Data Analysis 2 - Assigment 2

Modelling Fish Weights based on Length, Height, Width and Species

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Introduction

The objective of the analysis is to develop a multiple regression model predicting the **Weight** in grams (dependent variable) of a fish using its physical dimensions: Height, Width, Lengths (1,2,3) and its specie type. (Explanatory Variables).

The Lengths are continuous variables measured in centimeters (cm).

The Height is the maximum height as % of Length3 (Height = Height/Length3 * 100)

The Width is the maximum height as % of Length3 (Width = Width/Length3 * 100)

There are 7 species in the sample, with different numbers of observations of each.

The variable Species was transformed into dummy variables named from (Specie1 to Specie7)

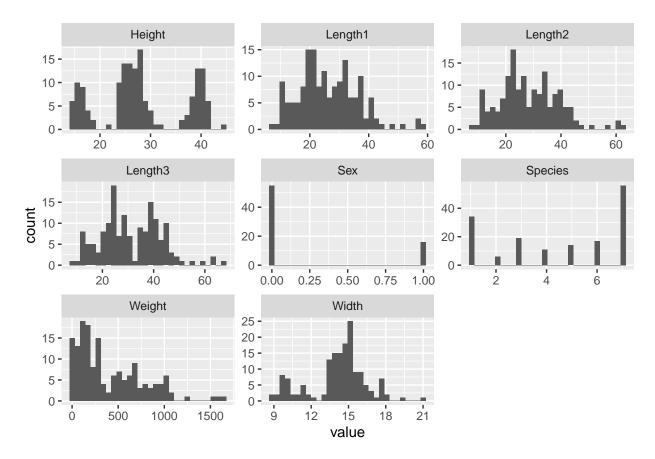
The data set is sorted by species from 1 to 7 and then approximately by length within a species.

Variables Descriptive Analysis

Summary Statistics Table (without Species Dummy Variables)

Species	Weight	Length1	Length2	Length3	Height	Width	Sex
Min.	Min. : 5.9	Min.:	Min.:	Min.:	Min.	Min.:	Min.
:1.000		7.50	8.40	8.80	:14.50	8.70	:0.0000
1st	1st Qu.:	1st	1st	1st	1st	1st	1st
Qu.:2.000	120.0	Qu.:19.10	Qu.:21.00	Qu.:23.20	Qu.:24.20	Qu.:13.40	Qu.:0.0000
Median	Median:	Median	Median	Median	Median	Median	Median
:5.000	273.0	:25.20	:27.30	:29.40	:26.90	:14.60	:0.0000
Mean	Mean:	Mean	Mean	Mean	Mean	Mean	Mean
:4.529	401.2	:26.27	:28.44	:31.24	:28.26	:14.12	:0.2253
3rd	3rd Qu.:	3rd	3rd	3rd	3rd	3rd	3rd
Qu.:7.000	650.0	Qu.:32.70	Qu.:36.00	Qu.:39.70	Qu.:37.80	Qu.:15.30	Qu.:0.0000
Max.	Max.	Max.	Max.	Max.	Max.	Max.	Max.
:7.000	:1650.0	:59.00	:63.40	:68.00	:44.50	:20.90	:1.0000
NA	NA	NA	NA	NA	NA	NA	NA's :86

Check the histograms distribution (without Species Dummy Variables)



Considering the analysis of the graphs and the summary statistics, it's possible to visualize that:

Whether **Height and Width** are closer to the regular **normal distribution**, the **Length(1,2,3)** and **Weight** variables are slightly skewed forming **Right-Tails** with the presence of some outliers.

Taking that into account, it's possible to apply some log transformations while the models are build, specially at the dependent variable **Weight** and the explanatory **Length**.

Variable **Sex** shouldn't be used for analysis because it has a good amount of **missing values** that may impact the model.

Variables Correlation Matrix

	Species	Weight	Length1	Length2	Length3	Height	Width
Species	1.0000000	-0.1327455	-0.0235951	-0.0355609	-0.1357486	-0.6619026	0.0891793
Weight	-0.1327455	1.0000000	0.9164552	0.9193671	0.9245350	0.1943756	0.1355157
Length1	-0.0235951	0.9164552	1.0000000	0.9995161	0.9921198	0.0351444	0.0317880
Length2	-0.0355609	0.9193671	0.9995161	1.0000000	0.9941898	0.0547546	0.0442597
Length3	-0.1357486	0.9245350	0.9921198	0.9941898	1.0000000	0.1326425	0.0377342
Height	-0.6619026	0.1943756	0.0351444	0.0547546	0.1326425	1.0000000	0.4560633
Width	0.0891793	0.1355157	0.0317880	0.0442597	0.0377342	0.4560633	1.0000000

