Biostat 203B Homework 4

Due Mar 9 @ 11:59PM

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Display machine information:

```
sessionInfo()
```

```
R version 4.4.2 (2024-10-31)
Platform: x86_64-pc-linux-gnu
Running under: Ubuntu 24.04.1 LTS

Matrix products: default
BLAS:   /usr/lib/x86_64-linux-gnu/blas/libblas.so.3.12.0
LAPACK: /usr/lib/x86_64-linux-gnu/lapack/liblapack.so.3.12.0

locale:
 [1] LC_CTYPE=C.UTF-8       LC_NUMERIC=C           LC_TIME=C.UTF-8
 [4] LC_COLLATE=C.UTF-8     LC_MONETARY=C.UTF-8    LC_MESSAGES=C.UTF-8
 [7] LC_PAPER=C.UTF-8       LC_NAME=C              LC_ADDRESS=C
[10] LC_TELEPHONE=C         LC_MEASUREMENT=C.UTF-8 LC_IDENTIFICATION=C

time zone: America/Los_Angeles
tzcode source: system (glibc)

attached base packages:
[1] stats     graphics  grDevices utils     datasets  methods   base

loaded via a namespace (and not attached):
 [1] htmlwidgets_1.6.4 compiler_4.4.2    fastmap_1.2.0     cli_3.6.3
 [5] tools_4.4.2       htmltools_0.5.8.1 rstudioapi_0.17.1 yaml_2.3.10
 [9] rmarkdown_2.29    knitr_1.49        jsonlite_1.8.9    xfun_0.50
[13] digest_0.6.37     rlang_1.1.4       evaluate_1.0.3
```

Display my machine memory.

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

```
Totalram:  15.463 GiB
Freeram:    9.338 GiB
```

Load database libraries and the tidyverse frontend:

```
library(bigrquery)
library(dbplyr)
```

```r
library(DBI)
library(gt)
library(gtsummary)
library(tidyverse)
```

```
── Attaching core tidyverse packages ──────────────────────── tidyverse 2.0.0 ──
✓ dplyr     1.1.4     ✓ readr     2.1.5
✓ forcats   1.0.0     ✓ stringr   1.5.1
✓ ggplot2   3.5.1     ✓ tibble    3.2.1
✓ lubridate 1.9.4     ✓ tidyr     1.3.1
✓ purrr     1.0.2
── Conflicts ──────────────────────────────────────── tidyverse_conflicts() ──
✗ dplyr::filter() masks stats::filter()
✗ dplyr::ident()  masks dbplyr::ident()
✗ dplyr::lag()    masks stats::lag()
✗ dplyr::sql()    masks dbplyr::sql()
ℹ Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
errors
```

```r
library(tidyr)
library(dplyr)
library(readr)
library(forcats)
library(ggplot2)
library(shiny)
library(DT)
```

```
Attaching package: 'DT'

The following objects are masked from 'package:shiny':

    dataTableOutput, renderDataTable
```

```r
library(lubridate)
library(arrow)
```

```
Attaching package: 'arrow'

The following object is masked from 'package:lubridate':

    duration

The following object is masked from 'package:utils':

    timestamp
```

Q1. Compile the ICU cohort in HW3 from the Google BigQuery database

Below is an outline of steps. In this homework, we exclusively work with the BigQuery database and should not use any MIMIC data files stored on our local computer. Transform data as much as possible in BigQuery database and `collect()` the tibble **only at the end of Q1.7**.

Q1.1 Connect to BigQuery

Authenticate with BigQuery using the service account token. Please place the service account token (shared via BruinLearn) in the working directory (same folder as your qmd file). Do **not** ever add this token to your Git repository. If you do so, you will lose 50 points.

```
# path to the service account token
satoken <- "biostat-203b-2025-winter-4e58ec6e5579.json"
# BigQuery authentication using service account
bq_auth(path = satoken)
```

Connect to BigQuery database `mimiciv_3_1` in GCP (Google Cloud Platform), using the project billing account `biostat-203b-2025-winter`.

```
# connect to the BigQuery database `biostat-203b-2025-mimiciv_3_1`
con_bq <- dbConnect(
    bigrquery::bigquery(),
    project = "biostat-203b-2025-winter",
    dataset = "mimiciv_3_1",
    billing = "biostat-203b-2025-winter"
)
con_bq
```

```
<BigQueryConnection>
  Dataset: biostat-203b-2025-winter.mimiciv_3_1
  Billing: biostat-203b-2025-winter
```

List all tables in the `mimiciv_3_1` database.

```
dbListTables(con_bq)
```

```
 [1] "admissions"        "caregiver"        "chartevents"
 [4] "d_hcpcs"           "d_icd_diagnoses"  "d_icd_procedures"
 [7] "d_items"           "d_labitems"       "datetimeevents"
[10] "diagnoses_icd"     "drgcodes"         "emar"
[13] "emar_detail"       "hcpcsevents"      "icustays"
[16] "ingredientevents"  "inputevents"      "labevents"
[19] "microbiologyevents" "omr"             "outputevents"
[22] "patients"          "pharmacy"         "poe"
[25] "poe_detail"        "prescriptions"    "procedureevents"
```

```
[28] "procedures_icd"      "provider"              "services"
[31] "transfers"
```

