

# B518 | Week 4 | Group Project | Submission 1

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2025-09-30

## 1 Project Idea One - Covid 19 (2021 ONLY - USA, UK, China, Belgium)

1.1 1) Original Source: URL: <https://docs.owid.io/projects/covid/en/latest/dataset.html>

1.2 2) Dataset moved: <https://catalog.ourworldindata.org/garden/covid/latest/compact/compact.csv>

### 1.2.1 Selection Criteria:

1.2.1.1 The data (see above) does meet the criteria of the assignment. In that it is relevant to health, publically accessible, sizable (61 columns and 530,292 rows), includes both categorical (e.g. country) and continuous variables (e.g. new\_cases\_per\_million, total\_deaths\_per\_million) and finally has been ethically sourced and de-identified.

```
## [1] "https://docs.owid.io/projects/covid/en/latest/dataset.html"
```

```
## n_rows n_cols
## 530292      61
```

```
## [1] "country"           "date"
## [3] "total_cases"       "new_cases"
## [5] "new_cases_smoothed" "total_cases_per_million"
## [7] "new_cases_per_million" "new_cases_smoothed_per_million"
## [9] "total_deaths"      "new_deaths"
## [11] "new_deaths_smoothed" "total_deaths_per_million"
```

```
## [1] "Ukraine"           "United Arab Emirates"
## [3] "United Kingdom"    "United States"
## [5] "United States Virgin Islands"
```

### 1.3 2) Introduction (4-6 Sentences)

This project uses Covid 19 data from ‘Our world in data’. We use this to primarily compare how daily new cases per million varied across four countries in 2021. We focus on 2021 to keep our comparisons on a common phase of the pandemic. The dataset itself does cover many more countries and years and also includes data on total cases and total deaths. We used the fields that have the suffix ‘per\_million’ as any comparisons scale by population size.

### 1.4 3) Why this dataset? (1 paragraph)

This dataset is highly relevant to health outcomes. The dataset is also very well documented, large (61 columns and 530k rows). For analysis potential, this dataset has both continuous fields (total\_deaths\_per\_million, new\_cases\_per\_million) and categorical (e.g. Country), feasible for tables, histograms, boxplots, time trends. Using the fields with the suffix “per\_million” allows better scaling for cross country comparisons and summaries.

### 1.5 4) Variables and structure

### 1.6 5) Research questions

- 1. What share of days exceed a threshold (to simulated a government policy threshold to “flatten the curve”) e.g 50 cases per million in each country
- 2. Which of the selected countries had the highest typical daily new cases per million in 2021
- 3. How did the monthly mean of new cases per million over 2021 for each country

### 1.7 6) Data clean up & Processing plan

We parsed the date field and derived a ‘year’ variable, then restricted the dataset to 2021 to keep figures more legible and comparable. We fixed our analysis to a small set of countries (United States, United Kingdom, China, Belgium) and then verified each has sufficient non missing values for ‘new\_cases\_per\_million’ in 2021. this processing prepares the data for descriptive statistics and many visualisations.

```
## United States United Kingdom      China      Belgium
##           365           365           365           365
```

### 1.8 7) Descriptive statistics & visualisations

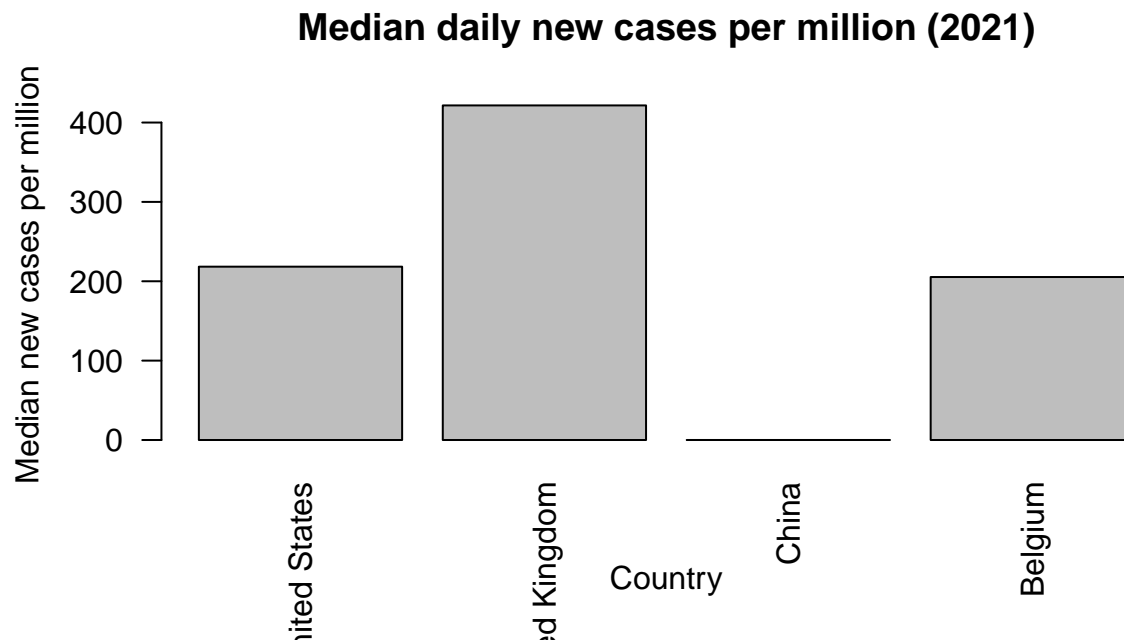
We summarise categories (counts/proportions), report center & spread for one numeric variable and add simple plots to visualise patterns ### 7.1) One-Way frequency table (categorical) Counts and proportions for a categorical variable

```
##
## United States United Kingdom      China      Belgium
##           365           365           365           365
```

```
##
## United States United Kingdom China Belgium
## 0.25 0.25 0.25 0.25
```

### 1.8.1 7.2) Bar Chart of Disease.Category (counts)

Bar chart / Bar plot of disease category by count



### 1.8.2 7.3) Two way table (category by category)

```
##
## FALSE TRUE
## United States 37 328
## United Kingdom 55 310
## China 365 0
## Belgium 20 345
```

```
##
## FALSE TRUE
## United States 0.101 0.899
## United Kingdom 0.151 0.849
## China 1.000 0.000
## Belgium 0.055 0.945
```

```
##
## FALSE TRUE
```

```
## United States 0.078 0.334
## United Kingdom 0.115 0.315
## China 0.765 0.000
## Belgium 0.042 0.351
```

```
##
## FALSE TRUE Sum
## United States 37 328 365
## United Kingdom 55 310 365
## China 365 0 365
## Belgium 20 345 365
## Sum 477 983 1460
```

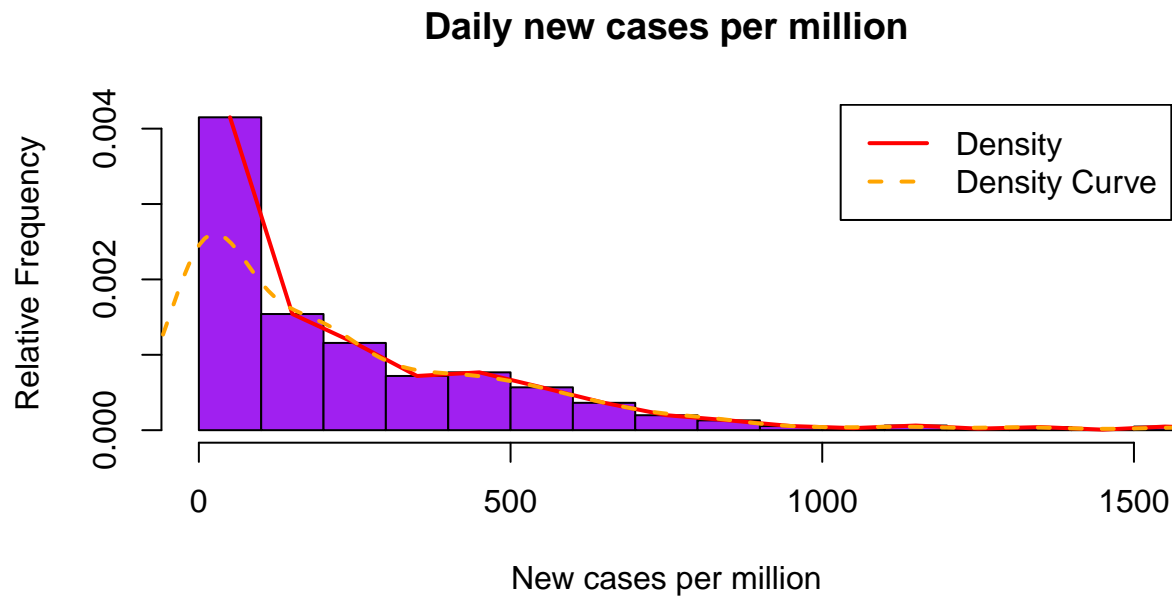
```
## United States United Kingdom China Belgium
## 0.899 0.849 0.000 0.945
```