As the correlation results between the Length's showed pretty much closer to 1 with significant p-values:

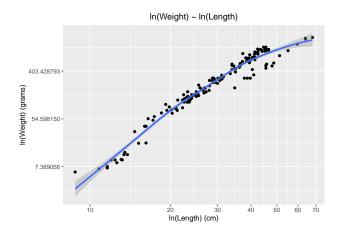
^{**} Length1 ~ Length3: r = 0.99 and p = 0.000.

```
** Length2 ~ Length3: r = 0.99 and p = 0.000.
```

Having the three of them at the same regression model may lead to errors and won't produce significant better adjustments. With that, the variables **Length1** and **Length2** won't be considered at any analysis. Only the variable **Length3**, the total size of the fish, is going to be used as the **Final_Length** measurement.

Weight and Length

 $ln(Weight) \sim ln(Final_Length)$



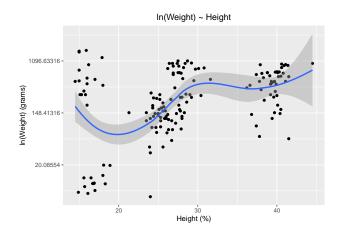
According to the results obtained from the graphs and model comparison (please review the following Appendix section for more detailed information), the log-log regression shows not only a better fit, but also an extraordinary R-squared adjusted with a really significant p-value. This is supported by the Right-Tailed Distributions Patterns encountered both in Weight and Length. The variable Length itself could be used alone for the predictions.

Considering that, the first regression model for analysis is:

Reg1: $\ln(\text{Weight}) \sim \ln(\text{Final Length})$

Weight and Height

$ln(Weight) \sim Height$



^{**} Length1 ~ Length2: r = 1.00 and p = 0.000.

The analysis of the graphs and the comparisons between the **models with and without** the log-transformations (please review the following **Appendix section** for more detailed information), didn't show **any significant statistic difference**. The **pattern** expressed in both graphs are quite **similar**, and the results obtained in terms of **R-squared** are pretty **close** as well. With that, due to the **ease and more tangible interpretation** of the results, the model **chosen** was the **log-lvl** one.

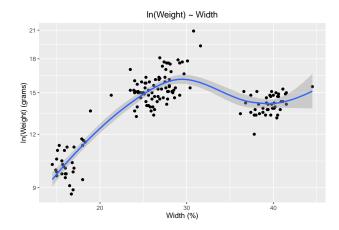
Another important aspect to highlight is the fact that the pattern observed can possibly fit a **quadratic** or cubic polynomial models. Although the cubic and quadratic models showed a slightly better fit than the simple log-lvl, the interpretation is much easier under the simple model. The differences aren't that significant enough in terms R-Squared as well.

Considering that, the second regression model for analysis is:

Reg2: $ln(Weight) \sim ln(Final_Length) + Height$

Weight and Width

$ln(Weight) \sim Width$



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Considering that, the second regression model for analysis is: Reg5: ln(Weight) ~ ln(Final Length) + Height + Width

Weight and Species

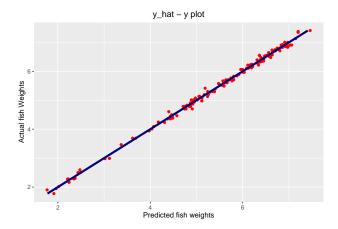
Considering what was done so far, the best possible model so far is represented by reg5. However, it's still missing the impact of the species variables. Another regression model can actually be created for that situation:

Reg8: ln(Weight) ~ ln(Final Length) + Height + Width + as.factor(Species)

Adding the dummy variables created for species, the final model, reg8, shows itself really representative with higher fit and prediction capabilities. Although the R-squared didn't present any substantive statistical difference from the previous models, the p-values were highly significant. It means that the model explains well the dataset chosen, using a good number of really representative variables.

For all of that, the **reg8** is going to be the model **used for predictions**. (Please check the **final table resume** at page number 15 - **Appendix**)

Predictions



As expected the model **predicted very well** the values for **Weight**. The **residuals** are pretty low and the graph shows small discrepancy between the actual and the predicted values. It's important to mention that that the **residuals were evaluated in terms of the log transformations**, as well the values **predicted**. The **conversion** from **logarithmic Weight to absolute values should be done carefully**, considering some possible error bias. The **variables** presented themselves significant (**lower p-values**) and important for the model description. The **R-squared value** is really **representative** and generates a great understanding of how the dependent variable **Weight is changing in the dataset**.

Conclusion

The main goal of this project, create a reasonable model that could predict the fish Weight based on its physical characteristics, was achieved successfully. Considering the high R-Squared and lower p-values, the model was build with a strong comprehend of the dataset structure and patterns generating on that way really accurate predictions.