Q1.2 `icustays` **data**

Connect to the `icustays` table.

```
# full ICU stays table
icustays_tble <- tbl(con_bq, "icustays") |>
  arrange(subject_id, hadm_id, stay_id) |>
  # show_query() |>
  print(width = Inf)
```

```
# Source:     SQL [?? x 8]
# Database:   BigQueryConnection
# Ordered by: subject_id, hadm_id, stay_id
   subject_id  hadm_id   stay_id first_careunit
        <int>    <int>     <int> <chr>
 1   10000032 29079034 39553978 Medical Intensive Care Unit (MICU)
 2   10000690 25860671 37081114 Medical Intensive Care Unit (MICU)
 3   10000980 26913865 39765666 Medical Intensive Care Unit (MICU)
 4   10001217 24597018 37067082 Surgical Intensive Care Unit (SICU)
 5   10001217 27703517 34592300 Surgical Intensive Care Unit (SICU)
 6   10001725 25563031 31205490 Medical/Surgical Intensive Care Unit (MICU/SICU)
 7   10001843 26133978 39698942 Medical/Surgical Intensive Care Unit (MICU/SICU)
 8   10001884 26184834 37510196 Medical Intensive Care Unit (MICU)
 9   10002013 23581541 39060235 Cardiac Vascular Intensive Care Unit (CVICU)
10   10002114 27793700 34672098 Coronary Care Unit (CCU)
   last_careunit                                     intime
   <chr>                                             <dttm>
 1 Medical Intensive Care Unit (MICU)                2180-07-23 14:00:00
 2 Medical Intensive Care Unit (MICU)                2150-11-02 19:37:00
 3 Medical Intensive Care Unit (MICU)                2189-06-27 08:42:00
 4 Surgical Intensive Care Unit (SICU)               2157-11-20 19:18:02
 5 Surgical Intensive Care Unit (SICU)               2157-12-19 15:42:24
 6 Medical/Surgical Intensive Care Unit (MICU/SICU) 2110-04-11 15:52:22
 7 Medical/Surgical Intensive Care Unit (MICU/SICU) 2134-12-05 18:50:03
 8 Medical Intensive Care Unit (MICU)                2131-01-11 04:20:05
 9 Cardiac Vascular Intensive Care Unit (CVICU)      2160-05-18 10:00:53
10 Coronary Care Unit (CCU)                          2162-02-17 23:30:00
   outtime             los
   <dttm>              <dbl>
 1 2180-07-23 23:50:47 0.410
 2 2150-11-06 17:03:17 3.89
 3 2189-06-27 20:38:27 0.498
 4 2157-11-21 22:08:00 1.12
 5 2157-12-20 14:27:41 0.948
 6 2110-04-12 23:59:56 1.34
 7 2134-12-06 14:38:26 0.825
 8 2131-01-20 08:27:30 9.17
```

```
 9 2160-05-19 17:33:33 1.31
10 2162-02-20 21:16:27 2.91
# i more rows
```

Q1.3 `admissions` **data**

Connect to the `admissions` table.

```
# # TODO
admissions_tble <- tbl(con_bq, "admissions") |>
  arrange(subject_id, hadm_id) |>
  # show_query() |>
  print(width = Inf)
```

```
# Source:      SQL [?? x 16]
# Database:    BigQueryConnection
# Ordered by: subject_id, hadm_id
   subject_id  hadm_id admittime           dischtime           deathtime
        <int>    <int> <dttm>              <dttm>              <dttm>
 1   10000032 22595853 2180-05-06 22:23:00 2180-05-07 17:15:00 NA
 2   10000032 22841357 2180-06-26 18:27:00 2180-06-27 18:49:00 NA
 3   10000032 25742920 2180-08-05 23:44:00 2180-08-07 17:50:00 NA
 4   10000032 29079034 2180-07-23 12:35:00 2180-07-25 17:55:00 NA
 5   10000068 25022803 2160-03-03 23:16:00 2160-03-04 06:26:00 NA
 6   10000084 23052089 2160-11-21 01:56:00 2160-11-25 14:52:00 NA
 7   10000084 29888819 2160-12-28 05:11:00 2160-12-28 16:07:00 NA
 8   10000108 27250926 2163-09-27 23:17:00 2163-09-28 09:04:00 NA
 9   10000117 22927623 2181-11-15 02:05:00 2181-11-15 14:52:00 NA
10   10000117 27988844 2183-09-18 18:10:00 2183-09-21 16:30:00 NA
   admission_type   admit_provider_id admission_location      discharge_location
   <chr>            <chr>             <chr>                   <chr>
 1 URGENT           P49AFC            TRANSFER FROM HOSPITAL  HOME
 2 EW EMER.         P784FA            EMERGENCY ROOM          HOME
 3 EW EMER.         P19UTS            EMERGENCY ROOM          HOSPICE
 4 EW EMER.         P06OTX            EMERGENCY ROOM          HOME
 5 EU OBSERVATION   P39NWO            EMERGENCY ROOM          <NA>
 6 EW EMER.         P42H7G            WALK-IN/SELF REFERRAL   HOME HEALTH CARE
 7 EU OBSERVATION   P35NE4            PHYSICIAN REFERRAL      <NA>
 8 EU OBSERVATION   P40JML            EMERGENCY ROOM          <NA>
 9 EU OBSERVATION   P47EY8            EMERGENCY ROOM          <NA>
10 OBSERVATION ADMIT P13ACE           WALK-IN/SELF REFERRAL   HOME HEALTH CARE
   insurance language marital_status race  edregtime
   <chr>     <chr>    <chr>          <chr> <dttm>
 1 Medicaid  English  WIDOWED        WHITE 2180-05-06 19:17:00
 2 Medicaid  English  WIDOWED        WHITE 2180-06-26 15:54:00
 3 Medicaid  English  WIDOWED        WHITE 2180-08-05 20:58:00
 4 Medicaid  English  WIDOWED        WHITE 2180-07-23 05:54:00
 5 <NA>      English  SINGLE         WHITE 2160-03-03 21:55:00
 6 Medicare  English  MARRIED        WHITE 2160-11-20 20:36:00
 7 Medicare  English  MARRIED        WHITE 2160-12-27 18:32:00
```

```
  8 <NA>        English  SINGLE         WHITE 2163-09-27 16:18:00
  9 Medicaid  English  DIVORCED       WHITE 2181-11-14 21:51:00
 10 Medicaid  English  DIVORCED       WHITE 2183-09-18 08:41:00
    edouttime              hospital_expire_flag
    <dttm>                              <int>
  1 2180-05-06 23:30:00                     0
  2 2180-06-26 21:31:00                     0
  3 2180-08-06 01:44:00                     0
  4 2180-07-23 14:00:00                     0
  5 2160-03-04 06:26:00                     0
  6 2160-11-21 03:20:00                     0
  7 2160-12-28 16:07:00                     0
  8 2163-09-28 09:04:00                     0
  9 2181-11-15 09:57:00                     0
 10 2183-09-18 20:20:00                     0
# i more rows
```