### 1.8.3 7.4) Center & Spread (overall, selected countries, 2021)

```
## median IQR sd
## 162.9 379.7 357.4
```

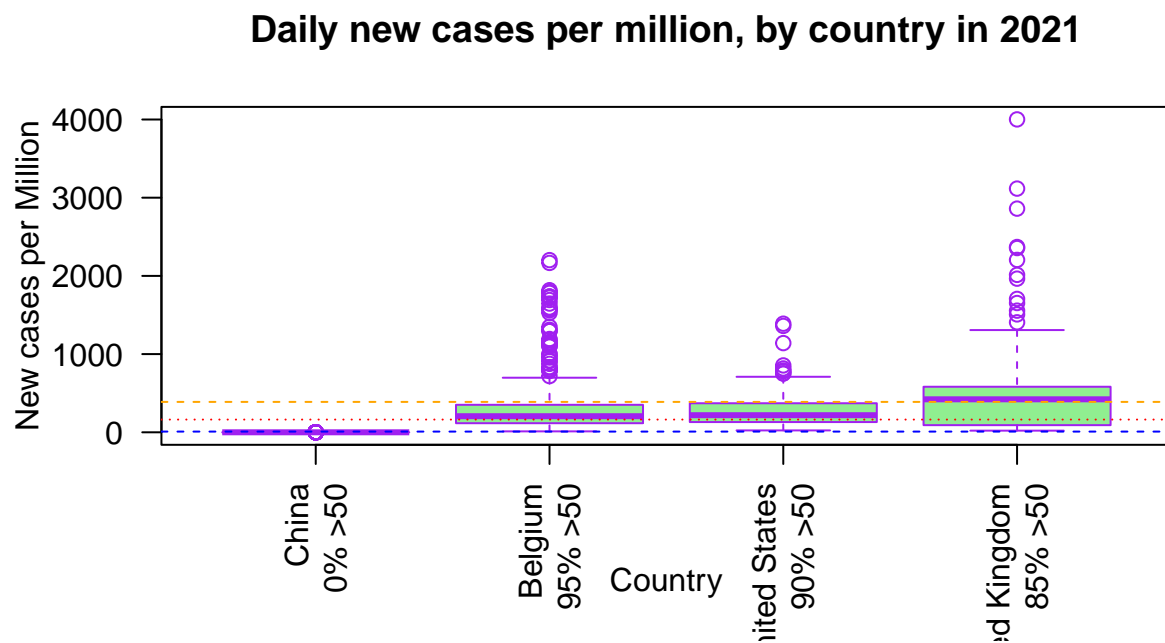
```
## country median IQR sd
## 1 United States 218.4 240.7 201.6
## 2 United Kingdom 421.6 491.1 456.4
## 3 China 0.0 0.1 0.1
## 4 Belgium 205.4 235.4 396.2
```

#### 1.8.4 7.5) Histogram (shape of the distribution)

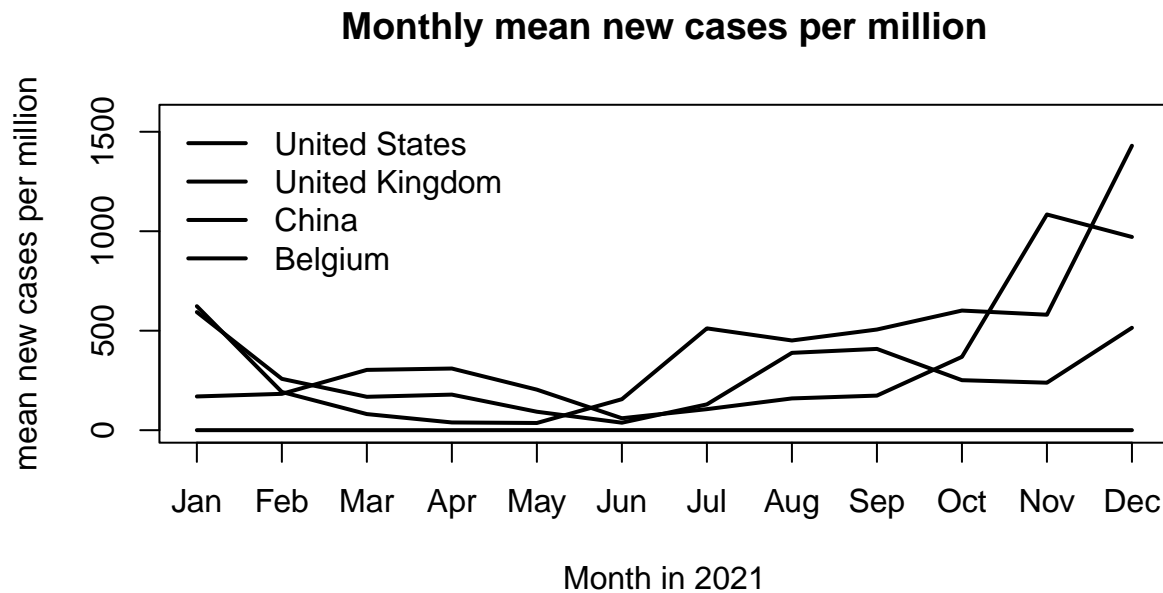


#### 1.8.5 7.6) Boxplot (numeric by category)

Boxplot (Mortality rate by category)



### 1.8.6 7.7) Simple time trend (average by year)



### 1.9 8) Planned statistical methods

We will compare 2021 distributions of 'new\_cases\_per\_million' across the four countries using....

### 1.10 9) Limitations

- Measurement differences - countries have different reporting rules, testing cadence & breadth.
- Scope - Only 2021 was analysed. Other years or waves of the disease may show other patterns.
- per million rates do not adjust for demographics of each country, which may show other patterns.
- China has several near zero analysis - This may reflect reporting practices of this specific country

## 2 Project Idea Two - Covid 19 Hospitalizations in France

### 3 Link to the dataset

Kaggle - Coronavirusdataset France (file: `chiffres-cles.csv`)

Actual URL: <https://www.kaggle.com/datasets/mclikmb4/coronavirusdataset-france?select=chiffres-cles.csv> Google drive URL: [https://drive.google.com/file/d/1rXHdGEDWFAMaitmkNSgehAt\\_e2FaC\\_PZ/view?usp=sharing](https://drive.google.com/file/d/1rXHdGEDWFAMaitmkNSgehAt_e2FaC_PZ/view?usp=sharing)

## 4 Introduction to the dataset

This dataset provides daily COVID-19 surveillance indicators for France at multiple geographic granularities (country, region, department, overseas collectivities). Each record includes a calendar date, a location code, and a location name, enabling comparisons across space and time. Indicators cover hospitalized patients, ICU occupancy, cumulative deaths, cumulative recoveries, and daily flows of new admissions (hospital and ICU). Source/provenance fields support auditability. The structure suits descriptive analyses and visualizations, with optional regional comparisons to highlight spatial heterogeneity. These indicators and their definitions are documented on the Kaggle dataset page (mclikmb4, 2020-2021).

