                                                            <dttm> 2180-05-06 22:23:00, 2180-06-26 18:27:00, 2180-0~
$ admittime
$ dischtime
                                                            <dttm> 2180-05-07 17:15:00, 2180-06-27 18:49:00, 2180-0~
                                                            $ deathtime
                                                            <chr> "URGENT", "EW EMER.", "EW EMER.", "EW EMER.", "EU~
$ admission_type
$ admit_provider_id
                                                            <chr> "P49AFC", "P784FA", "P19UTS", "P060TX", "P39NW0",~
                                                            <chr> "TRANSFER FROM HOSPITAL", "EMERGENCY ROOM", "EMER~
$ admission_location
                                                            <chr> "HOME", "HOME", "HOSPICE", "HOME", NA, "HOME HEAL~
$ discharge_location
$ insurance
                                                            <chr> "Medicaid", "Medicaid", "Medicaid", N~
                                                            <chr> "English", "Engl
$ language
                                                            <chr> "WIDOWED", "WIDOWED", "WIDOWED", "WIDOWED", "SING~
$ marital_status
                                                            <chr> "WHITE", "WHITE", "WHITE", "WHITE", "WHITE", "WHI-
$ race
                                                            <dttm> 2180-05-06 19:17:00, 2180-06-26 15:54:00, 2180-0~
$ edregtime
                                                            <dttm> 2180-05-06 23:30:00, 2180-06-26 21:31:00, 2180-0~
$ edouttime
```

## str(fread("~/mimic/hosp/admissions.csv.gz"))

```
Classes 'data.table' and 'data.frame': 546028 obs. of 16 variables:
                     : int 10000032 10000032 10000032 10000032 10000068 10000084 10000084
$ subject id
$ hadm id
                     : int 22595853 22841357 25742920 29079034 25022803 23052089 29888819
                     : POSIXct, format: "2180-05-06 22:23:00" "2180-06-26 18:27:00" ...
$ admittime
$ dischtime
                     : POSIXct, format: "2180-05-07 17:15:00" "2180-06-27 18:49:00" ...
 $ deathtime
                     : POSIXct, format: NA NA ...
                            "URGENT" "EW EMER." "EW EMER." "EW EMER." ...
 $ admission_type
                     : chr
 $ admit_provider_id : chr "P49AFC" "P784FA" "P19UTS" "P060TX" ...
                            "TRANSFER FROM HOSPITAL" "EMERGENCY ROOM" "EMERGENCY ROOM" "EM
$ admission_location : chr
$ discharge_location : chr "HOME" "HOME" "HOSPICE" "HOME" ...
$ insurance
                     : chr "Medicaid" "Medicaid" "Medicaid" ...
 $ language
                     : chr
                            "English" "English" "English" ...
$ marital_status
                     : chr "WIDOWED" "WIDOWED" "WIDOWED" ...
                     : chr "WHITE" "WHITE" "WHITE" ...
$ race
                     : POSIXct, format: "2180-05-06 19:17:00" "2180-06-26 15:54:00" ...
$ edregtime
$ edouttime
                      : POSIXct, format: "2180-05-06 23:30:00" "2180-06-26 21:31:00" ...
$ hospital_expire_flag: int  0 0 0 0 0 0 0 0 0 ...
- attr(*, ".internal.selfref")=<externalptr>
```

# pryr::object\_size(read.csv("~/mimic/hosp/admissions.csv.gz"))

200.10 MB

```
pryr::object_size(read_csv("~/mimic/hosp/admissions.csv.gz"))

Rows: 546028 Columns: 16
-- Column specification -------
Delimiter: ","
chr (8): admission_type, admit_provider_id, admission_location, discharge_l...
dbl (3): subject_id, hadm_id, hospital_expire_flag
dttm (5): admittime, dischtime, deathtime, edregtime, edouttime

i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.