Q1.4 `patients` data

Connect to the `patients` table.

```
# # TODO
patients_tble <- tbl(con_bq, "patients") |>
  # show_query() |>
  print(width = Inf)
```

```
# Source:   table<`patients`> [?? x 6]
# Database: BigQueryConnection
   subject_id gender anchor_age anchor_year anchor_year_group dod
        <int> <chr>      <int>       <int> <chr>             <date>
 1   10078138 F             18        2110 2017 - 2019       NA
 2   10180372 M             18        2110 2008 - 2010       NA
 3   10686175 M             18        2110 2011 - 2013       NA
 4   10851602 F             18        2110 2014 - 2016       NA
 5   10902424 F             18        2110 2017 - 2019       NA
 6   11092326 M             18        2110 2008 - 2010       NA
 7   11289691 F             18        2110 2017 - 2019       NA
 8   11595073 M             18        2110 2011 - 2013       NA
 9   11739764 F             18        2110 2017 - 2019       NA
10   11776346 F             18        2110 2008 - 2010       NA
# i more rows
```

Q1.5 `labevents` data

Connect to the `labevents` table and retrieve a subset that only contain subjects who appear in `icustays_tble` and the lab items listed in HW3. Only keep the last lab measurements (by `storetime`) before the ICU stay and pivot lab items to become variables/columns. Write all steps in *one* chain of pipes. **Solution:**

```
labevents_tble <- tbl(con_bq, "labevents") |>
```

```r
# Select relevant columns
select(subject_id, itemid, storetime, valuenum) |>

# Join d_labitems table and Filter to include only items of interest
inner_join(select(tbl(con_bq, "d_labitems") %>%
filter(itemid %in% c(50912, 50971, 50983, 50902,
                     50882, 51221,  51301, 50931)) %>%
mutate(itemid = as.integer(itemid)), itemid), by = "itemid") |>

# Join icustays table
left_join(
  select(tbl(con_bq, "icustays") |>
arrange(subject_id, hadm_id, stay_id),
subject_id, stay_id, intime),
  by = c("subject_id"),
  copy = TRUE) |>

# Keep only lab items before ICU stay
filter(storetime < intime) |>

# Group by subject_id, stay_id, and itemid
group_by(subject_id, stay_id, itemid) |>

# Keep only the last lab value before ICU stay
slice_max(order_by = storetime, n = 1) |>

# Remove unnecessary columns
select(-storetime, -intime) |>

ungroup() |>

# Pivot wider to reshape data
pivot_wider(names_from = itemid, values_from = valuenum) |>

# Rename columns using mutate() instead of rename()
mutate(
  Bicarbonate = `50882`,
  Chloride = `50902`,
  Creatinine = `50912`,
  Glucose = `50931`,
  Potassium = `50971`,
  Sodium = `50983`,
  Hematocrit = `51221`,
  White_Blood_Cells = `51301`
) |>

select(-`50882`, -`50902`, -`50912`, -`50931`, -`50971`, -`50983`, -`51221`,
       -`51301`) |>

# Sort for grading purposes
arrange(subject_id, stay_id) |>
```

```
# Print SQL translation (without computing results)
show_query(labevents_tble) |>
```

```
print(labevents_tble)
```

Warning: ORDER BY is ignored in subqueries without LIMIT
ℹ Do you need to move arrange() later in the pipeline or use window_order() instead?
ORDER BY is ignored in subqueries without LIMIT
ℹ Do you need to move arrange() later in the pipeline or use window_order() instead?

```
<SQL>
SELECT
  `subject_id`,
  `stay_id`,
  MAX(IF(`itemid` = 50882, `valuenum`, NULL)) AS `Bicarbonate`,
  MAX(IF(`itemid` = 50902, `valuenum`, NULL)) AS `Chloride`,
  MAX(IF(`itemid` = 50912, `valuenum`, NULL)) AS `Creatinine`,
  MAX(IF(`itemid` = 50931, `valuenum`, NULL)) AS `Glucose`,
  MAX(IF(`itemid` = 50971, `valuenum`, NULL)) AS `Potassium`,
  MAX(IF(`itemid` = 50983, `valuenum`, NULL)) AS `Sodium`,
  MAX(IF(`itemid` = 51221, `valuenum`, NULL)) AS `Hematocrit`,
  MAX(IF(`itemid` = 51301, `valuenum`, NULL)) AS `White_Blood_Cells`
FROM (
  SELECT `subject_id`, `itemid`, `valuenum`, `stay_id`
  FROM (
    SELECT
      `q01`.*,
      RANK() OVER (PARTITION BY `subject_id`, `stay_id`, `itemid` ORDER BY `storetime` DESC) AS
`col01`
    FROM (
      SELECT
        `labevents`.`subject_id` AS `subject_id`,
        `labevents`.`itemid` AS `itemid`,
        `storetime`,
        `valuenum`,
        `stay_id`,
        `intime`
      FROM `labevents`
      INNER JOIN (
        SELECT SAFE_CAST(`itemid` AS INT64) AS `itemid`
        FROM `d_labitems`
        WHERE (`itemid` IN (50912.0, 50971.0, 50983.0, 50902.0, 50882.0, 51221.0, 51301.0,
50931.0))
      ) `...2`
        ON (`labevents`.`itemid` = `...2`.`itemid`)
      LEFT JOIN (
        SELECT `subject_id`, `stay_id`, `intime`
        FROM `icustays`
      ) `...3`
```

```
        ON (`labevents`.`subject_id` = `...3`.`subject_id`)
    ) `q01`
    WHERE (`storetime` < `intime`)
  ) `q01`
  WHERE (`col01` <= 1)
) `q01`
GROUP BY `subject_id`, `stay_id`
ORDER BY `subject_id`, `stay_id`

Warning: `...` must be empty in `format.tbl()`
ORDER BY is ignored in subqueries without LIMIT
i Do you need to move arrange() later in the pipeline or use window_order() instead?
Caused by error in `format_tbl()`:
! `...` must be empty.
X Problematic argument:
• ..1 = labevents_tble
i Did you forget to name an argument?


# Source:      SQL [?? x 10]
# Database:    BigQueryConnection
# Ordered by: subject_id, stay_id
   subject_id  stay_id Bicarbonate Chloride Creatinine Glucose Potassium Sodium
        <int>    <int>       <dbl>    <dbl>      <dbl>   <dbl>     <dbl>  <dbl>
 1   10000032 39553978          25       95        0.7     102       6.7    126
 2   10000690 37081114          26      100          1      85       4.8    137
 3   10000980 39765666          21      109        2.3      89       3.9    144
 4   10001217 34592300          30      104        0.5      87       4.1    142
 5   10001217 37067082          22      108        0.6     112       4.2    142
 6   10001725 31205490          NA       98         NA      NA       4.1    139
 7   10001843 39698942          28       97        1.3     131       3.9    138
 8   10001884 37510196          30       88        1.1     141       4.5    130
 9   10002013 39060235          24      102        0.9     288       3.5    137
10   10002114 34672098          18       NA        3.1      95       6.5    125
# i more rows
# i 2 more variables: Hematocrit <dbl>, White_Blood_Cells <dbl>
```

