# **Final Project**

#### **DATASET REFERENCE:**

"https://data.europa.eu/data/datasets/e50ca3e5-787e-4911-bc9e-d8012fa5af7a?locale=en"

```
# Load necessary library
library(dplyr)
```

```
Attaching package: 'dplyr'

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

filter, lag

The following objects are masked from 'package:base':
```

intersect, setdiff, setequal, union

	BIRTH_YEAR	COUNTY_ORIGIN	BIRTH_MONTH	CALF_BREED_TYPE	TOTAL_BIRTH
1	2016	CARLOW	4	BEEF	3670
2	2016	CARLOW	4	DAIRY	680
3	2016	CARLOW	8	BEEF	778
4	2016	CARLOW	8	DAIRY	80
5	2016	CARLOW	12	BEEF	835
6	2016	CARLOW	12	DAIRY	123

PART:1

#### summary(data)

```
BIRTH_YEAR
              COUNTY_ORIGIN
                                   BIRTH_MONTH
                                                  CALF_BREED_TYPE
                                        : 1.00
Min.
       :2016
              Length: 624
                                  Min.
                                                  Length: 624
1st Qu.:2016
              Class :character
                                  1st Qu.: 3.75
                                                  Class :character
              Mode :character
Median :2016
                                  Median : 6.50
                                                  Mode :character
Mean
       :2016
                                  Mean : 6.50
3rd Qu.: 2016
                                  3rd Qu.: 9.25
                                  Max.
                                        :12.00
       :2016
Max.
```

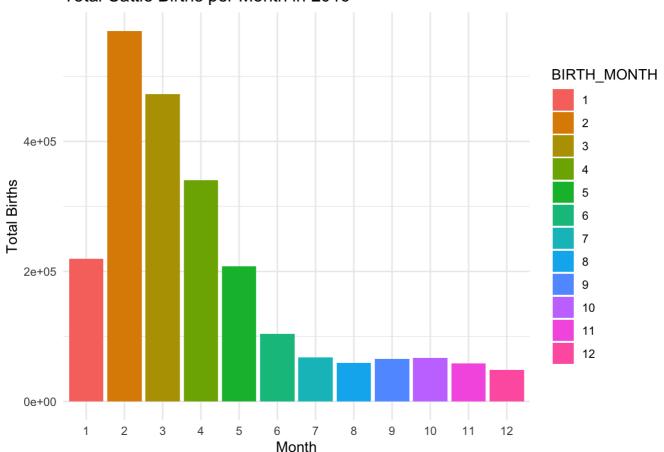
## TOTAL\_BIRTH

Min. : 6.0 1st Qu.: 454.5 Median : 1561.5 Mean : 3655.0 3rd Qu.: 3825.8 Max. :95720.0

```
library(ggplot2)
library(dplyr)

# Plot 1: Total Cattle Births per Month in 2016
data %>%
    group_by(BIRTH_MONTH) %>%
    summarise(TOTAL_BIRTH = sum(TOTAL_BIRTH)) %>%
    mutate(BIRTH_MONTH = factor(BIRTH_MONTH, levels=c(1,2,3,4,5,6,7,8,9,10,11,12))) %>%
    ggplot(aes(x = BIRTH_MONTH, y = TOTAL_BIRTH, fill = BIRTH_MONTH)) +
    geom_bar(stat = "identity") +
    theme_minimal() +
    labs(title = "Total Cattle Births per Month in 2016", x = "Month", y = "Total Births")
```

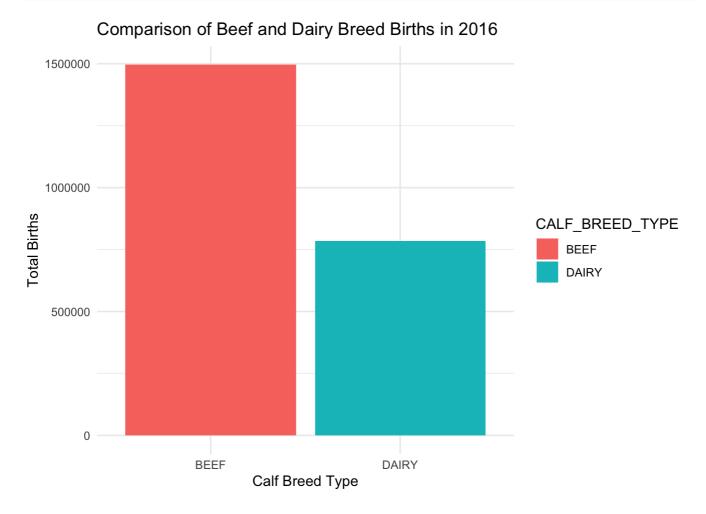
# Total Cattle Births per Month in 2016



```
# Plot 2: Comparison of Beef and Dairy Breed Births in 2016
data %>%
  group_by(CALF_BREED_TYPE) %>%
  summarise(TOTAL_BIRTH = sum(TOTAL_BIRTH)) %>%
  ggplot(aes(x = CALF_BREED_TYPE, y = TOTAL_BIRTH, fill = CALF_BREED_TYPE)) +
  geom_bar(stat = "identity") +
```