## 5 Dataset justification

**Relevance:** Directly biomedical/public-health, reflecting real-world hospital and ICU loads during COVID-19.

**Size/structure:** The file far exceeds the minimum requirements (well over 100 rows and more than 20 columns) and includes both categorical (granularity, location IDs, sources) and continuous (counts) variables.

**Accessibility/ethics:** Publicly accessible aggregated, de-identified counts suitable for academic use.

**Analytical potential:** Enables trend estimation, wave identification, geographic comparison, and lead-lag analysis between admissions (“flow”) and occupancy (“stock”).

**Ethical use.** The dataset consists of aggregated, de-identified counts without PII; no patient-level identifiers are present, aligning with course requirements for ethical, public data.

## 6 Variables description

### Key columns:

`date` (daily), `granularity` (country, region, department), `location_code` (location code), `location_name` (location name).

### Indicators:

- `hospitalized` - current hospitalized patients
- `icu_patients` - current ICU patients
- `deaths` - cumulative deaths
- `recovered` - cumulative recoveries
- `new_hospitalizations` - new daily hospital admissions
- `new_icu_admissions` - new daily ICU admissions

### Additional fields:

`confirmed_cases` and `tested` may be present with different levels of completeness.

**Note:** Due to several missing/invalid values (NaN/Inf), the `tested` column is largely unusable for analysis and is excluded from primary summaries and plots.

### Source metadata:

source\_name, source\_url, source\_archive, source\_type.

Table 1: Row counts by geographic granularity

granularity	n
department	40715
region	7708
country	817
overseas_collectivity	131
world	83

Table 2: Summary statistics for key numeric indicators

variable	n	mean	sd	median	min	max
confirmed_cases	3081	121010.685	508142.429	27.0	0	3560764
deaths	47928	920.086	4150.452	135.0	0	70574
hospitalized	46826	578.225	2597.057	91.0	0	33497
icu_patients	46743	80.489	387.667	10.0	0	7148
new_hospitalizations	46095	32.664	166.648	4.0	0	4281
new_icu_admissions	46095	5.421	28.033	0.0	0	771
recovered	46712	3949.800	17835.138	645.5	0	299624
tested	0	NaN	NA	NA	Inf	-Inf

## 7 Research question(s)

1. **National waves:** How did France’s national hospitalization and ICU occupancy evolve across early pandemic waves (2020-2021)?
2. **Flow-stock timing:** Do peaks in new hospital admissions precede peaks in current hospitalizations, and by roughly how many days?

## 8 Data cleanup and processing plan

- **Parsing and types:** Ensure the `date` field is properly parsed as a date variable and convert indicator fields into numeric types for consistency.
- **Subsetting:** For national trends, include only rows classified as country with `location_code` = “FRA”. For geographic comparisons, restrict the dataset to rows where `granularity` is region.
- **Missingness:** Quantify missing values for each column and handle them transparently by applying listwise deletion for plotted series (no imputation).



- **Duplicates:** Identify and remove duplicate entries defined by the combination of `date` and `location_code`.
- **Provenance:** Retain all source metadata fields, and include them in the appendix when relevant for transparency.

## 9 Descriptive statistics (figures in Appendix)

France’s national indicators exhibit multi-wave patterns during 2020-2021. Hospital occupancy and ICU burden rise and fall in tandem with case surges, while cumulative deaths increase monotonically. The timing relationship between new admissions (flow) and current occupancy (stock) suggests admissions lead occupancy by several days. For visuals supporting these statements, see Appendix Figures A1-A3. Tables above summarize structure and central tendencies.

Across all rows, the median current hospitalizations was 91, with an IQR of 25-285; ICU occupancy had a much lower median, which is expected since ICU is a subset of the total hospital (median 10), consistent with ICU being a subset of total hospital burden.

## 10 Planned statistical methods

- **Lagged cross-correlation** between `new_hospitalizations` (flow) and `hospitalized` (stock) to estimate lead time from admissions to occupancy.
- **Regional comparison** of ICU vs hospital burden by wave period (medians, IQRs).
- **Simple time-series decomposition** on national hospitalizations to separate trend/seasonal/residual components (if applicable).

## 11 Limitations

Several fields like `tested` and early `confirmed_cases` have bad coverage over time, and indicators are hospital-centric rather than community-representative. Counts are aggregated and de-identified, so patient-level cannot be controlled. Because the dataset mixes granularities (national, regional, departmental), comparing across levels requires careful subsetting (`granularity == "country"` for national trends). These constraints limit causal interpretation, so we have to focus more on descriptive trends and clearly labeled comparisons.

## 12 Appendix

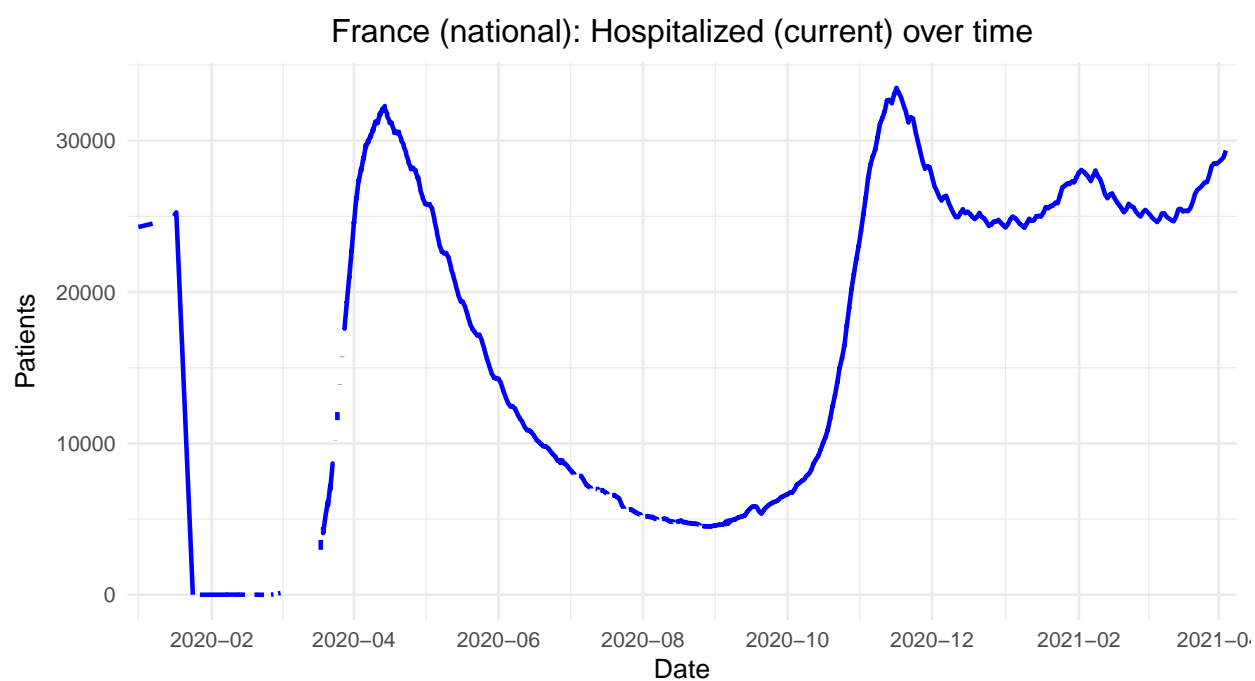


Figure 1: France (national): Hospitalized (current) over time.