Although it's **impossible** to infer any **causality**, this model seems a **good income** for **further studies** in the area.

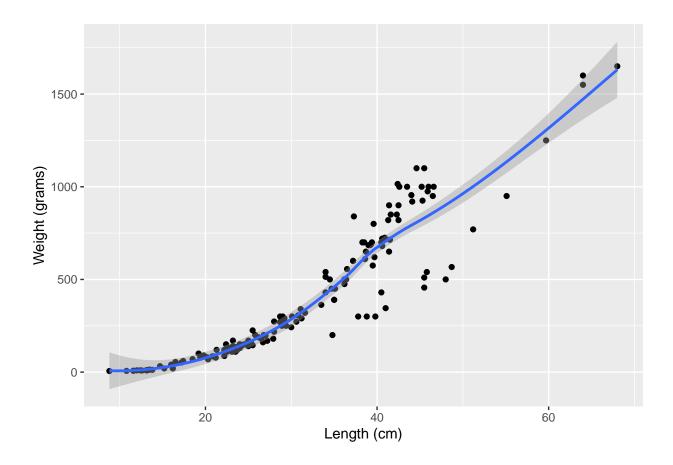
Appendix

Correlation Matrix

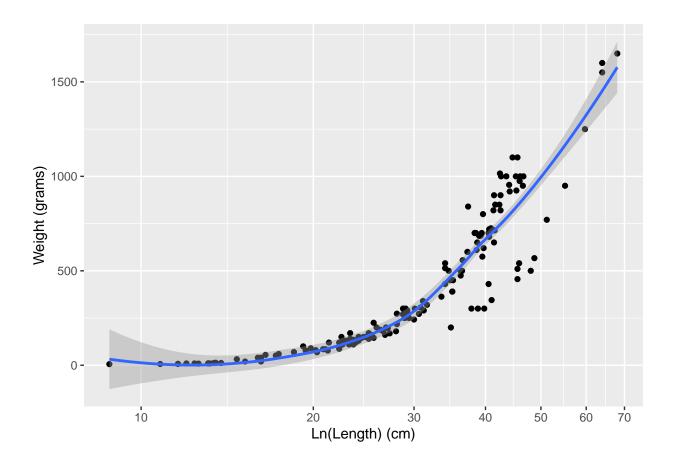
```
Species
                          Weight
                                    Length1
                                                Length2
                                                           Length3
                                                                         Height
## Species 1.00000000 -0.1327455 -0.02359505 -0.03556088 -0.13574860 -0.66190261
## Weight -0.13274545 1.0000000 0.91645518 0.91936709 0.92453500 0.19437562
## Length1 -0.02359505 0.9164552 1.00000000 0.99951615 0.99211983 0.03514437
## Length2 -0.03556088 0.9193671 0.99951615 1.00000000 0.99418981 0.05475455
## Length3 -0.13574860 0.9245350 0.99211983 0.99418981 1.00000000 0.13264248
## Height -0.66190261 0.1943756 0.03514437 0.05475455 0.13264248 1.00000000
## Width
           0.08917930 \quad 0.1355157 \quad 0.03178795 \quad 0.04425967 \quad 0.03773417 \quad 0.45606333
##
               Width
## Species 0.08917930
## Weight 0.13551573
## Length1 0.03178795
## Length2 0.04425967
## Length3 0.03773417
## Height 0.45606333
## Width 1.0000000
             Species
                        Weight Length1 Length2
                                                      Length3
## Species
                  NA 0.09744931 0.7692749 0.6583765 0.09003959 0.000000e+00
                           NA 0.0000000 0.0000000 0.00000000 1.471291e-02
## Weight 0.09744931
## Length1 0.76927491 0.00000000 NA 0.0000000 0.00000000 6.621304e-01
## Length2 0.65837646 0.00000000 0.0000000 NA 0.00000000 4.958074e-01
## Length3 0.09003959 0.00000000 0.0000000 0.0000000
                                                           NA 9.771168e-02
## Height 0.00000000 0.01471291 0.6621304 0.4958074 0.09771168
          0.26669809 0.09059758 0.6926813 0.5820380 0.63892903 1.943337e-09
## Width
                 Width
## Species 2.666981e-01
## Weight 9.059758e-02
## Length1 6.926813e-01
## Length2 5.820380e-01
## Length3 6.389290e-01
## Height 1.943337e-09
## Width
                    NA
```