Q1.6 `chartevents` **data**

Connect to `chartevents` table and retrieve a subset that only contain subjects who appear in `icustays_tble` and the chart events listed in HW3. Only keep the first chart events (by `storetime`) during ICU stay and pivot chart events to become variables/columns. Write all steps in *one* chain of pipes. Similary to HW3, if a vital has multiple measurements at the first `storetime`, average them. **Soulution:**

```
chartevents_tble <- tbl(con_bq, "chartevents") |>

  # Select relevant columns
  select(subject_id, stay_id, itemid, storetime, valuenum) |>

  # Filter for selected vital signs
  filter(itemid %in% c(220045, 220179, 220180, 223761, 220210)) |>
```

```r
  # Join ICU stays table to get stay_id, intime, outtime
  left_join(
    select(tbl(con_bq, "icustays"), subject_id, stay_id, intime, outtime),
    by = c("subject_id", "stay_id"),
    copy = TRUE
  ) |>

  # Filter data within ICU stay (storetime between intime and outtime)
  filter(storetime >= intime & storetime <= outtime) |>

  # Group by subject_id, stay_id, itemid, storetime to apply the mean
  group_by(subject_id, stay_id, itemid, storetime) |>

  # Compute the average vital measurement per storetime
  mutate(valuenum = mean(valuenum, na.rm = TRUE)) |>
  ungroup() |>

  # Group again to get the first recorded vital per subject, stay, item
  group_by(subject_id, stay_id, itemid) |>
  slice_min(order_by = storetime, n = 1) |>

  # Remove unneeded columns
  select(-storetime, -intime) |>

  ungroup() |>

  # Pivot data wider to reshape by itemid
  pivot_wider(names_from = itemid, values_from = valuenum) |>

  # Rename columns dynamically inside BigQuery
  mutate(
    heart_rate = `220045`,
    systolic_non_invasive_blood_pressure = `220179`,
    diastolic_non_invasive_blood_pressure = `220180`,
    temperature_fahrenheit = `223761`,
    respiratory_rate = `220210`
  ) |>

  # Remove old numeric columns
  select(-`220045`, -`220179`, -`220180`, -`223761`, -`220210`) |>

  # Sort data
  arrange(subject_id, stay_id)

 # Print SQL query translation (without computing results)
 show_query(chartevents_tble) |>
   print(chartevents_tble)
```

<SQL>
SELECT

```
    `subject_id`,
    `stay_id`,
    `outtime`,
    MAX(IF(`itemid` = 220045, `valuenum`, NULL)) AS `heart_rate`,
    MAX(IF(`itemid` = 220179, `valuenum`, NULL)) AS `systolic_non_invasive_blood_pressure`,
    MAX(IF(`itemid` = 220180, `valuenum`, NULL)) AS `diastolic_non_invasive_blood_pressure`,
    MAX(IF(`itemid` = 223761, `valuenum`, NULL)) AS `temperature_fahrenheit`,
    MAX(IF(`itemid` = 220210, `valuenum`, NULL)) AS `respiratory_rate`
  FROM (
    SELECT `subject_id`, `stay_id`, `itemid`, `valuenum`, `outtime`
    FROM (
      SELECT
        `q01`.*,
        RANK() OVER (PARTITION BY `subject_id`, `stay_id`, `itemid` ORDER BY `storetime`) AS
`col01`
      FROM (
        SELECT
          `subject_id`,
          `stay_id`,
          `itemid`,
          `storetime`,
          AVG(`valuenum`) OVER (PARTITION BY `subject_id`, `stay_id`, `itemid`, `storetime`) AS
`valuenum`,
          `intime`,
          `outtime`
        FROM (
          SELECT `LHS`.*, `intime`, `outtime`
          FROM (
            SELECT `subject_id`, `stay_id`, `itemid`, `storetime`, `valuenum`
            FROM `chartevents`
            WHERE (`itemid` IN (220045.0, 220179.0, 220180.0, 223761.0, 220210.0))
          ) `LHS`
          LEFT JOIN `icustays`
            ON (
              `LHS`.`subject_id` = `icustays`.`subject_id` AND
              `LHS`.`stay_id` = `icustays`.`stay_id`
            )
        ) `q01`
        WHERE (`storetime` >= `intime` AND `storetime` <= `outtime`)
      ) `q01`
    ) `q01`
    WHERE (`col01` <= 1)
  ) `q01`
  GROUP BY `subject_id`, `stay_id`, `outtime`
  ORDER BY `subject_id`, `stay_id`
  Warning: `...` must be empty in `format.tbl()`
  Caused by error in `format_tbl()`:
  ! `...` must be empty.
  ✗ Problematic argument:
```

- ..1 = chartevents_tble

ℹ Did you forget to name an argument?

```
# Source:      SQL [?? x 8]
# Database:    BigQueryConnection
# Ordered by: subject_id, stay_id
   subject_id  stay_id outtime               heart_rate systolic_non_invasive_bl…¹
        <int>    <int> <dttm>                     <dbl>                      <dbl>
 1   10000032 39553978 2180-07-23 23:50:47          91                         84
 2   10000690 37081114 2150-11-06 17:03:17          78                        106
 3   10000980 39765666 2189-06-27 20:38:27          76                        154
 4   10001217 34592300 2157-12-20 14:27:41        79.3                        156
 5   10001217 37067082 2157-11-21 22:08:00          86                        151
 6   10001725 31205490 2110-04-12 23:59:56          86                         73
 7   10001843 39698942 2134-12-06 14:38:26        124.                        110
 8   10001884 37510196 2131-01-20 08:27:30          49                       174.
 9   10002013 39060235 2160-05-19 17:33:33          80                       98.5
10   10002114 34672098 2162-02-20 21:16:27        110.                        112
# ℹ more rows
# ℹ abbreviated name: ¹systolic_non_invasive_blood_pressure
# ℹ 3 more variables: diastolic_non_invasive_blood_pressure <dbl>,
#   temperature_fahrenheit <dbl>, respiratory_rate <dbl>
```