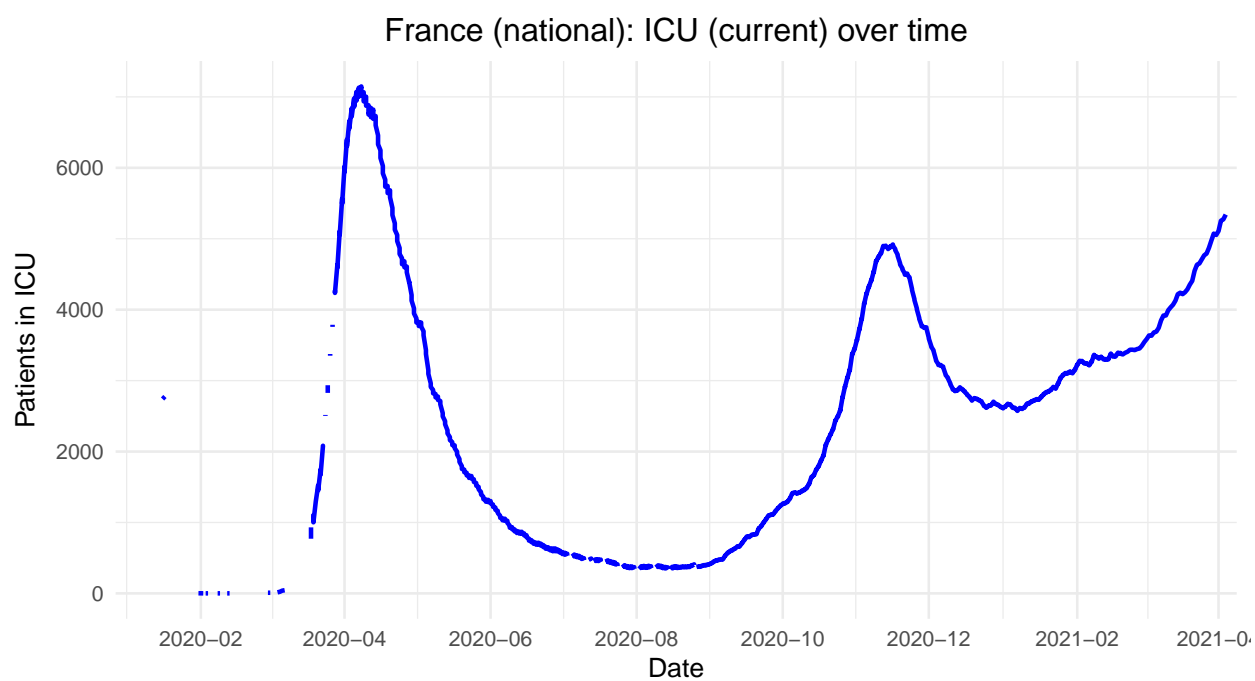


Figure 2: France (national): ICU (current) over time.

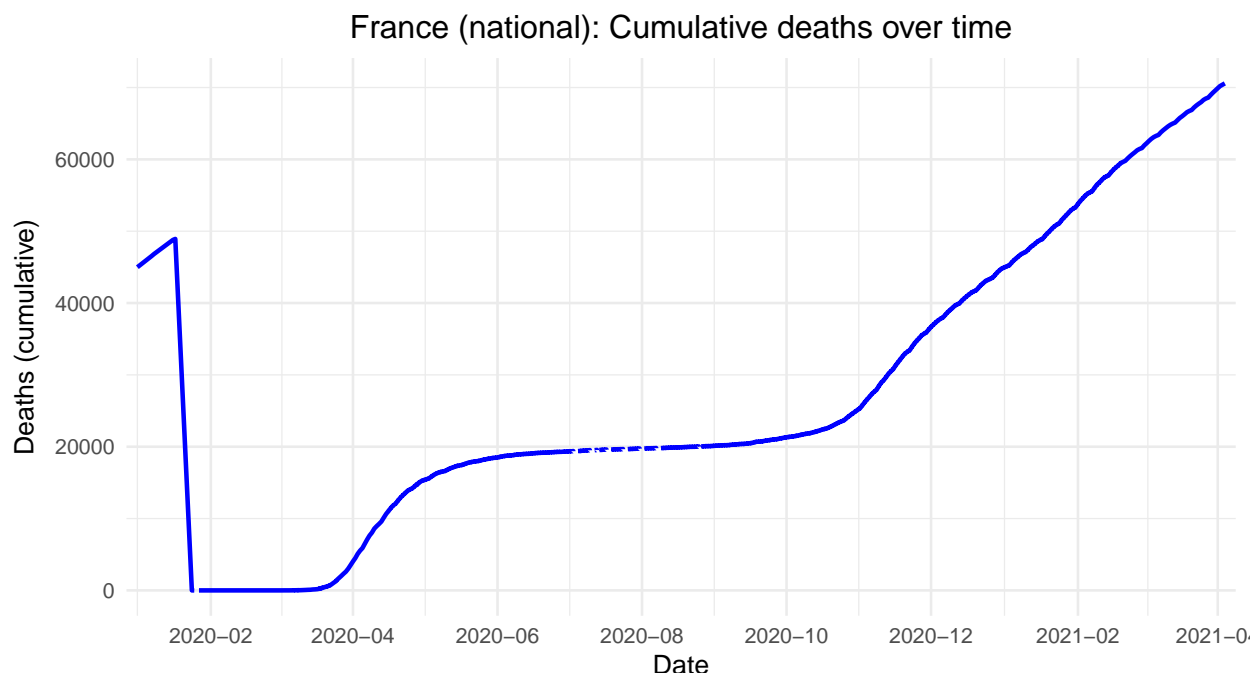


Figure 3: France (national): Cumulative deaths over time.

## 13 Project Idea Three - Heart attack

### 14 1.Link to the dataset

<https://www.kaggle.com/datasets/iamsouravbanerjee/heart-attack-prediction-dataset>

### 15 2. Introduction to dataset

The Heart Attack Prediction Dataset, available on Kaggle, is a comprehensive resource for studying the clinical, lifestyle, and demographic factors associated with cardiovascular risk. It consists of 8,763 de-identified patient records, including continuous variables such as age, cholesterol, blood pressure, and heart rate, as well as categorical features like sex, chest pain type, smoking habits, diabetes status, and dietary patterns. Socioeconomic and geographic attributes, including income and region, further enrich the dataset by adding broader context to heart health predictors. The primary outcome variable indicates whether a patient is at risk of a heart attack, making the dataset well-suited for statistical analysis, visualization, and classification tasks. Its diverse mix of variables supports exploration of correlations, risk factors, and group comparisons, while also providing an ethical and accessible foundation for predictive modeling in cardiovascular health research.

## 16 3. Dataset justification

I chose the Heart Attack Prediction Dataset because it directly addresses a critical biomedical challenge cardiovascular disease which remains one of the leading causes of mortality worldwide. The dataset integrates clinical, lifestyle, and demographic variables, making it highly relevant for exploring the multifactorial nature of heart health. With its balanced mix of categorical and continuous features, it offers strong potential for applying a variety of statistical methods, visualizations, and predictive modeling techniques. Its size and diversity of attributes make it complex enough to yield meaningful insights, yet still manageable for academic analysis. Overall, this dataset provides both real-world relevance and analytical richness, making it an excellent candidate for this project.

## 17 4. Variables description

Key columns include Patient ID (unique identifier for each record), Age (in years), Sex (male or female), Cholesterol (cholesterol levels in mg/dL), Blood Pressure (systolic/diastolic in mmHg), Heart Rate (beats per minute), and BMI (body mass index, kg/m<sup>2</sup>). Clinical indicators capture Diabetes status (Yes/No), Family History of heart problems (1 = Yes, 0 = No), Previous Heart Problems (1 = Yes, 0 = No), Medication Use (1 = Yes, 0 = No), and Triglyceride levels (mg/dL). Lifestyle-related attributes include Smoking (1 = Smoker, 0 = Non-smoker), Obesity (1 = Obese, 0 = Not obese), Alcohol Consumption (None, Light, Moderate, Heavy), Diet (Healthy, Average, Unhealthy), Exercise Hours Per Week, Physical Activity Days Per Week, Stress Level (1–10 scale), Sedentary Hours Per Day, and Sleep Hours Per Day. Socioeconomic and demographic fields consist of Income, Country, Continent, and Hemisphere. The target variable, Heart Attack Risk, is a binary indicator (1 = Yes, 0 = No) denoting whether the patient is at risk of a heart attack.