70.02 MB

pryr::object_size(fread("~/mimic/hosp/admissions.csv.gz"))
```

63.47 MB

# Q1.2 User-supplied data types

Re-ingest admissions.csv.gz by indicating appropriate column data types in read\_csv. Does the run time change? How much memory does the result tibble use? (Hint: col\_types argument in read\_csv.)

Solution: The run time is faster than the default data types. The memory usage is stay same

```
user system elapsed 0.794 0.055 0.406
```

70.02 MB

# Q2. Ingest big data files



Let us focus on a bigger file, labevents.csv.gz, which is about 130x bigger than admissions.csv.gz.

# ls -l ~/mimic/hosp/labevents.csv.gz

-rw-rw-r--@ 1 ningkezhang staff 2592909134 Oct 3 06:08 /Users/ningkezhang/mimic/hosp/labe

#### zcat < ~/mimic/hosp/labevents.csv.gz | head -10</pre>

```
labevent_id,subject_id,hadm_id,specimen_id,itemid,order_provider_id,charttime,storetime,value1,10000032,,2704548,50931,P69FQC,2180-03-23 11:51:00,2180-03-23 15:56:00,___,95,mg/dL,70,100 2,10000032,,36092842,51071,P69FQC,2180-03-23 11:51:00,2180-03-23 16:00:00,NEG,,,,,ROUTINE, 3,10000032,,36092842,51074,P69FQC,2180-03-23 11:51:00,2180-03-23 16:00:00,NEG,,,,,ROUTINE, 4,10000032,,36092842,51075,P69FQC,2180-03-23 11:51:00,2180-03-23 16:00:00,NEG,,,,,ROUTINE,"5,10000032,,36092842,51079,P69FQC,2180-03-23 11:51:00,2180-03-23 16:00:00,NEG,,,,,ROUTINE, 6,10000032,,36092842,51087,P69FQC,2180-03-23 11:51:00,,,,,,,ROUTINE,RANDOM.
7,10000032,,36092842,51089,P69FQC,2180-03-23 11:51:00,2180-03-23 16:15:00,,,,,,ROUTINE,PRESS 8,10000032,,36092842,51090,P69FQC,2180-03-23 11:51:00,2180-03-23 16:00:00,NEG,,,,,ROUTINE,MS 9,10000032,,36092842,51092,P69FQC,2180-03-23 11:51:00,2180-03-23 16:00:00,NEG,,,,,,ROUTINE,MS 9,10000032,NEG,MS 9,10000032,NEG,MS 9,10000032,NEG,MS 9,10000032,NEG,MS 9,10000032,NEG,MS 9,10000032,NEG,M
```

## Q2.1 Ingest labevents.csv.gz by read\_csv



Figure 1: readr

Try to ingest labevents.csv.gz using read\_csv. What happens? If it takes more than 3 minutes on your computer, then abort the program and report your findings.

**Solution:** It takes more than 5 minutes, and timing stopped, only 79.594 MiB of free RAM is available. Also, when I try **fread**, it shows: vector memory limit of 16.0 Gb reached, so processing Big Data is very demanding on RAM.

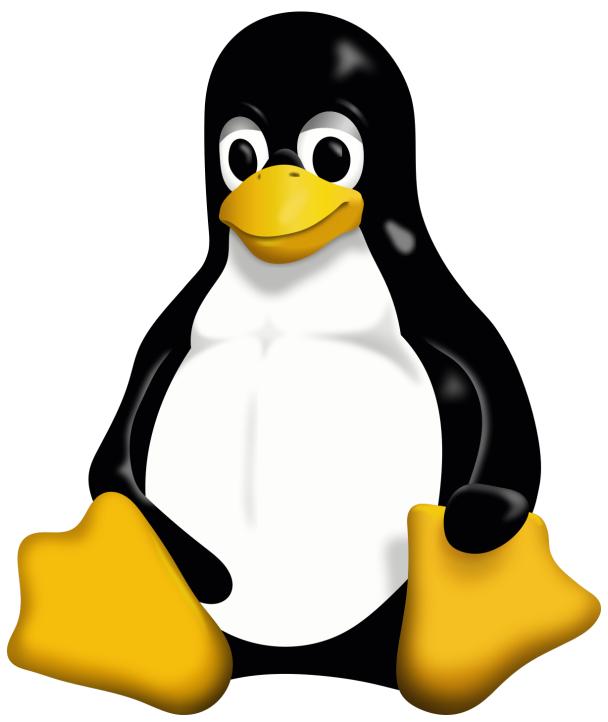
```
system.time(read_csv("~/mimic/hosp/labevents.csv.gz"))
pryr::object_size(read_csv("~/mimic/hosp/labevents.csv.gz"))
system.time(fread("~/mimic/hosp/labevents.csv.gz"))
```

## Q2.2 Ingest selected columns of labevents.csv.gz by read\_csv

Try to ingest only columns subject\_id, itemid, charttime, and valuenum in labevents.csv.gz using read\_csv. Does this solve the ingestion issue? (Hint: col\_select argument in read\_csv.)

Solution: For read\_csv still can not solve the ingestion issue, but fread can take 36.124s to process, which menas fewer columns come with less memory usage and less processing to get faster ingestion. read\_csv maybe cannot ingest gz files directly.

 ${\bf Q2.3\ Ingest\ a\ subset\ of\ labevents.csv.gz}$ 



Our first strategy to handle this big data file is to make a subset of the labevents data.

Read the MIMIC documentation for the content in data file labevents.csv.

In later exercises, we will only be interested in the following lab items: creatinine (50912), potassium (50971), sodium (50983), chloride (50902), bicarbonate (50882), hematocrit (51221), white blood cell count (51301), and glucose (50931) and the following columns: subject\_id, itemid, charttime, valuenum. Write a Bash command to extract these columns and rows from labevents.csv.gz and save the result to a new file labevents\_filtered.csv.gz in the current working directory. (Hint: Use zcat < to pipe the output of labevents.csv.gz to awk and then to gzip to compress the output. Do not put labevents\_filtered.csv.gz in Git! To save render time, you can put #| eval: false at the beginning of this code chunk. TA will change it to #| eval: true before rendering your qmd file.)

Display the first 10 lines of the new file labevents\_filtered.csv.gz. How many lines are in this new file, excluding the header? How long does it take read\_csv to ingest labevents\_filtered.csv.gz?

**Solution:** There are 32679896 lines, excluding the header. It takes 12.492s, with 1.05 GB RAM.