Q1.7 Put things together

This step is similar to Q7 of HW3. Using *one* chain of pipes `|>` to perform following data wrangling steps: (i) start with the `icustays_tble`, (ii) merge in admissions and patients tables, (iii) keep adults only (age at ICU intime >= 18), (iv) merge in the labevents and chartevents tables, (v) `collect` the tibble, (vi) sort `subject_id`, `hadm_id`, `stay_id` and `print(width = Inf)`.

```r
mimic_icu_cohort <- icustays_tble %>%
  left_join(admissions_tble, by = c("subject_id", "hadm_id")) %>%
  left_join(patients_tble, by = "subject_id") %>%
  left_join(labevents_tble, by = c("subject_id", "stay_id")) %>%
  left_join(chartevents_tble, by = c("subject_id", "stay_id")) %>%
  mutate(age_intime = year(intime) - anchor_year + anchor_age) %>%
  filter(age_intime >= 18) %>%
  collect() %>%
print(mimic_icu_cohort, width = Inf)
```

```
Warning: ORDER BY is ignored in subqueries without LIMIT
ℹ Do you need to move arrange() later in the pipeline or use window_order() instead?
ORDER BY is ignored in subqueries without LIMIT
ℹ Do you need to move arrange() later in the pipeline or use window_order() instead?
ORDER BY is ignored in subqueries without LIMIT
ℹ Do you need to move arrange() later in the pipeline or use window_order() instead?
ORDER BY is ignored in subqueries without LIMIT
ℹ Do you need to move arrange() later in the pipeline or use window_order() instead?
ORDER BY is ignored in subqueries without LIMIT
ℹ Do you need to move arrange() later in the pipeline or use window_order() instead?
```

```
Warning: `...` must be empty in `format.tbl()`
Caused by error in `format_tbl()`:
! `...` must be empty.
✗ Problematic argument:
• ..1 = mimic_icu_cohort
ℹ Did you forget to name an argument?
```

```
# A tibble: 94,458 × 42
   subject_id  hadm_id   stay_id first_careunit last_careunit intime
        <int>    <int>     <int> <chr>          <chr>         <dttm>
 1   10270110 20171261 35854639 PACU           PACU          2134-03-25 03:32:02
 2   10270110 20171261 36372959 PACU           PACU          2134-03-24 01:31:39
 3   10270644 20019675 35548343 PACU           PACU          2159-12-03 16:20:31
 4   10368426 21588639 39194905 PACU           PACU          2164-12-30 13:29:21
 5   10464753 28216499 32421516 PACU           PACU          2183-01-10 20:51:04
 6   10640410 25898987 34344828 PACU           PACU          2112-02-03 12:55:23
 7   10691194 24438843 37799251 PACU           PACU          2147-06-01 17:38:48
 8   10710188 21362776 34067486 PACU           PACU          2147-06-22 11:48:40
 9   10710188 21362776 36638120 PACU           PACU          2147-05-28 16:18:40
10   10826759 28468289 37075137 PACU           PACU          2121-05-19 18:07:00
   outtime_x                   los admittime           dischtime
   <dttm>                    <dbl> <dttm>              <dttm>
 1 2134-03-25 14:20:42      0.450  2134-03-22 04:57:00 2134-04-26 14:17:00
 2 2134-03-25 03:31:52      1.08   2134-03-22 04:57:00 2134-04-26 14:17:00
 3 2159-12-08 17:28:42      5.05   2159-12-03 01:17:00 2159-12-28 17:30:00
 4 2164-12-30 14:00:38      0.0217 2164-12-26 15:39:00 2165-01-03 16:30:00
 5 2183-01-11 22:58:45      1.09   2182-12-27 19:24:00 2183-01-27 17:39:00
 6 2112-02-08 15:14:54      5.10   2112-02-03 12:54:00 2112-02-19 18:00:00
 7 2147-06-01 17:58:44      0.0138 2147-04-25 08:30:00 2147-06-11 15:22:00
 8 2147-06-23 11:35:59      0.991  2147-05-28 16:17:00 2147-06-23 14:21:00
 9 2147-06-22 11:48:30     24.8    2147-05-28 16:17:00 2147-06-23 14:21:00
10 2121-05-20 16:32:39      0.934  2121-05-19 17:00:00 2121-05-24 12:30:00
   deathtime           admission_type   admit_provider_id
   <dttm>              <chr>            <chr>
 1 NA                  EW EMER.         P44KDZ
 2 NA                  EW EMER.         P44KDZ
 3 NA                  EW EMER.         P68D28
 4 NA                  EW EMER.         P46834
 5 NA                  OBSERVATION ADMIT P411FD
 6 NA                  OBSERVATION ADMIT P55X3P
 7 NA                  ELECTIVE         P93BYT
 8 2147-06-23 14:21:00 EW EMER.         P502T3
 9 2147-06-23 14:21:00 EW EMER.         P502T3
10 NA                  EW EMER.         P20PIB
   admission_location              discharge_location       insurance
   <chr>                           <chr>                    <chr>
 1 TRANSFER FROM HOSPITAL          HOSPICE                  Medicaid
 2 TRANSFER FROM HOSPITAL          HOSPICE                  Medicaid
 3 PHYSICIAN REFERRAL              SKILLED NURSING FACILITY Medicare
 4 WALK-IN/SELF REFERRAL           SKILLED NURSING FACILITY Medicare
```

```
 5 TRANSFER FROM HOSPITAL              HOSPICE                  Medicare
 6 CLINIC REFERRAL                     HOME HEALTH CARE         Private
 7 PHYSICIAN REFERRAL                  SKILLED NURSING FACILITY Medicare
 8 TRANSFER FROM SKILLED NURSING FACILITY DIED                 Medicare
 9 TRANSFER FROM SKILLED NURSING FACILITY DIED                 Medicare
10 TRANSFER FROM HOSPITAL              REHAB                    Medicare
   language marital_status race                  edregtime
   <chr>    <chr>          <chr>                 <dttm>
 1 English  MARRIED        WHITE                 2134-03-22 01:01:00
 2 English  MARRIED        WHITE                 2134-03-22 01:01:00
 3 English  DIVORCED       WHITE                 2159-12-02 19:45:00
 4 English  WIDOWED        WHITE                 2164-12-26 08:22:00
 5 English  MARRIED        UNABLE TO OBTAIN      2182-12-27 18:59:00
 6 English  MARRIED        BLACK/AFRICAN         2112-02-03 08:05:00
 7 English  WIDOWED        WHITE                 NA
 8 English  MARRIED        WHITE - OTHER EUROPEAN 2147-05-28 11:58:00
 9 English  MARRIED        WHITE - OTHER EUROPEAN 2147-05-28 11:58:00
10 English  SINGLE         WHITE - BRAZILIAN     2121-05-19 08:03:00
   edouttime           hospital_expire_flag gender anchor_age anchor_year
   <dttm>                             <int> <chr>       <int>      <int>
 1 2134-03-22 07:40:00                    0 M             78       2134
 2 2134-03-22 07:40:00                    0 M             78       2134
 3 2159-12-03 02:51:00                    0 F             84       2152
 4 2164-12-26 21:43:00                    0 M             80       2154
 5 2182-12-27 21:24:00                    0 M             86       2182
 6 2112-02-03 14:15:00                    0 F             44       2112
 7 NA                                     0 F             74       2144
 8 2147-05-28 18:23:00                    1 M             86       2147
 9 2147-05-28 18:23:00                    1 M             86       2147
10 2121-05-19 18:07:00                    0 F             77       2121
   anchor_year_group dod        Bicarbonate Chloride Creatinine Glucose
   <chr>             <date>           <dbl>    <dbl>      <dbl>   <dbl>
 1 2020 - 2022       2134-04-30          23      105          1     178
 2 2020 - 2022       2134-04-30          24      104        0.8      98
 3 2014 - 2016       2160-06-25          20      108        0.5      75
 4 2011 - 2013       2165-03-18          24      109        0.8     131
 5 2020 - 2022       2183-01-28          22      107        1.2     111
 6 2017 - 2019       NA                  21       96       10.3     102
 7 2017 - 2019       2147-09-16          24       94          5      95
 8 2020 - 2022       2147-06-23          31      107        0.6     173
 9 2020 - 2022       2147-06-23          20      110        4.1     120
10 2020 - 2022       NA                  NA       NA         NA      NA
   Potassium Sodium Hematocrit White_Blood_Cells outtime_y           heart_rate
       <dbl>  <dbl>      <dbl>             <dbl> <dttm>                    <dbl>
 1       3.8    136       19.6              13.3 2134-03-25 14:20:42          86
 2         4    137       24.8              42.1 2134-03-25 03:31:52         101
 3       3.8    145       31.4               9.2 2159-12-08 17:28:42          55
 4         4    138       29.4               4.8 NA                          NA
 5       3.6    140       31                12.8 2183-01-11 22:58:45          96
 6       5.4    137       30.5              11.2 2112-02-08 15:14:54          97
 7         4    135       26.2               8.7 NA                          NA
```