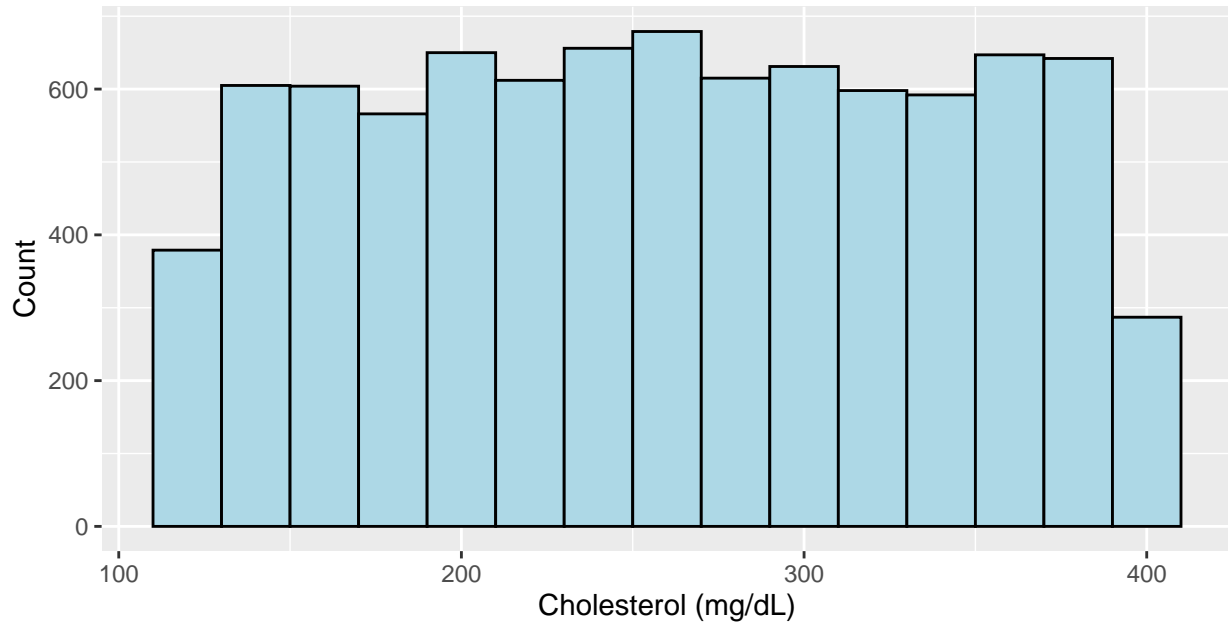
```
## 'data.frame':    8763 obs. of  26 variables:
##  $ Patient.ID      : chr  "BMW7812" "CZE1114" "BNI9906" "JLN3497" ...
##  $ Age              : int   67 21 21 84 66 54 90 84 20 43 ...
##  $ Sex              : chr   "Male" "Male" "Female" "Male" ...
##  $ Cholesterol       : int  208 389 324 383 318 297 358 220 145 248 ...
##  $ Blood.Pressure    : chr   "158/88" "165/93" "174/99" "163/100" ...
##  $ Heart.Rate        : int   72 98 72 73 93 48 84 107 68 55 ...
##  $ Diabetes          : int    0 1 1 1 1 1 0 0 1 0 ...
##  $ Family.History    : int    0 1 0 1 1 1 0 0 0 1 ...
##  $ Smoking           : int    1 1 0 1 1 1 1 1 1 1 ...
##  $ Obesity           : int    0 1 0 0 1 0 0 1 1 1 ...
##  $ Alcohol.Consumption : int    0 1 0 1 0 1 1 1 0 1 ...
##  $ Exercise.Hours.Per.Week : num  4.17 1.81 2.08 9.83 5.8 ...
##  $ Diet              : chr   "Average" "Unhealthy" "Healthy" "Average" ...
##  $ Previous.Heart.Problems : int    0 1 1 1 1 1 0 0 0 0 ...
##  $ Medication.Use     : int    0 0 1 0 0 1 0 1 0 0 ...
##  $ Stress.Level       : int    9 1 9 9 6 2 7 4 5 4 ...
##  $ Sedentary.Hours.Per.Day : num  6.62 4.96 9.46 7.65 1.51 ...
##  $ Income            : int  261404 285768 235282 125640 160555 241339 190450 1...
##  $ BMI               : num   31.3 27.2 28.2 36.5 21.8 ...
##  $ Triglycerides      : int   286 235 587 378 231 795 284 370 790 232 ...
```

```
## $ Physical.Activity.Days.Per.Week: int 0 1 4 3 1 5 4 6 7 7 ...
## $ Sleep.Hours.Per.Day              : int 6 7 4 4 5 10 10 7 4 7 ...
## $ Country                          : chr "Argentina" "Canada" "France" "Canada" ...
## $ Continent                        : chr "South America" "North America" "Europe" "North Am
## $ Hemisphere                       : chr "Southern Hemisphere" "Northern Hemisphere" "North
## $ Heart.Attack.Risk                 : int 0 0 0 0 0 1 1 1 0 0 ...
```