```
zcat < labevents_filtered.csv.gz | head -10
zcat < labevents_filtered.csv.gz | tail -n +2 | wc -1</pre>
```

```
subject_id,itemid,charttime,valuenum 10000032,50931,2180-03-23 11:51:00,95 10000032,50882,2180-03-23 11:51:00,27 10000032,50902,2180-03-23 11:51:00,101 10000032,50912,2180-03-23 11:51:00,0.4 10000032,50971,2180-03-23 11:51:00,3.7 10000032,50983,2180-03-23 11:51:00,136 10000032,51221,2180-03-23 11:51:00,45.4 10000032,51221,2180-03-23 11:51:00,3 10000032,51221,2180-05-06 22:25:00,42.6 32679896
```

```
system.time(
  lab <- read_csv("labevents_filtered.csv.gz")
)

Rows: 32679896 Columns: 4
-- Column specification ------
Delimiter: ","
dbl (3): subject_id, itemid, valuenum
dttm (1): charttime

i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.

user system elapsed
12.805 1.272 5.156

pryr::object_size(lab)</pre>
```

1.05 GB

#### Q2.4 Ingest labevents.csv by Apache Arrow

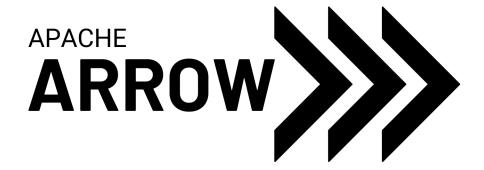


Figure 2: arrow

Our second strategy is to use Apache Arrow for larger-than-memory data analytics. Unfortunately Arrow does not work with gz files directly. First decompress labevents.csv.gz to labevents.csv and put it in the current working directory (do not add it in git!). To save render time, put #| eval: false at the beginning of this code chunk. TA will change it to #| eval: true when rendering your qmd file.

Then use arrow::open\_dataset to ingest labevents.csv, select columns, and filter itemid as in Q2.3. How long does the ingest+select+filter process take? Display the number of rows and the first 10 rows of the result tibble, and make sure they match those in Q2.3. (Hint: use dplyr verbs for selecting columns and filtering rows.)

Write a few sentences to explain what is Apache Arrow. Imagine you want to explain it to a layman in an elevator.

**Solution:** The process takes 37.127s. Apache Arrow keeps data in a special format that we can read instantly, without wasting time converting files.

