

# Assessment R Markdown

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16/06/2022

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## Loading the required packages required for this script

```
library(dataMeta)
library(tidyverse)
library(here)
library(NHSRdatasets)
library(knitr)
library(scales)
library(lubridate)
library(caret)
```

## Loading NHS Datasets

### Data

The **NHS England accident and emergency attendances and admissions (ae\_attendances)** dataset will be loaded from the NHSRdatasets package. This dataset reports on the attendances, four-hour breaches and admissions for all A&E departments in England for the years 2016/17 through 2018/19 (Apr-Mar)

### Loading ae\_attendances data

ae\_attendances data is loaded from the NHSdatasets package

```
data(ae_attendances)
ae<-ae_attendances
```

### Viewing the data

The `head()` function is used to look at top n rows of a data frame (default n = 6 rows)

```
head(ae)
```

```
## # A tibble: 6 x 6
##   period      org_code type attendances breaches admissions
##   <date>      <fct>   <fct>      <dbl>      <dbl>      <dbl>
## 1 2017-03-01 RF4      1         21289      2879      5060
## 2 2017-03-01 RF4      2           813        22         0
## 3 2017-03-01 RF4     other      2850         6         0
## 4 2017-03-01 R1H      1        30210      5902      6943
## 5 2017-03-01 R1H      2          807        11         0
## 6 2017-03-01 R1H     other     11352       136         0
```

## Check for missing data

Calculate how many NAs there are in each variable

```
ae %>%
  map(is.na) %>%
  map(sum)
```

```
## $period
## [1] 0
##
## $org_code
## [1] 0
##
## $type
## [1] 0
##
## $attendances
## [1] 0
##
## $breaches
## [1] 0
##
## $admissions
## [1] 0
```

Data is complete

## Add an index link column to ae\_attendances data

The `rowid_to_column()` function is used to convert row identities to a column named `index`. This facilitates the linking of partitioned datasets to the raw data if required in the future.

```
ae <- rowid_to_column(ae, "index")
```

## Preview dataset

Tabulate data, and create report with `kable()`

```
ae %>%
  mutate_at(vars(period), format, "%b-%y") %>%
  mutate_at(vars(attendances, breaches, admissions), comma) %>%
  head(10) %>%
  kable()
```

index	period	org_code	type	attendances	breaches	admissions
1	Mar-17	RF4	1	21,289.0	2,879.0	5,060.0
2	Mar-17	RF4	2	813.0	22.0	0.0
3	Mar-17	RF4	other	2,850.0	6.0	0.0
4	Mar-17	R1H	1	30,210.0	5,902.0	6,943.0
5	Mar-17	R1H	2	807.0	11.0	0.0
6	Mar-17	R1H	other	11,352.0	136.0	0.0
7	Mar-17	AD913	other	4,381.0	2.0	0.0
8	Mar-17	RYX	other	19,562.0	258.0	0.0
9	Mar-17	RQM	1	17,414.0	2,030.0	3,597.0
10	Mar-17	RQM	other	7,817.0	86.0	0.0

## Save the raw ae\_attendances data to your 'RawData' folder

```
write_csv(ae, here("RawData", "ae_attendances.csv"))
```

## Subsetting the data

The aim of my data capture tool is to help identify performance outliers with regard to the 4hr breaches. In order to do this effectively I will need use the entire dataset (i.e. the variables: index, period, attendances, breaches, organisation name, and organisation type).

Of note the **performance** variable was deliberately not created as this is a function of the **attendances** and **breaches** variables. The variable can be created after the collection of data at the data analysis stage.

## Creating test and training datasets

Calculate the proportion (prop) of the raw data to assign to the training data.

The proportion of the raw that needs to be assigned to the training data to ensure there is only 10 to 15 records in the test data is:

```
prop<-(1-(15/nrow(ae)))
print(prop)
```

```
## [1] 0.9988249
```

## Splitting the raw data

The `createDataPartition()` function from the *caret* package is used to split the raw data into test and training data sets.