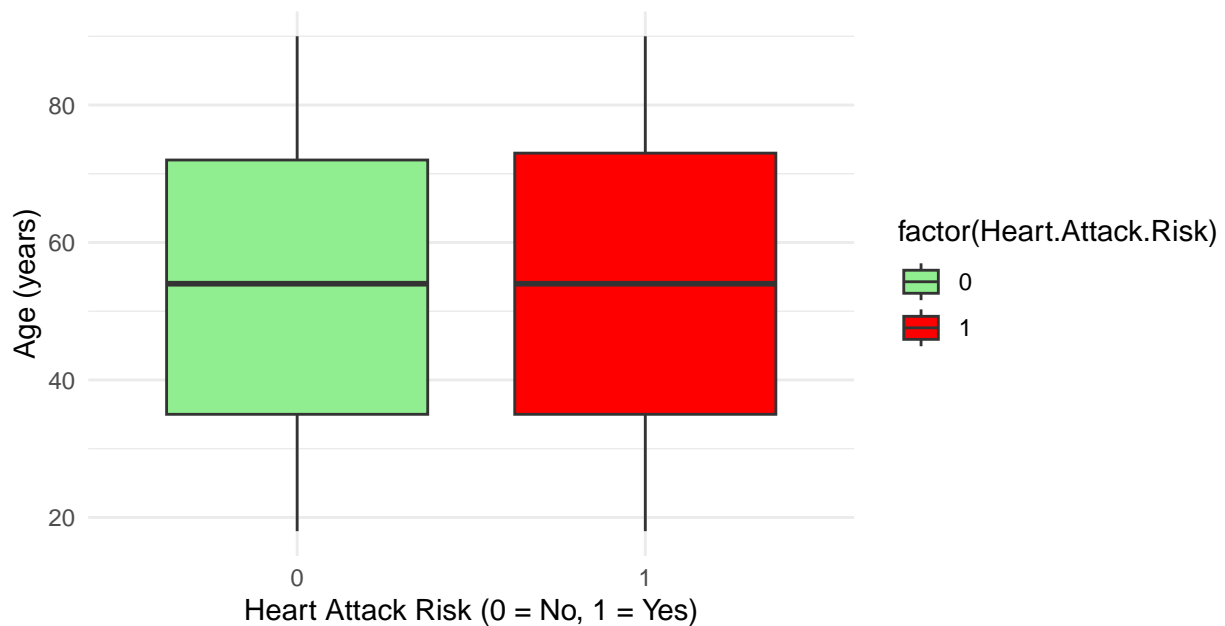
```
## Patient.ID           Age           Sex           Cholesterol
## Length:8763          Min.   :18.00   Length:8763          Min.   :120.0
## Class :character     1st Qu.:35.00   Class :character     1st Qu.:192.0
## Mode  :character     Median :54.00   Mode  :character     Median :259.0
##                               Mean   :53.71               Mean   :259.9
##                               3rd Qu.:72.00               3rd Qu.:330.0
##                               Max.   :90.00               Max.   :400.0
## Blood.Pressure       Heart.Rate       Diabetes       Family.History
## Length:8763          Min.   : 40.00   Min.   :0.0000   Min.   :0.000
## Class :character     1st Qu.: 57.00   1st Qu.:0.0000   1st Qu.:0.000
## Mode  :character     Median : 75.00   Median :1.0000   Median :0.000
##                               Mean   : 75.02   Mean   :0.6523   Mean   :0.493
##                               3rd Qu.: 93.00   3rd Qu.:1.0000   3rd Qu.:1.000
##                               Max.   :110.00   Max.   :1.0000   Max.   :1.000
## Smoking              Obesity          Alcohol.Consumption Exercise.Hours.Per.Week
## Min.   :0.0000   Min.   :0.0000   Min.   :0.0000   Min.   : 0.002442
## 1st Qu.:1.0000   1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.: 4.981579
## Median :1.0000   Median :1.0000   Median :1.0000   Median :10.069559
## Mean   :0.8968   Mean   :0.5014   Mean   :0.5981   Mean   :10.014284
## 3rd Qu.:1.0000   3rd Qu.:1.0000   3rd Qu.:1.0000   3rd Qu.:15.050018
## Max.   :1.0000   Max.   :1.0000   Max.   :1.0000   Max.   :19.998709
## Diet                Previous.Heart.Problems Medication.Use       Stress.Level
## Length:8763          Min.   :0.0000           Min.   :0.0000   Min.   : 1.00
## Class :character     1st Qu.:0.0000           1st Qu.:0.0000   1st Qu.: 3.00
## Mode  :character     Median :0.0000           Median :0.0000   Median : 5.00
##                               Mean   :0.4958           Mean   :0.4983   Mean   : 5.47
##                               3rd Qu.:1.0000           3rd Qu.:1.0000   3rd Qu.: 8.00
##                               Max.   :1.0000           Max.   :1.0000   Max.   :10.00
## Sedentary.Hours.Per.Day Income          BMI           Triglycerides
## Min.   : 0.001263   Min.   : 20062   Min.   :18.00   Min.   : 30.0
## 1st Qu.: 2.998794   1st Qu.: 88310   1st Qu.:23.42   1st Qu.:225.5
## Median : 5.933622   Median :157866   Median :28.77   Median :417.0
## Mean   : 5.993690   Mean   :158263   Mean   :28.89   Mean   :417.7
## 3rd Qu.: 9.019125   3rd Qu.:227749   3rd Qu.:34.32   3rd Qu.:612.0
## Max.   :11.999313   Max.   :299954   Max.   :40.00   Max.   :800.0
## Physical.Activity.Days.Per.Week Sleep.Hours.Per.Day Country
## Min.   :0.00           Min.   : 4.000   Length:8763
## 1st Qu.:2.00           1st Qu.: 5.000   Class :character
## Median :3.00           Median : 7.000   Mode  :character
## Mean   :3.49           Mean   : 7.024
## 3rd Qu.:5.00           3rd Qu.: 9.000
```

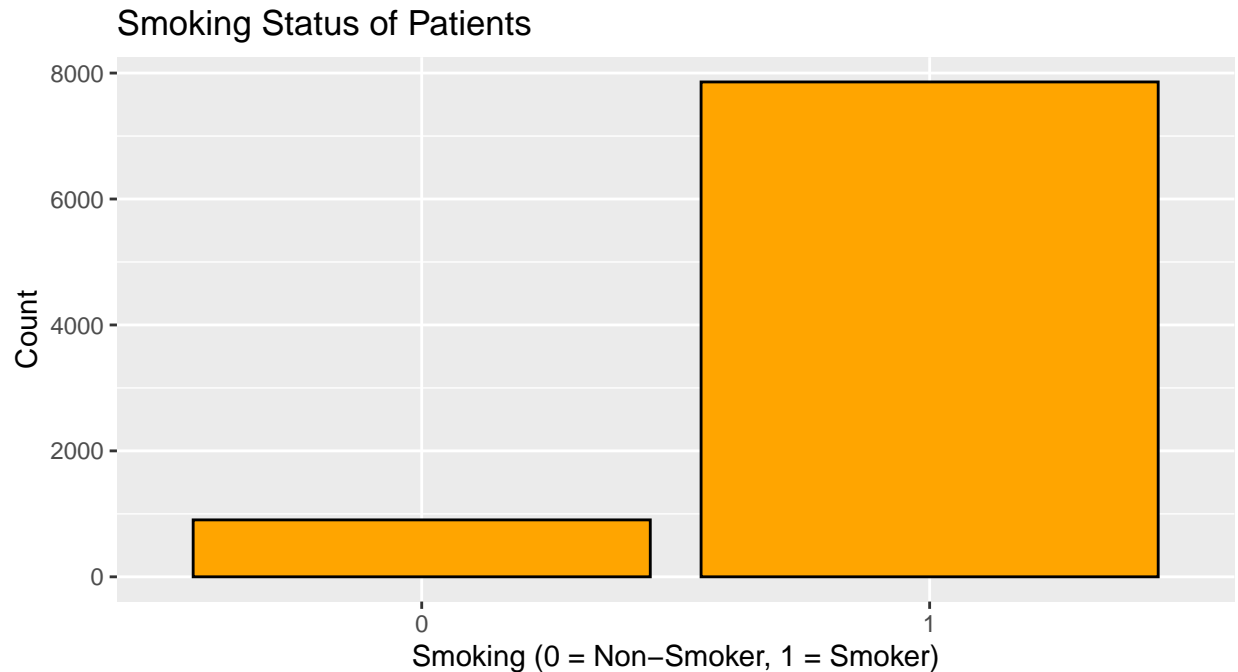
```
## Max.      :7.00                      Max.      :10.000
## Continent      Hemisphere      Heart.Attack.Risk
## Length:8763    Length:8763    Min.       :0.0000
## Class :character Class :character 1st Qu.:0.0000
## Mode  :character Mode  :character Median  :0.0000
##                                     Mean   :0.3582
##                                     3rd Qu.:1.0000
##                                     Max.   :1.0000
```

Distribution of Cholesterol Levels



Age Distribution by Heart Attack Risk





## 18 5. Research questions

1. Which clinical, lifestyle, and demographic factors are most strongly associated with the risk of heart attack in patients?
2. Which features contribute most to a machine learning model's decision boundary for predicting heart attack risk?

## 19 6. Data cleanup and processing plan

- Check for missing values: Identify NAs using `colSums(is.na(hd))`; if very few, remove those rows; if moderate, impute using mean/median for continuous variables (e.g., Cholesterol, BMI) and mode for categorical variables (e.g., Diet, Alcohol Consumption).
- Remove duplicate entries: Drop exact duplicates or repeated Patient IDs to avoid over-representation using `hd <- hd[!duplicated(hd), ]`.
- Fix inconsistent formats: Split Blood Pressure into two numeric columns (Systolic and Diastolic) and convert binary indicators (0/1) like Diabetes, Smoking, and Heart Attack Risk into categorical factors.
- Validate ranges & handle outliers: Review continuous variables (e.g., Cholesterol, Triglycerides, BMI, Sleep Hours) for biologically implausible values; correct, cap, or remove extreme outliers as appropriate.
- Standardize categorical variables: Ensure consistent levels for Sex (Male/Female), Diet (Healthy/Average/Unhealthy), and Alcohol Consumption (None/Light/Moderate/Heavy).

- Create derived variables: Add new groupings such as Age Groups (e.g., 18–30, 31–50, 51–70, 71–90) and BMI Categories (Underweight, Normal, Overweight, Obese) to facilitate group comparisons in descriptive statistics and visualization.