```
gunzip -c ~/mimic/hosp/labevents.csv.gz > labevents.csv
```

```
user system elapsed 37.771 4.019 35.867
```

```
nrow(arrow)
```

#### [1] 32679896

```
print(head(arrow, 10))
```

```
2
     10000032 50882 2180-03-23 04:51:00
                                             27
3
     10000032 50902 2180-03-23 04:51:00
                                            101
4
     10000032 50912 2180-03-23 04:51:00
                                              0.4
5
     10000032 50971 2180-03-23 04:51:00
                                              3.7
6
     10000032 50983 2180-03-23 04:51:00
                                            136
7
     10000032 51221 2180-03-23 04:51:00
                                             45.4
8
     10000032 51301 2180-03-23 04:51:00
                                              3
9
     10000032 51221 2180-05-06 15:25:00
                                             42.6
10
     10000032 51301 2180-05-06 15:25:00
                                              5
```

#### Q2.5 Compress labevents.csv to Parquet format and ingest/select/filter



Figure 3: parquet

Re-write the csv file labevents.csv in the binary Parquet format (Hint: arrow::write\_dataset.) How large is the Parquet file(s)? How long does the ingest+select+filter process of the Parquet file(s) take? Display the number of rows and the first 10 rows of the result tibble and make sure they match those in Q2.3. (Hint: use dplyr verbs for selecting columns and filtering rows.)

Write a few sentences to explain what is the Parquet format. Imagine you want to explain it to a layman in an elevator.

**Solution:** The Parquet is 2.5G, process takes 6.801s. Parquet is like a ZIP file for big data. It compresses and organizes information, only reads what's needed, so it takes up less space and loads much faster than read\_csv.

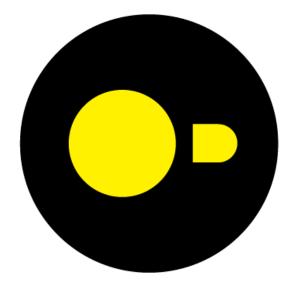
user system elapsed 6.556 1.990 2.967

```
nrow(parquet_df)
```

#### [1] 32679896

```
print(head(parquet_df, 10))
```

```
# A tibble: 10 \times 4
   subject_id itemid charttime
                                         valuenum
        <int> <int> <dttm>
                                            <dbl>
     10000032 50931 2180-03-23 04:51:00
                                             95
 1
 2
     10000032 50882 2180-03-23 04:51:00
                                             27
 3
     10000032 50902 2180-03-23 04:51:00
                                            101
     10000032 50912 2180-03-23 04:51:00
 4
                                              0.4
5
     10000032 50971 2180-03-23 04:51:00
                                              3.7
6
     10000032 50983 2180-03-23 04:51:00
                                            136
7
     10000032 51221 2180-03-23 04:51:00
                                             45.4
8
     10000032 51301 2180-03-23 04:51:00
                                              3
9
     10000032 51221 2180-05-06 15:25:00
                                             42.6
10
     10000032 51301 2180-05-06 15:25:00
                                              5
```



# DuckDB

Figure 4: duckdb

Ingest the Parquet file, convert it to a DuckDB table by arrow::to\_duckdb, select columns, and filter rows as in Q2.5. How long does the ingest+convert+select+filter process take? Display the number of rows and the first 10 rows of the result tibble and make sure they match those in Q2.3. (Hint: use dplyr verbs for selecting columns and filtering rows.)

Write a few sentences to explain what is DuckDB. Imagine you want to explain it to a layman in an elevator.

**Solution:** It takes 7.782s to process. DuckDB is a high-performance, lightweight database that runs locally. If opening an excel table with millions of rows of data will get stuck, use DuckDB to filter, calculate and merge data will be on the fly.

user system elapsed 7.340 2.195 3.022

```
nrow(duckdb_df)
```

#### [1] 32679896

```
print(head(duckdb_df, 10))
```

```
# A tibble: 10 x 4
  subject_id itemid charttime
                                        valuenum
                                           <dbl>
       <dbl> <dbl> <dttm>
1
    10003417 51301 2111-01-11 11:42:00
                                             7
2
    10003417 50882 2111-01-24 18:03:00
                                            25
    10003417 50902 2111-01-24 18:03:00
3
                                           103
 4
    10003417 50912 2111-01-24 18:03:00
                                             2.5
5
    10003417 50971 2111-01-24 18:03:00
6
    10003417 50983 2111-01-24 18:03:00
                                           138
7
    10003417 50882 2111-02-18 13:34:00
                                            23
8
    10003417 50902 2111-02-18 13:34:00
                                           101
9
    10003417 50912 2111-02-18 13:34:00
                                             2.8
    10003417 50931 2111-02-18 13:34:00
10
                                            90
```

#### dbDisconnect(con)

#### Q3. Ingest and filter chartevents.csv.gz

chartevents.csv.gz contains all the charted data available for a patient. During their ICU stay, the primary repository of a patient's information is their electronic chart. The itemid variable indicates a single measurement type in the database. The value variable is the value measured for itemid. The first 10 lines of chartevents.csv.gz are