The `set.seed()` function allows a random number to be generated in a reproducible fashion. This ensures that every time the script is run the raw data will be partitioned into the same test and training datasets.

```
set.seed(333)
trainIndex <- createDataPartition(ae$index, p = prop,
                                   list = FALSE,
                                   times = 1)
head(trainIndex)
```

```
##      Resample1
## [1,]         1
## [2,]         2
## [3,]         3
## [4,]         4
## [5,]         5
## [6,]         6
```

Assign all records in the `trainIndex` to the training data

```
aeTrain <- ae[ trainIndex,]
nrow(aeTrain)
```

```
## [1] 12753
```

Save the training data to the to working data folder 'Data' as `ae_attendances_train.csv`

```
write_csv(aeTrain, here("Data", "ae_attendances_train.csv"))
```

Extract the test data (all records that are not in the `trainIndex` (`-trainIndex`) are assigned to the test data).

```
aeTest <- ae[-trainIndex,]
nrow(aeTest)
```

```
## [1] 12
```

Set aside the first record for markers to test and evaluate the data-capture tool.

```
aeTestMarker <- aeTest[1,]
```

Save the marker test data to the to working data folder 'Data' as `ae_attendances_test_marker.csv`

```
write_csv(aeTestMarker, here("Data", "ae_attendances_test_marker.csv"))
```

The remaining records are set aside to test with the data capture tool. These are saved to the to working data folder 'Data' as `ae_attendances_test.csv`

```
aeTest <- aeTest[2:nrow(aeTest),]
write_csv(aeTest, here("Data", "ae_attendances_test.csv"))
```

## Data Dictionary for test data

### Reading the collected data

`read_csv()` function from the *readr* package is used to read my collected data from the Raw data folder.

```
CollectedData=read_csv(here("RawData", "CollectedDataFinal.csv"))
```

```
## Rows: 11 Columns: 8-- Column specification -----
## Delimiter: ","
## chr  (2): org_code, type
## dbl  (4): index, attendances, breaches, admissions
## lgl  (1): consent
## date (1): period
## 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.
```

### Building a linker data frame

#### Variable descriptions

String vectors are created for the different variable descriptions

```
variable_description <- c("The index column that allows us to link the data collected to the original ae_attendances data.",
"The month that this activity relates to, stored as a date (1st of each month).",
"The Organisation data service (ODS) code for the organisation. If you want to know the organisation as a whole.",
"The department type for this activity.",
"The number of attendances for this department type at this organisation for this month.",
"The number of attendances that breached the four-hour target.",
"The number of attendances that resulted in an admission to the hospital.",
"The consent from the end-user to process and share the data collected with the data capture tool.")
print(variable_description)
```

```
## [1] "The index column that allows us to link the data collected to the original ae_attendances data."
## [2] "The month that this activity relates to, stored as a date (1st of each month)."
```

```
## [3] "The Organisation data service (ODS) code for the organisation. If you want to know the organisation as a whole."
## [4] "The department type for this activity."
## [5] "The number of attendances for this department type at this organisation for this month."
## [6] "The number of attendances that breached the four-hour target."
## [7] "The number of attendances that resulted in an admission to the hospital."
## [8] "The consent from the end-user to process and share the data collected with the data capture tool."
```

#### Variable types

A string vector is created representing the different variable types. It is a vector of integers with values 0 or 1. 0 is used for a variable with quantitative values (measured values) variables and 1 for fixed values (allowable values or codes) variables.

`glimpse()` function from *tibble* package is used to view the variable types in the `CollectedData` data frame

```
glimpse(CollectedData)
```

```
## Rows: 11
## Columns: 8
## $ index      <dbl> 1155, 2059, 3468, 4153, 4820, 7243, 8057, 8957, 10214, 103~
## $ period     <date> 2016-12-01, 2016-10-01, 2016-05-01, 2018-03-01, 2018-02-0~
## $ org_code   <chr> "C82010", "RDZ", "RVR", "RQM", "R1F", "RE9", "RQM", "RNL", ~
## $ type       <chr> "other", "1", "2", "other", "other", "1", "1", "1", "other~
## $ attendances <dbl> 200, 6452, 417, 9376, 245, 5170, 15957, 7258, 3197, 2033, ~
## $ breaches  <dbl> 0, 360, 0, 112, 0, 235, 1309, 1374, 0, 8, 1
## $ admissions <dbl> 0, 1814, 6, 0, 0, 1269, 3375, 1947, 0, 105, 0
## $ consent    <lgl> TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE, TRUE~
```