8	5	144	29.6	10.8	2147-06-23	11:35:59	92
9	5.2	150	47.5	17.8	2147-06-22	11:48:30	62
10	NA	NA	37.4	8.3	2121-05-20	16:32:39	77

	systolic_non_invasive_blood_pressure	diastolic_non_invasive_blood_pressure
	<dbl>	<dbl>
1	NA	NA
2	NA	NA
3	92	62
4	NA	NA
5	106	63
6	173	107
7	NA	NA
8	89	61
9	104	61
10	116	58

	temperature_fahrenheit	respiratory_rate	age_intime
	<dbl>	<dbl>	<int>
1	97.7	14	78
2	96.7	15	78
3	97.9	16	91
4	NA	NA	90
5	97.3	20	87
6	97.8	22	44
7	NA	NA	77
8	101.	23	86
9	98.3	23	86
10	98.5	16.5	77

i 94,448 more rows

Q1.8 Preprocessing

Perform the following preprocessing steps. (i) Lump infrequent levels into "Other" level for `first_careunit`, `last_careunit`, `admission_type`, `admission_location`, and `discharge_location`. (ii) Collapse the levels of `race` into `ASIAN`, `BLACK`, `HISPANIC`, `WHITE`, and `Other`. (iii) Create a new variable `los_long` that is `TRUE` when `los` is greater than or equal to 2 days. (iv) Summarize the data using `tbl_summary()`, stratified by `los_long`. Hint: `fct_lump_n` and `fct_collapse` from the `forcats` package are useful.

Hint: Below is a numerical summary of my tibble after preprocessing:

Characteristic	TRUE N = 46,337[1]	FALSE N = 48,107[1]
first_careunit		
Cardiac Vascular Intensive Care Unit (CVICU)	7,353 (16%)	7,416 (15%)
Medical Intensive Care Unit (MICU)	9,837 (21%)	10,862 (23%)
Medical/Surgical Intensive Care Unit (MICU/SICU)	6,667 (14%)	8,780 (18%)
Surgical Intensive Care Unit (SICU)	6,434 (14%)	6,574 (14%)
Other	16,046 (35%)	14,475 (30%)
last_careunit		
Cardiac Vascular Intensive Care Unit (CVICU)	7,353 (16%)	7,416 (15%)
Medical Intensive Care Unit (MICU)	9,837 (21%)	10,862 (23%)
Medical/Surgical Intensive Care Unit		

