# Assignment 1 for Data Analysis 2 and Coding with R

### Pauline Broussolle

#### 2020-11-24

#### Introduction

We analyse covid data from CSSE at Johns Hopkins University and we use also population data for 2019 from World Bank (using WDI). The dataset contains 182 countries and registers the number of confirmed covid cases and number of death due by covid on 15/10/2020.

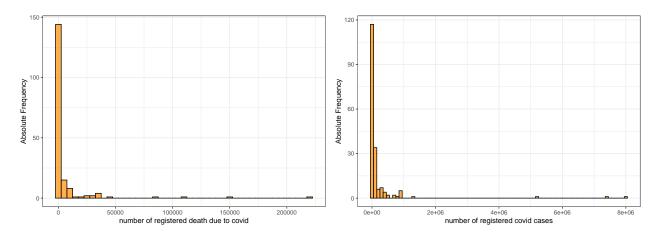
My outcome variable is the **number of registered death due to covid** and my explanatory variable: **number of registered cases**. My goal is to analyse the pattern of association between registered covid-19 cases and registered number of death due to covid-19 on 15/10/2020.

### Executive summary

## 1. Summary statistics and Distribution for x and y

Table 1: Summary for the number of registered death caused by covid and registered covid cases

mean	median	min	max	std	variable
213757 $6032$	16483.0 281.5	3			confirmed cases nb of death



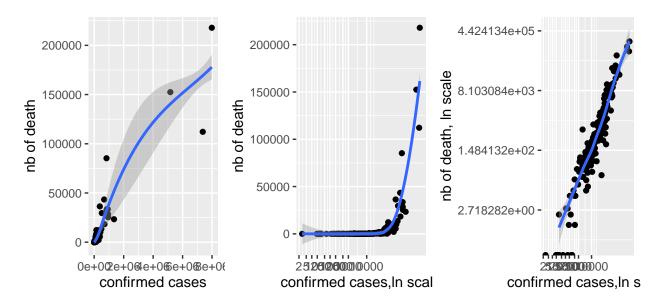
2-3 sentence, explain the main features and distribution - use histograms and summary statistics table (mean, median, min, max, standard deviation) We see that there are a few extreme values.

#### Chech extreme values

We check countries which have confirmed cases aboves 2 milion and registered number of death above 50,000. These are India, Brazil and United States, which are not measurement errors. We keep these values.

#### 2. Pattern of association

#### Transformation of the variables: Scaling, Take Logs?



For the simple model without scaling the pattern is non-linear, most of observations are concentrated and there are some extreme observations, corresponding to Brazil, US, India.

Make a substantive and statistical reasoning, where and when to use ln transformation. You do not need to fit any model here, only use statistical reasoning based on the graphs. i. Take care when it is possible to make ln transformation: you may need to drop or change some variables. We should use the log-log transformation, which is the only to provide a linear pattern. We create new variables which are ln\_confirmed and ln\_death

#### 3. Estimating different models

We choose the log transformation. We estimate four different regression models: Simple linear regression, Quadratic regression, Piecewise linear spline regression, Weighted linear regression (using population as weights).

#### Presentation of model choice

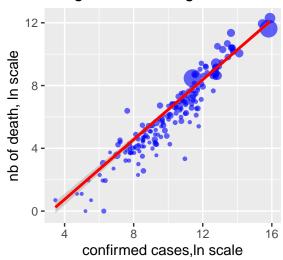
The model comparison (all the estimated model results) is reported in the appendix of the report.

The best model is Weighted linear regression, using population as weights: reg4:  $\ln_{death} = alpha + beta * \ln_{confirmed}$ , weights: population

```
##
## Call:
## lm_robust(formula = ln_death ~ ln_confirmed, data = df, weights = population)
```

```
##
## Weighted, Standard error type: HC2
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|) CI Lower CI Upper
                 -3.0661
                            0.78208
                                      -3.92 1.285e-04
                                                       -4.6100
                                                                  -1.522 168
##
  (Intercept)
## ln confirmed
                  0.9531
                            0.06364
                                      14.98 9.286e-33
                                                         0.8275
                                                                   1.079 168
## Multiple R-squared: 0.9285,
                                    Adjusted R-squared: 0.9281
## F-statistic: 224.3 on 1 and 168 DF, p-value: < 2.2e-16
## 'geom_smooth()' using formula 'y ~ x'
```

# Weighted linear regression, usir



Compare the models and choose your preferred one Use substantive and statistical reasoning for your chosen model. ii. Show the model results in the report along with the graph.

## 4. Hypothesis Test and Analysis of the residuals

```
##
## Call:
## lm_robust(formula = ln_death ~ ln_confirmed, data = df, weights = population)
## Weighted, Standard error type: HC2
##
##
  Coefficients:
                Estimate Std. Error t value Pr(>|t|) CI Lower CI Upper DF
##
                            0.78208
                                      -3.92 1.285e-04
                                                       -4.6100
## (Intercept)
                 -3.0661
                                                                  -1.522 168
## ln_confirmed
                  0.9531
                            0.06364
                                      14.98 9.286e-33
                                                        0.8275
                                                                   1.079 168
## Multiple R-squared: 0.9285,
                                    Adjusted R-squared: 0.9281
## F-statistic: 224.3 on 1 and 168 DF, p-value: < 2.2e-16
```

The estimated t-statistics is 14.98, with p-value: 9.286e-33. Thus we reject the H0, which means that number of recorded death due to covid is not uncorrelated with number of confirmed covid cases.

Table 2: Countries with largest negative errors

country	$\ln_{-death}$	${\rm reg4\_y\_pred}$	reg4_res
Burundi	0.000000	2.910794	-2.910794
Iceland	2.302585	4.799315	-2.496730
Qatar	5.402677	8.148109	-2.745432
Singapore	3.332205	7.385913	-4.053708
Sri Lanka	2.564949	5.097056	-2.532107

Table 3: Countries with largest positive errors

country	$\ln_{-death}$	${\rm reg4\_y\_pred}$	reg4_res
Ecuador	9.417842	8.295599	1.1222430
Italy	10.501554	9.183262	1.3182927
Mexico	11.353754	9.929485	1.4242690
United Kingdom	10.677823	9.728898	0.9489249
Yemen	6.390241	4.203259	2.1869819

# Conclusion

- We investigated ...
- $\bullet\,\,$  and we have found
  - X and Y are ... correlated
- Our analysis can be
  - strengthened by...
  - weakened by  $\dots$

# Appendix

	Linear model
(Intercept)	-4.29*
	[-4.88; -3.69]
$ln\_confirmed$	$1.03^{*}$
	[0.97; 1.09]
$\mathbb{R}^2$	0.89
$Adj. R^2$	0.89
Num. obs.	170
RMSE	0.83

 $<sup>^{*}</sup>$  0 outside the confidence interval.

"