## 20 7. Descriptive statistics and data visualizations

```
##
## Variable: Age
## Mean: 53.70798
## Median: 54
## Range: 72
## Standard Deviation: 21.24951
##
## Variable: Cholesterol
## Mean: 259.8772
## Median: 259
## Range: 280
## Standard Deviation: 80.86328
##
## Variable: Heart.Rate
## Mean: 75.02168
## Median: 75
## Range: 70
## Standard Deviation: 20.55095
##
## Variable: BMI
## Mean: 28.89145
## Median: 28.769
## Range: 21.99487
## Standard Deviation: 6.319181
##
## Variable: Triglycerides
## Mean: 417.6771
## Median: 417
## Range: 770
## Standard Deviation: 223.7481
##
## Variable: Exercise.Hours.Per.Week
## Mean: 10.01428
## Median: 10.06956
## Range: 19.99627
## Standard Deviation: 5.783745
##
## Variable: Stress.Level
## Mean: 5.469702
## Median: 5
## Range: 9
```



```

## Standard Deviation: 2.859622
##
## Variable: Sedentary.Hours.Per.Day
## Mean: 5.99369
## Median: 5.933622
## Range: 11.99805
## Standard Deviation: 3.466359
##
## Variable: Income
## Mean: 158263.2
## Median: 157866
## Range: 279892
## Standard Deviation: 80575.19
##
## Variable: Physical.Activity.Days.Per.Week
## Mean: 3.489672
## Median: 3
## Range: 7
## Standard Deviation: 2.282687
##
## Variable: Sleep.Hours.Per.Day
## Mean: 7.023508
## Median: 7
## Range: 6
## Standard Deviation: 1.988473

##
## Variable: Sex
##
## Female    Male
##    2652    6111
##
## Variable: Diabetes
##
##      0      1
## 3047 5716
##
## Variable: Family.History
##
##      0      1
## 4443 4320
##
## Variable: Smoking
##
##      0      1
##  904 7859
##
## Variable: Obesity

```

```

##
##      0      1
## 4369 4394
##
## Variable: Alcohol.Consumption
##
##      0      1
## 3522 5241
##
## Variable: Diet
##
##      Average      Healthy Unhealthy
##      2912      2960      2891
##
## Variable: Previous.Heart.Problems
##
##      0      1
## 4418 4345
##
## Variable: Medication.Use
##
##      0      1
## 4396 4367
##
## Variable: Country
##
##      Argentina      Australia      Brazil      Canada      China
##      471      449      462      440      436
##      Colombia      France      Germany      India      Italy
##      429      446      477      412      431
##      Japan      New Zealand      Nigeria      South Africa      South Korea
##      433      435      448      425      409
##      Spain      Thailand United Kingdom United States      Vietnam
##      430      428      457      420      425
##
## Variable: Continent
##
##      Africa      Asia      Australia      Europe North America
##      873      2543      884      2241      860
## South America
##      1362
##
## Variable: Hemisphere
##
## Northern Hemisphere Southern Hemisphere
##      5660      3103
##
## Variable: Heart.Attack.Risk

```

```
##
##      0      1
## 5624 3139
```

```
## Categorical Columns:
```

```
## [1] "Patient.ID"      "Sex"              "Blood.Pressure"  "Diet"
## [5] "Country"          "Continent"        "Hemisphere"
```

```
##
```

```
## Numerical Columns:
```

```
## [1] "Age"                "Cholesterol"
## [3] "Heart.Rate"         "Diabetes"
## [5] "Family.History"     "Smoking"
## [7] "Obesity"            "Alcohol.Consumption"
## [9] "Exercise.Hours.Per.Week" "Previous.Heart.Problems"
## [11] "Medication.Use"     "Stress.Level"
## [13] "Sedentary.Hours.Per.Day" "Income"
## [15] "BMI"                "Triglycerides"
## [17] "Physical.Activity.Days.Per.Week" "Sleep.Hours.Per.Day"
## [19] "Heart.Attack.Risk"
```

```
##                                variable    n    mean
## Age                           Age 8763    53.71
## Cholesterol                   Cholesterol 8763    259.88
## Heart.Rate                    Heart.Rate 8763    75.02
## Diabetes                      Diabetes 8763     0.65
## Family.History                Family.History 8763     0.49
## Smoking                       Smoking 8763     0.90
## Obesity                       Obesity 8763     0.50
## Alcohol.Consumption           Alcohol.Consumption 8763     0.60
## Exercise.Hours.Per.Week       Exercise.Hours.Per.Week 8763    10.01
## Previous.Heart.Problems       Previous.Heart.Problems 8763     0.50
## Medication.Use                Medication.Use 8763     0.50
## Stress.Level                  Stress.Level 8763     5.47
## Sedentary.Hours.Per.Day       Sedentary.Hours.Per.Day 8763     5.99
## Income                        Income 8763 158263.18
## BMI                           BMI 8763     28.89
## Triglycerides                 Triglycerides 8763    417.68
## Physical.Activity.Days.Per.Week Physical.Activity.Days.Per.Week 8763     3.49
## Sleep.Hours.Per.Day           Sleep.Hours.Per.Day 8763     7.02
## Heart.Attack.Risk             Heart.Attack.Risk 8763     0.36
##                                median    range
## Age                           54.00    72.00
## Cholesterol                   259.00    280.00
## Heart.Rate                    75.00    70.00
```

## Diabetes	1.00	1.00
## Family.History	0.00	1.00
## Smoking	1.00	1.00
## Obesity	1.00	1.00
## Alcohol.Consumption	1.00	1.00
## Exercise.Hours.Per.Week	10.07	20.00
## Previous.Heart.Problems	0.00	1.00
## Medication.Use	0.00	1.00
## Stress.Level	5.00	9.00
## Sedentary.Hours.Per.Day	5.93	12.00
## Income	157866.00	279892.00
## BMI	28.77	21.99
## Triglycerides	417.00	770.00
## Physical.Activity.Days.Per.Week	3.00	7.00
## Sleep.Hours.Per.Day	7.00	6.00
## Heart.Attack.Risk	0.00	1.00

## 21 8. Planned statistical methods

As the project progresses, I plan to use chi-square tests to assess associations between categorical factors (e.g., smoking, diabetes) and heart attack risk, and t-tests/ANOVA to compare continuous measures (e.g., cholesterol, BMI) across groups. To build predictive insight, I will apply logistic regression and may explore machine learning models such as decision trees or random forests. These methods will help identify key risk factors and evaluate their predictive power.

## 22 ) JOINT PROJECTS - References

Project 1 - Our World in Data. “Coronavirus Pandemic (COVID-19) dataset.” source location - Source: URL: <https://docs.owid.io/projects/covid/en/latest/dataset.html> Project 2 - mclikmb4, (2021, April 4), Coronavirus-dataset France, Kaggle, <https://www.kaggle.com/datasets/mclikmb4/coronavirusdataset-france?select=chiffres-cles.csv>

Project 3 - Banerjee, S. (2021). Heart Attack Prediction Dataset. Kaggle. <https://www.kaggle.com/datasets/iamsouravbanerjee/heart-attack-prediction-dataset>