Solution:*

```
# Inspect unique levels of the 'race' variable
unique_race_levels <- mimic_icu_cohort %>%
  pull(race) %>%
  unique()

print(unique_race_levels)
```

```
 [1] "WHITE"
 [2] "UNABLE TO OBTAIN"
 [3] "BLACK/AFRICAN"
 [4] "WHITE - OTHER EUROPEAN"
 [5] "WHITE - BRAZILIAN"
 [6] "UNKNOWN"
 [7] "BLACK/AFRICAN AMERICAN"
 [8] "PORTUGUESE"
 [9] "OTHER"
[10] "ASIAN"
[11] "WHITE - RUSSIAN"
[12] "HISPANIC/LATINO - DOMINICAN"
[13] "BLACK/CARIBBEAN ISLAND"
[14] "HISPANIC/LATINO - PUERTO RICAN"
```

```
[15] "ASIAN - ASIAN INDIAN"
[16] "PATIENT DECLINED TO ANSWER"
[17] "AMERICAN INDIAN/ALASKA NATIVE"
[18] "HISPANIC/LATINO - SALVADORAN"
[19] "HISPANIC/LATINO - CENTRAL AMERICAN"
[20] "WHITE - EASTERN EUROPEAN"
[21] "ASIAN - SOUTH EAST ASIAN"
[22] "ASIAN - CHINESE"
[23] "HISPANIC/LATINO - GUATEMALAN"
[24] "NATIVE HAWAIIAN OR OTHER PACIFIC ISLANDER"
[25] "BLACK/CAPE VERDEAN"
[26] "HISPANIC/LATINO - MEXICAN"
[27] "HISPANIC OR LATINO"
[28] "HISPANIC/LATINO - HONDURAN"
[29] "SOUTH AMERICAN"
[30] "HISPANIC/LATINO - COLUMBIAN"
[31] "ASIAN - KOREAN"
[32] "HISPANIC/LATINO - CUBAN"
[33] "MULTIPLE RACE/ETHNICITY"
```

```r
mimic_icu_cohort <- mimic_icu_cohort %>%
  mutate(race = toupper(trimws(race)))

mimic_icu_cohort <- mimic_icu_cohort %>%
  mutate(
    race = fct_collapse(race,
      ASIAN = c("ASIAN", "ASIAN - ASIAN INDIAN", "ASIAN - SOUTH EAST ASIAN",
                "ASIAN - CHINESE", "ASIAN - KOREAN"),

      BLACK = c("BLACK/AFRICAN", "BLACK/AFRICAN AMERICAN",
                "BLACK/CARIBBEAN ISLAND", "BLACK/CAPE VERDEAN"),

      HISPANIC = c("HISPANIC OR LATINO", "HISPANIC/LATINO - DOMINICAN",
                   "HISPANIC/LATINO - PUERTO RICAN",
                   "HISPANIC/LATINO - SALVADORAN",
                   "HISPANIC/LATINO - CENTRAL AMERICAN",
                   "HISPANIC/LATINO - GUATEMALAN",
                   "HISPANIC/LATINO - MEXICAN", "HISPANIC/LATINO - HONDURAN",
                   "HISPANIC/LATINO - COLUMBIAN", "HISPANIC/LATINO - CUBAN"),
      WHITE = c("WHITE", "WHITE - OTHER EUROPEAN", "WHITE - BRAZILIAN",
                "WHITE - RUSSIAN", "WHITE - EASTERN EUROPEAN", "PORTUGUESE"),

      Other = c("UNKNOWN", "OTHER", "UNABLE TO OBTAIN",
                "PATIENT DECLINED TO ANSWER", "AMERICAN INDIAN/ALASKA NATIVE",
                "NATIVE HAWAIIAN OR OTHER PACIFIC ISLANDER",
                "SOUTH AMERICAN", "MULTIPLE RACE/ETHNICITY")
    )
  ) %>%

  # Lump infrequent levels for specified categorical variables
```

```
  mutate(across(c(first_careunit, last_careunit, admission_type, admission_location, discharge_lo
                ~ fct_lump_n(.x, n = 5, other_level = "Other"))) %>%

  # Create los_long variable
  mutate(los_long = los >= 2)

# Generate summary table stratified by 'los_long'
summary_table <- mimic_icu_cohort %>%
  select(first_careunit, last_careunit, los, admission_type, admission_location,
         discharge_location, insurance, language, marital_status, race,
         hospital_expire_flag, gender, dod, Chloride, Creatinine, Sodium,
         Potassium, Glucose, Hematocrit, White_Blood_Cells, Bicarbonate,
         systolic_non_invasive_blood_pressure,
         diastolic_non_invasive_blood_pressure, respiratory_rate,
         temperature_fahrenheit, heart_rate, age_intime, los_long) %>%
  tbl_summary(by = los_long)
```

14 missing rows in the "los_long" column have been removed.
The following errors were returned during `tbl_summary()`:
✗ For variable `dod` (`los_long = FALSE`) and "p75" statistic: * not defined
  for "Date" objects

```
summary_table %>%
  as_gt()
```

Characteristic	TRUE N = 46,337[1]	FALSE N = 48,107[1]
first_careunit		
Cardiac Vascular Intensive Care Unit (CVICU)	7,353 (16%)	7,416 (15%)
Coronary Care Unit (CCU)	5,433 (12%)	5,338 (11%)
Medical Intensive Care Unit (MICU)	9,837 (21%)	10,862 (23%)
Medical/Surgical Intensive Care Unit (MICU/SICU)	6,667 (14%)	8,780 (18%)
Surgical Intensive Care Unit (SICU)	6,434 (14%)	6,574 (14%)
Other	10,613 (23%)	9,137 (19%)
last_careunit		

[1] n (%); Median (Q1, Q3)

Characteristic	TRUE N = 46,337[1]	FALSE N = 48,107[1]
Cardiac Vascular Intensive Care Unit (CVICU)	7,353 (16%)	7,416 (15%)
Coronary Care Unit (CCU)	5,433 (12%)	5,338 (11%)
Medical Intensive Care Unit (MICU)	9,837 (21%)	10,862 (23%)
Medical/Surgical Intensive Care Unit (MICU/SICU)	6,667 (14%)	8,780 (18%)
Surgical Intensive Care Unit (SICU)	6,434 (14%)	6,574 (14%)
Other	10,613 (23%)	9,137 (19%)
los	3.9 (2.7, 6.8)	1.1 (0.8, 1.5)
admission_type		
DIRECT EMER.	1,726 (3.7%)	1,590 (3.3%)
EW EMER.	23,012 (50%)	25,337 (53%)
OBSERVATION ADMIT	7,393 (16%)	6,638 (14%)
SURGICAL SAME DAY ADMISSION	4,001 (8.6%)	5,543 (12%)
URGENT	8,691 (19%)	6,683 (14%)
Other	1,514 (3.3%)	2,316 (4.8%)
admission_location		
EMERGENCY ROOM	17,058 (37%)	20,443 (42%)
PHYSICIAN REFERRAL	11,013 (24%)	12,684 (26%)
TRANSFER FROM HOSPITAL	13,904 (30%)	10,400 (22%)
TRANSFER FROM SKILLED NURSING FACILITY	803 (1.7%)	713 (1.5%)
WALK-IN/SELF REFERRAL	2,169 (4.7%)	2,308 (4.8%)
Other	1,390 (3.0%)	1,559 (3.2%)

[1] n (%); Median (Q1, Q3)