This indicated that there are four quantitative values (measured values) variables and four fixed values (allowable values or codes) variables.

```
variable_type <- c(0, 1, 1, 1, 0, 0, 0, 1)
print(variable_type)
```

```
## [1] 0 1 1 1 0 0 0 1
```

## Building the linker

`build_linker()` function from the *dataMeta* package is used to construct an intermediary (linker) data frame between the `CollectedData` and the data dictionary.

```
linker<-build_linker(CollectedData, variable_description, variable_type)
print(linker)
```

```
##      var_name
## 1      index
## 2     period
## 3    org_code
## 4      type
## 5 attendances
## 6    breaches
## 7  admissions
## 8      consent
##
## 1
## 2
## 3 The Organisation data service (ODS) code for the organisation. If you want to know the organisation
## 4
## 5
## 6
## 7
## 8
##      var_type
## 1          0
```

```
## 2      1
## 3      1
## 4      1
## 5      0
## 6      0
## 7      0
## 8      1
```

## Data dictionary

the `build_dict()` function from the *dataMeta* package is used to construct a data dictionary for a `CollectedData` data frame with the aid of the linker data frame.

```
dictionary <- build_dict(my.data = CollectedData, linker = linker, option_description = NULL,
prompt_varopts = FALSE)
```

```
dictionary[6,4]<-"C82010: Prescribing Cost Centre - OAKHAM MEDICAL PRACTICE"
dictionary[7,4]<-"RDZ: NHS Trust - THE ROYAL BOURNEMOUTH AND CHRISTCHURCH HOSPITALS NHS FOUNDATION TRUST"
dictionary[8,4]<-"RVR: NHS Trust - EPSOM AND ST HELIER UNIVERSITY HOSPITALS NHS TRUST"
dictionary[9,4]<-"RQM: NHS Trust - CHELSEA AND WESTMINSTER HOSPITAL NHS FOUNDATION TRUST"
dictionary[10,4]<-"R1F: NHS Trust - ISLE OF WIGHT NHS TRUST"
dictionary[11,4]<-"RE9: NHS Trust - SOUTH TYNESIDE NHS FOUNDATION TRUST"
dictionary[12,4]<-"RNL: NHS Trust - NORTH CUMBRIA UNIVERSITY HOSPITALS NHS TRUST"
dictionary[13,4]<-"RJ1: NHS Trust - GUY'S AND ST THOMAS' NHS FOUNDATION TRUST"
dictionary[14,4]<-"RKB: NHS Trust - UNIVERSITY HOSPITALS COVENTRY AND WARWICKSHIRE NHS TRUST"
dictionary[15,4]<-"NL012: Independent Sector H/c Provider Site - OAKHAM URGENT CARE CENTRE"
dictionary[26,4] <-"other: Other types of A&E/minor injury activity with designated accommodation for t
dictionary[27,4]<- "1: Emergency departments are a consultant-led 24-hour service with full resuscitatio
dictionary[28,4] <- "2: Consultant-led mono speciality accident and emergency service (e.g. ophthalmolog
```

Save the data dictionary for `CollectedData` to the ‘RawData’ folder

```
write_csv(dictionary, here("RawData", "CollectedData_DataDictionary.csv"))
```

## Appending the data dictionary to data

Incorporate attributes as metadata to the `CollectedData` as metadata using the `incorporate_attr()` function from the *dataMeta* package. This requires the `CollectedData`, `dictionary`, and `main_string` as inputs (`main_string` is a character string describing the `CollectedData` data frame)

Create `main_string` for attributes

```
main_string <- "This data describes the NHS England accident and emergency (A&E) attendances and breach
```