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theme\_minimal() + labs(title = "Comparison of Beef and Dairy Breed Births in 2016", <math>x = "Calf Breed Ty



The bar chart illustrates the total number of cattle births for each month in the year 2016, revealing a seasonal pattern in cattle breeding. The most significant number of births occur in February, with nearly 400,000, followed closely by March. After the spring peak, there's a steady decline in births from April through to the end of the year. The chart indicates that the cattle industry likely plans breeding schedules to coincide with favorable conditions, leading to a concentration of births in the early part of the year. The consistency in birth numbers from June to December suggests a deliberate management of breeding cycles to suit operational or environmental conditions.

The bar chart "Comparison of Beef and Dairy Breed Births in 2016" presents a side-by-side comparison of the total number of births for beef and dairy cattle breeds within the year. The red bar represents beef breeds and towers significantly, indicating over 1,000,000 births. In contrast, the blue bar represents dairy breeds, showing more than 500,000 births but considerably fewer than beef breeds. The disparity suggests a higher number of beef breed cattle were born in 2016 compared to dairy breeds, which could be attributed to various factors such as market demand or specific breeding programs within the cattle industry during that year. The chart clearly demonstrates the predominance of beef breed births over dairy in the recorded period.

### PART:2

The `janitor` package in R significantly streamlines data cleaning and preparation, making it a valuable asset for data analysts and researchers. Key features include `clean\_names()`, which standardizes column names for consistency, and `tabyl()`, which simplifies the creation of frequency tables for data examination. It also offers robust tools for managing duplicates, reordering and

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cleaning factor levels, and performing advanced cross-tabulations. Additionally, the package provides functions for decorative formatting, like adding totals and subtotals to tables, and facilitates flexible data type conversions. Renowned for its user-friendly and intuitive functions, 'janitor' enhances the efficiency and effectiveness of data cleaning processes, making it a popular choice in the R community.

```
#Package
library(janitor)
```

Attaching package: 'janitor'

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

chisq.test, fisher.test

```
#Function 1
cattle_birth <- data %>% clean_names()
head(cattle_birth)
```

```
birth_year county_origin birth_month calf_breed_type total_birth
1
        2016
                      CARLOW
                                        4
                                                                    3670
                                                       BEEF
2
        2016
                     CARLOW
                                        4
                                                                     680
                                                      DAIRY
3
                                        8
                                                                     778
        2016
                     CARLOW
                                                       BEEF
4
                                                                      80
        2016
                      CARLOW
                                        8
                                                      DAIRY
5
        2016
                      CARLOW
                                       12
                                                       BEEF
                                                                     835
6
        2016
                      CARLOW
                                       12
                                                      DAIRY
                                                                     123
```

```
#Function 2
breed_frequency <- cattle_birth %>% tabyl(calf_breed_type)
print(breed_frequency)
```

```
calf_breed_type n percent
BEEF 312 0.5
DAIRY 312 0.5
```

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```
5
             207919
    6
             103965
    7
              67717
    8
              59592
    9
              65185
   10
              66817
   11
              58691
   12
              48352
Total
            2280698
```

#### PART:3