Characteristic	TRUE N = 46,337[1]	FALSE N = 48,107[1]
discharge_location		
DIED	6,884 (15%)	4,436 (9.4%)
HOME	6,879 (15%)	15,210 (32%)
HOME HEALTH CARE	10,620 (23%)	13,422 (28%)
REHAB	5,574 (12%)	2,445 (5.2%)
SKILLED NURSING FACILITY	8,785 (19%)	7,489 (16%)
Other	7,518 (16%)	4,334 (9.2%)
Unknown	77	771
insurance		
Medicaid	6,768 (15%)	7,469 (16%)
Medicare	26,330 (58%)	25,485 (54%)
No charge	5 (<0.1%)	3 (<0.1%)
Other	1,091 (2.4%)	1,237 (2.6%)
Private	11,515 (25%)	13,018 (28%)
Unknown	628	895
language		
American Sign Language	29 (<0.1%)	34 (<0.1%)
Amharic	14 (<0.1%)	9 (<0.1%)
Arabic	87 (0.2%)	62 (0.1%)
Armenian	12 (<0.1%)	13 (<0.1%)
Bengali	22 (<0.1%)	12 (<0.1%)
Chinese	550 (1.2%)	611 (1.3%)
English	41,563 (90%)	43,483 (91%)

[1] n (%); Median (Q1, Q3)

Characteristic	TRUE N = 46,337[1]	FALSE N = 48,107[1]
French	18 (<0.1%)	14 (<0.1%)
Haitian	375 (0.8%)	252 (0.5%)
Hindi	24 (<0.1%)	21 (<0.1%)
Italian	101 (0.2%)	107 (0.2%)
Japanese	5 (<0.1%)	7 (<0.1%)
Kabuverdianu	301 (0.7%)	345 (0.7%)
Khmer	50 (0.1%)	37 (<0.1%)
Korean	40 (<0.1%)	32 (<0.1%)
Modern Greek (1453-)	102 (0.2%)	88 (0.2%)
Other	152 (0.3%)	153 (0.3%)
Persian	42 (<0.1%)	35 (<0.1%)
Polish	36 (<0.1%)	38 (<0.1%)
Portuguese	351 (0.8%)	314 (0.7%)
Russian	601 (1.3%)	659 (1.4%)
Somali	8 (<0.1%)	15 (<0.1%)
Spanish	1,472 (3.2%)	1,429 (3.0%)
Thai	21 (<0.1%)	22 (<0.1%)
Vietnamese	151 (0.3%)	129 (0.3%)
Unknown	210	186
marital_status		
DIVORCED	3,377 (8.0%)	3,555 (8.0%)
MARRIED	20,557 (49%)	21,344 (48%)
SINGLE	12,745 (30%)	14,039 (31%)

[1] n (%); Median (Q1, Q3)

Characteristic	TRUE N = 46,337[1]	FALSE N = 48,107[1]
WIDOWED	5,319 (13%)	5,752 (13%)
Unknown	4,339	3,417
race		
Other	7,802 (17%)	6,689 (14%)
ASIAN	1,369 (3.0%)	1,516 (3.2%)
BLACK	4,933 (11%)	5,452 (11%)
HISPANIC	1,687 (3.6%)	1,908 (4.0%)
WHITE	30,546 (66%)	32,542 (68%)
hospital_expire_flag	6,831 (15%)	4,512 (9.4%)
gender		
F	20,106 (43%)	21,471 (45%)
M	26,231 (57%)	26,636 (55%)
dod	2155-09-06 (2135-07-16, 2175-10-08)	2155-12-18 (2136-04-26, NA)
Unknown	25,846	30,639
Chloride	102 (98, 105)	102 (98, 105)
Unknown	6,184	5,167
Creatinine	1.00 (0.80, 1.60)	1.00 (0.80, 1.40)
Unknown	4,541	3,486
Sodium	138.0 (135.0, 141.0)	139.0 (136.0, 141.0)
Unknown	6,167	5,163
Potassium	4.20 (3.90, 4.70)	4.20 (3.90, 4.60)
Unknown	6,200	5,187

[1] n (%); Median (Q1, Q3)

Characteristic	TRUE N = 46,337[1]	FALSE N = 48,107[1]
Glucose	122 (100, 159)	118 (98, 154)
Unknown	6,340	5,314
Hematocrit	35 (29, 40)	36 (30, 41)
Unknown	3,857	2,894
White_Blood_Cells	9.7 (7.0, 13.8)	9.0 (6.6, 12.6)
Unknown	3,906	2,944
Bicarbonate	24.0 (21.0, 27.0)	24.0 (21.0, 27.0)
Unknown	6,272	5,277
systolic_non_invasive_blood_pressure	119 (104, 137)	122 (107, 138)
Unknown	348	1,022
diastolic_non_invasive_blood_pressure	67 (57, 79)	68 (58, 80)
Unknown	351	1,024
respiratory_rate	19.0 (16.0, 23.0)	18.0 (15.0, 22.0)
Unknown	15	183
temperature_fahrenheit	98.20 (97.70, 98.80)	98.10 (97.60, 98.60)
Unknown	231	1,444
heart_rate	87 (75, 102)	84 (73, 99)
Unknown	1	85
age_intime	67 (56, 77)	66 (54, 77)

[1] n (%); Median (Q1, Q3)

Q1.9 Save the final tibble

Solution: Save the final tibble to an R data file `mimic_icu_cohort.rds` in the `mimiciv_shiny` folder.

```
# make a directory mimiciv_shiny
if (!dir.exists("mimiciv_shiny")) {
```

```
    dir.create("mimiciv_shiny")
}
# save the final tibble
mimic_icu_cohort |>
    write_rds("mimiciv_shiny/mimic_icu_cohort.rds", compress = "gz")
```

Done.

Close database connection and clear workspace.

```
if (exists("con_bq")) {
    dbDisconnect(con_bq)
}
rm(list = ls())
```

Done.

Although it is not a good practice to add big data files to Git, for grading purpose, please add `mimic_icu_cohort.rds` to your Git repository.

Q2. Shiny app

Develop a Shiny app for exploring the ICU cohort data created in Q1. The app should reside in the `mimiciv_shiny` folder. The app should contain at least two tabs. One tab provides easy access to the graphical and numerical summaries of variables (demographics, lab measurements, vitals) in the ICU cohort, using the `mimic_icu_cohort.rds` you curated in Q1. The other tab allows user to choose a specific patient in the cohort and display the patient's ADT and ICU stay information as we did in Q1 of HW3, by dynamically retrieving the patient's ADT and ICU stay information from BigQuery database. Again, do **not** ever add the BigQuery token to your Git repository. If you do so, you will lose 50 points.

Solution: The detailed code for shiny app can be found in the file app.R.