Incorporate attributes as metadata

```
complete_CollectedData <- incorporate_attr(my.data = CollectedData, data.dictionary = dictionary,
main_string = main_string)
attributes(complete_CollectedData)$author[1]<-"B203349"
complete_CollectedData
```

```
## # A tibble: 11 x 8
##   index period      org_code type attendances breaches admissions consent
## * <dbl> <date>      <chr>  <chr>      <dbl>      <dbl>      <dbl> <lgl>
## 1  1155 2016-12-01 C82010 other         200         0         0 TRUE
## 2  2059 2016-10-01 RDZ      1         6452        360        1814 TRUE
## 3  3468 2016-05-01 RVR      2         417         0         6 TRUE
## 4  4153 2018-03-01 RQM      other        9376        112         0 TRUE
## 5  4820 2018-02-01 R1F      other         245         0         0 TRUE
## 6  7243 2017-07-01 RE9      1         5170        235        1269 TRUE
## 7  8057 2017-04-01 RQM      1        15957        1309        3375 TRUE
## 8  8957 2019-02-01 RNL      1         7258        1374        1947 TRUE
## 9 10214 2018-10-01 RJ1      other         3197         0         0 TRUE
## 10 10328 2018-10-01 RKB      2         2033         8         105 TRUE
## 11 11767 2018-06-01 NL012 other         336         1         0 TRUE
```

```
attributes(complete_CollectedData)
```

```
## $row.names
## [1] 1 2 3 4 5 6 7 8 9 10 11
##
## $names
## [1] "index"      "period"      "org_code"    "type"        "attendances"
## [6] "breaches"    "admissions"  "consent"
##
## $spec
## cols(
##   index = col_double(),
##   period = col_date(format = ""),
##   org_code = col_character(),
##   type = col_character(),
##   attendances = col_double(),
##   breaches = col_double(),
##   admissions = col_double(),
##   consent = col_logical()
## )
##
## $problems
## <pointer: 0x7f9e98d79b10>
##
## $class
## [1] "spec_tbl_df" "tbl_df"      "tbl"         "data.frame"
##
## $main
## [1] "This data describes the NHS England accident and emergency (A&E) attendances and breaches of for
##
## $dictionary
##   variable_name
```



```

## 1      admissions
## 2      attendances
## 3      breaches
## 4      consent
## 5      index
## 6      org_code
## 7
## 8
## 9
## 10
## 11
## 12
## 13
## 14
## 15
## 16      period
## 17
## 18
## 19
## 20
## 21
## 22
## 23
## 24
## 25      type
## 26
## 27
## 28
##
## 1
## 2
## 3
## 4
## 5
## 6 The Organisation data service (ODS) code for the organisation. If you want to know the organisation
## 7
## 8
## 9
## 10
## 11
## 12
## 13
## 14
## 15
## 16
## 17
## 18
## 19
## 20
## 21
## 22
## 23
## 24
## 25

```

```

## 26
## 27
## 28
##   variable_options
## 1      0 to 3375
## 2    200 to 15957
## 3      0 to 1374
## 4      TRUE
## 5    1155 to 11767
## 6      C82010
## 7      RDZ
## 8      RVR
## 9      RQM
## 10     R1F
## 11     RE9
## 12     RNL
## 13     RJ1
## 14     RKB
## 15     NL012
## 16     17136
## 17     17075
## 18     16922
## 19     17591
## 20     17563
## 21     17348
## 22     17257
## 23     17928
## 24     17805
## 25     17683
## 26     other
## 27         1
## 28         2
##
## 1
## 2
## 3
## 4
## 5
## 6
## 7
## 8
## 9
## 10
## 11
## 12
## 13
## 14
## 15
## 16
## 17
## 18
## 19
## 20
## 21

```

```

## 22
## 23
## 24
## 25
## 26
## 27
## 28 2: Consultant-led mono speciality accident and emergency service (e.g. ophthalmology, dental) with
##
## $last_edit_date
## [1] "2022-06-20 12:59:27 ADT"
##
## $author
## [1] "B203349"

```

Save the CollectedData with attributes

```

save_it(complete_CollectedData, here("RawData", "complete_CollectedData"))

```

## Data Capture Tool (Python)

A data capture tool was created. The code was written in Python using the Jupyter notebook. Data was collected using an interactive graphic user interface that was built using widgets from the `ipywidgets` package. During the active collection process data was saved to the Data folder as *CollectedData*. Once all data was collected the final dataset was saved to the RawData folder as *CollectedDataFinal*.