## 23 ) PROJECTS - Appendix

### 23.1 Project 1

### 23.2 Project 2

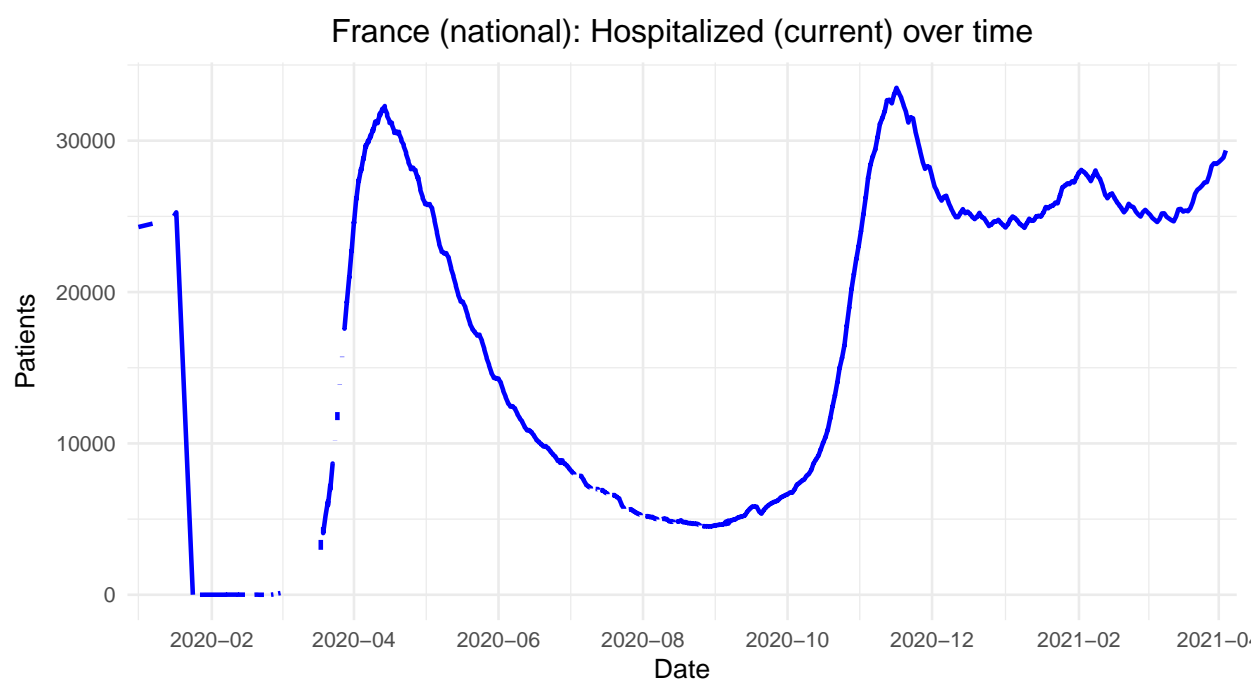


Figure 4: France (national): Hospitalized (current) over time.

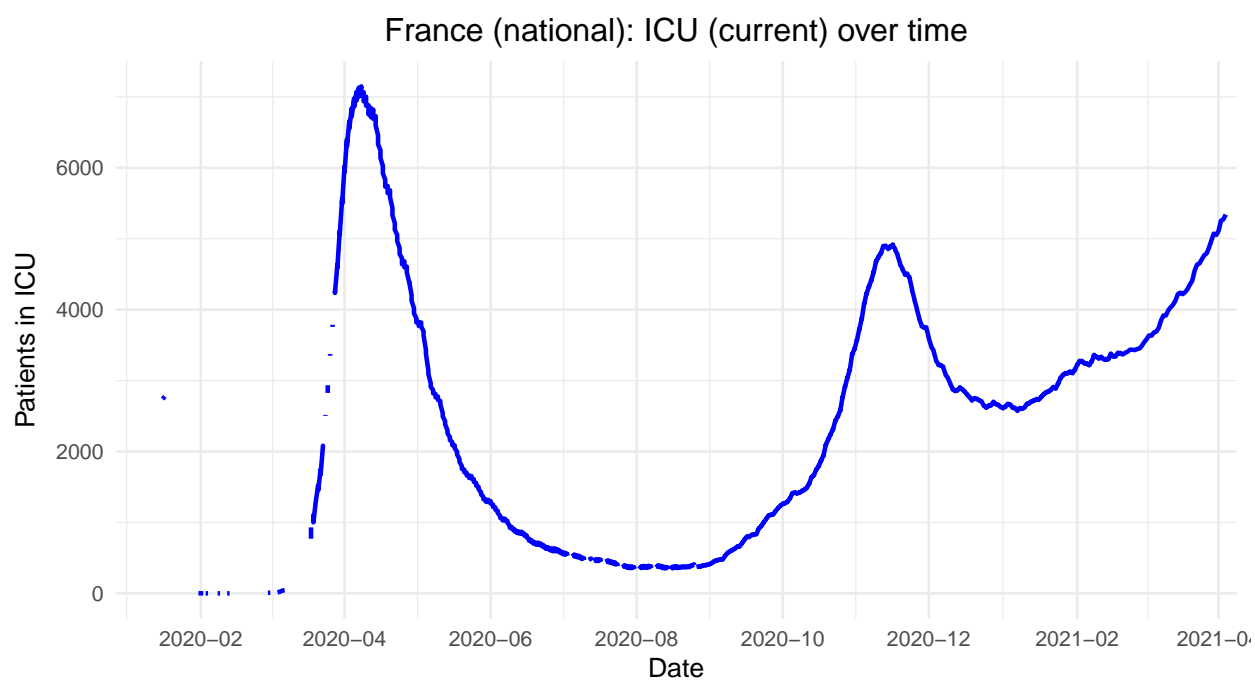


Figure 5: France (national): ICU (current) over time.

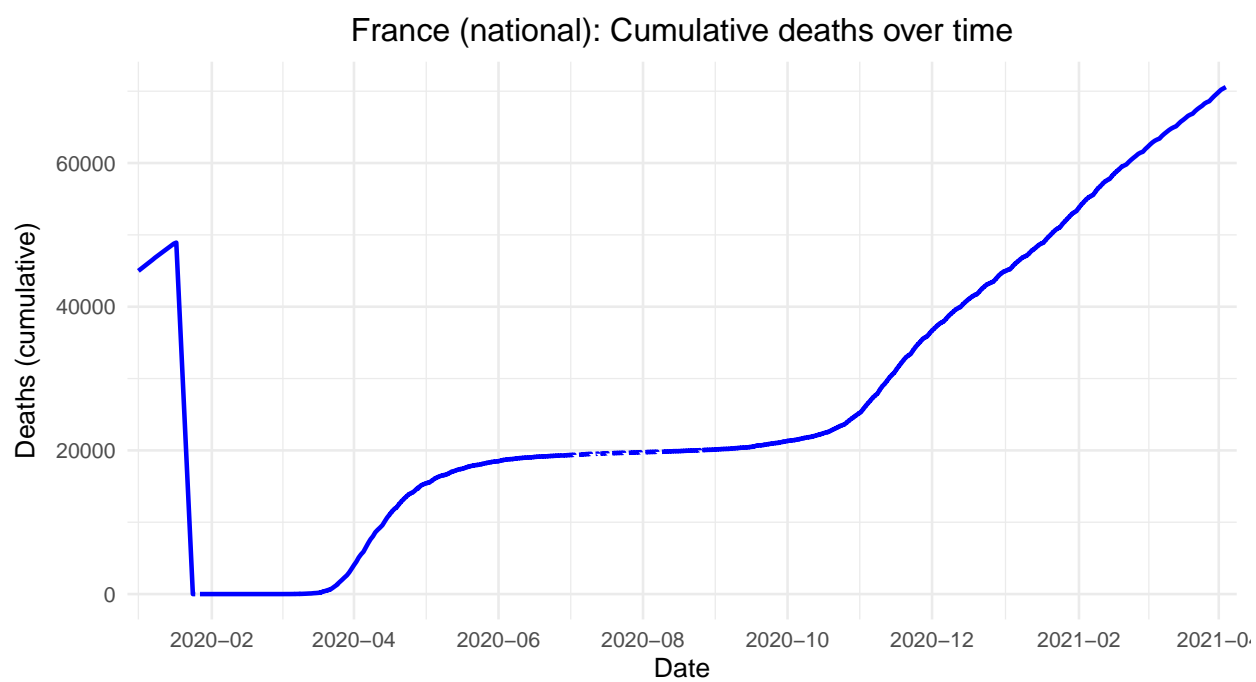


Figure 6: France (national): Cumulative deaths over time.