```
zcat < ~/mimic/icu/chartevents.csv.gz | head -10</pre>
```

```
subject_id,hadm_id,stay_id,caregiver_id,charttime,storetime,itemid,value,valuenum,valueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,walueuom,auueuom,walueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuom,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuoon,auueuo
```

How many rows? 433 millions.

```
zcat < ~/mimic/icu/chartevents.csv.gz | tail -n +2 | wc -l</pre>
```

d\_items.csv.gz is the dictionary for the itemid in chartevents.csv.gz.

```
zcat < ~/mimic/icu/d_items.csv.gz | head -10</pre>
```

```
itemid, label, abbreviation, linksto, category, unitname, param_type, lownormal value, highnormal value 220001, Problem List, Problem List, chartevents, General, Text,,
220003, ICU Admission date, ICU Admission date, datetime events, ADT, Date and time,,
220045, Heart Rate, HR, chartevents, Routine Vital Signs, bpm, Numeric,,
220046, Heart rate Alarm - High, HR Alarm - High, chartevents, Alarms, bpm, Numeric,,
220047, Heart Rate Alarm - Low, HR Alarm - Low, chartevents, Alarms, bpm, Numeric,,
220048, Heart Rhythm, Heart Rhythm, chartevents, Routine Vital Signs, Text,,
220050, Arterial Blood Pressure systolic, ABPs, chartevents, Routine Vital Signs, mmHg, Numeric, 60
220051, Arterial Blood Pressure diastolic, ABPd, chartevents, Routine Vital Signs, mmHg, Numeric, 60
220052, Arterial Blood Pressure mean, ABPm, chartevents, Routine Vital Signs, mmHg, Numeric,
```

In later exercises, we are interested in the vitals for ICU patients: heart rate (220045), mean non-invasive blood pressure (220181), systolic non-invasive blood pressure (220179), body temperature in Fahrenheit (223761), and respiratory rate (220210). Retrieve a subset of chartevents.csv.gz only containing these items, using the favorite method you learnt in Q2.

Document the steps and show code. Display the number of rows and the first 10 rows of the result tibble.

**Solution:** The file chartevents\_filtered.csv.gz which is after compressed contains 30195426 rows, the first 10 rows are displaied.

```
# Open dataset as arrow table
chartevents <- open_dataset("~/mimic/icu/chartevents.csv.gz", format = "csv")

# Filter and select relevant columns
chartevents_filtered <- chartevents |>
    select(subject_id, itemid, charttime, value) |>
    filter(itemid %in% c(220045, 220181, 220179, 223761, 220210)) |>
    collect()
```

```
# Save as a compressed CSV
write_csv_arrow(chartevents_filtered, "chartevents_filtered.csv.gz")
```

```
#Display first 10 rows
zcat < chartevents_filtered.csv.gz | head -10
#Display the number of rows
zcat < chartevents_filtered.csv.gz | tail -n +2 | wc -1</pre>
```

```
"subject_id","itemid","charttime","value"

10000032,223761,2180-07-23 06:00:00.000000-0800,"98.7"

10000032,220179,2180-07-23 06:11:00.000000-0800,"84"

10000032,220181,2180-07-23 06:11:00.000000-0800,"56"

10000032,220045,2180-07-23 06:12:00.000000-0800,"91"

10000032,220210,2180-07-23 06:12:00.000000-0800,"24"

10000032,220045,2180-07-23 06:30:00.000000-0800,"93"

10000032,220179,2180-07-23 06:30:00.000000-0800,"95"

10000032,220181,2180-07-23 06:30:00.000000-0800,"67"

10000032,220210,2180-07-23 06:30:00.000000-0800,"21"

30195426
```