```
library(dplyr)
# Define the S3 class
setClass("CattleStats", slots = c(stats = "data.frame"))
# The main function
analyzeCattleBirths <- function(data, group_by) {</pre>
  require(dplyr)
  # Calculating statistics
  stats <- data %>%
    group_by(!!sym(group_by)) %>%
    summarise(Mean = mean(TOTAL_BIRTH),
              Median = median(TOTAL_BIRTH),
              Std_Dev = sd(TOTAL_BIRTH)) %>%
    as.data.frame()
  # Creating S3 object
  stats_obj <- new("CattleStats", stats = stats)</pre>
  class(stats_obj) <- "CattleStats"</pre>
  return(stats_obj)
}
# Print method for CattleStats
print.CattleStats <- function(object) {</pre>
  print(object@stats)
}
# Summary method for CattleStats
summary.CattleStats <- function(object) {</pre>
  summary(object@stats)
}
# Plot method for CattleStats
plot.CattleStats <- function(object, group_by, ...) {</pre>
  barplot(as.matrix(object@stats[, -1]), beside = TRUE,
          main = "Cattle Birth Statistics",
          ylab = "Value", xlab = group_by,
          col = rainbow(nrow(object@stats)),
          legend.text = rownames(object@stats), ...)
}
# Now execute the function to create the result object
result <- analyzeCattleBirths(data, "BIRTH MONTH")</pre>
```

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```
# Now you can print, summarize, and plot the result
print(result)
```

```
BIRTH_MONTH
                   Mean Median
                                  Std_Dev
1
            1 4225.3077 2345.5 6666.5186
2
            2 10959.6731 5747.0 15630.9017
3
              9090.6538 6012.5 9642.6629
4
            4 6540.9038 4041.0 7228.8914
5
            5
               3998.4423 2027.5 4431.4446
6
            6 1999.3269 884.5 2268.3274
7
            7 1302.2500 601.0 1461.9106
8
              1146.0000 634.0 1243.7477
            8
9
            9 1253.5577 909.5 1138.9503
10
           10 1284.9423 928.0 1047.1170
11
           11 1128.6731 868.0
                                908.3569
12
           12
                929.8462 696.5
                                 807.0202
```

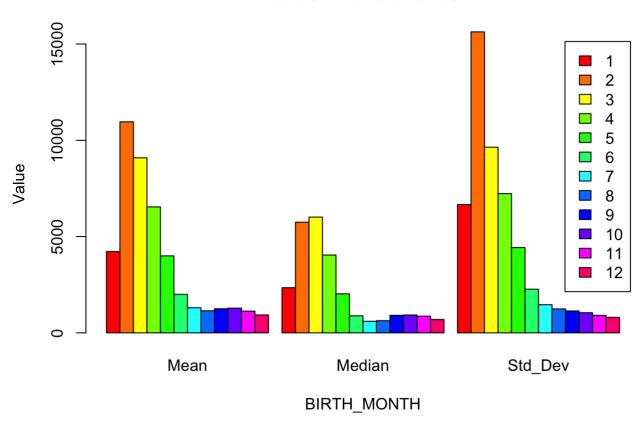
## summary(result)

BIRTH_MONTH	Mean	Median	Std_Dev
Min. : 1.00	Min. : 929.8	Min. : 601.0	Min. : 807
1st Qu.: 3.75	1st Qu.: 1226.7	1st Qu.: 825.1	1st Qu.: 1116
Median : 6.50	Median : 1650.8	Median : 918.8	Median : 1865
Mean : 6.50	Mean : 3655.0	Mean :2141.2	Mean : 4373
3rd Qu.: 9.25	3rd Qu.: 4804.2	3rd Qu.:2769.4	3rd Qu.: 6807
Max. :12.00	Max. :10959.7	Max. :6012.5	Max. :15631

plot(result, group\_by = "BIRTH\_MONTH") # The corrected call to your plot method

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# **Cattle Birth Statistics**



The "Cattle Birth Statistics" bar chart compares the mean, median, and standard deviation of cattle births by month for an unspecified year. The mean and median values peak in February, suggesting that this month typically has the highest average number of births. The standard deviation is greatest in January, indicating a larger variability in the number of births during this month compared to others. This could reflect a seasonal breeding pattern, with a more consistent number of births around the peak month and greater fluctuations at the beginning of the year. Overall, the chart provides a clear visual representation of cattle birth trends and variability throughout the year.

#### Citation

```
citation("janitor")

To cite package 'janitor' in publications use:

Firke S (2023). _janitor: Simple Tools for Examining and Cleaning
   Dirty Data_. R package version 2.2.0,
   <a href="https://CRAN.R-project.org/package=janitor">https://CRAN.R-project.org/package=janitor</a>.

A BibTeX entry for LaTeX users is

@Manual{,
   title = {janitor: Simple Tools for Examining and Cleaning Dirty Data},
   author = {Sam Firke},
   year = {2023},
   note = {R package version 2.2.0},
   url = {https://CRAN.R-project.org/package=janitor},
}
```

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