Weight and Length

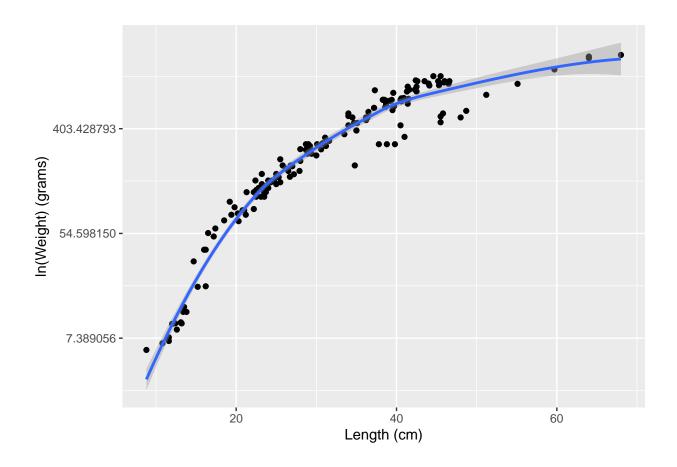
```
## 'geom_smooth()' using formula 'y ~ x'
```



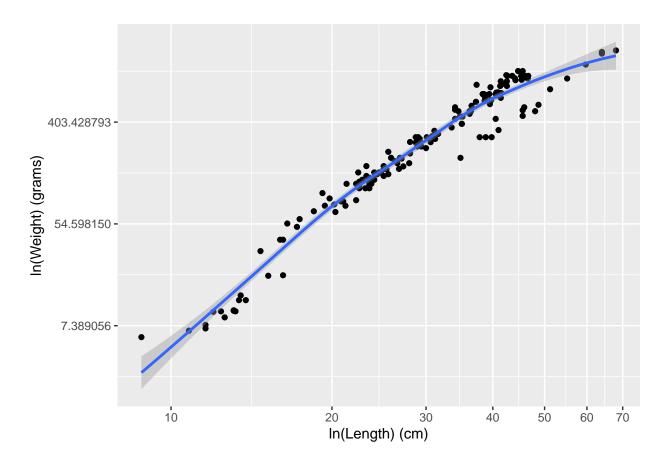
'geom_smooth()' using formula 'y ~ x'



'geom_smooth()' using formula 'y ~ x'



'geom_smooth()' using formula 'y ~ x'

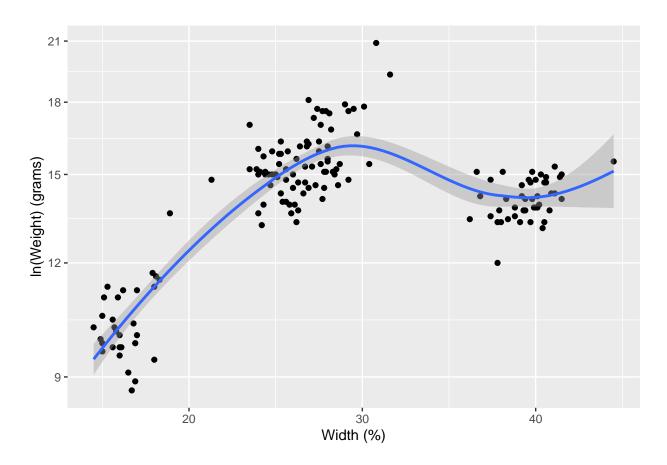


```
##
    studentized Breusch-Pagan test
##
##
## data: lvl_lvl
## BP = 28.102, df = 1, p-value = 1.151e-07
##
##
    studentized Breusch-Pagan test
##
## data: lvl_log
## BP = 6.9663, df = 1, p-value = 0.008306
##
##
    studentized Breusch-Pagan test
##
## data: log_lvl
## BP = 0.0068874, df = 1, p-value = 0.9339
##
##
    studentized Breusch-Pagan test
##
## data: log_log
## BP = 0.18215, df = 1, p-value = 0.6695
```

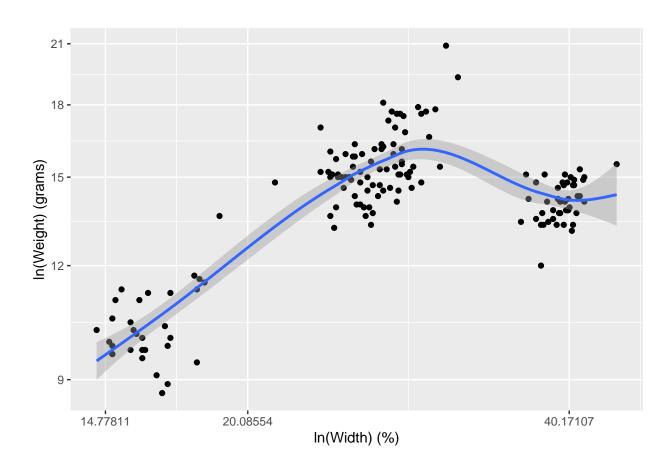
The table was written to the file 'D:/CEU/Data_Analysis_2/DA2_Final_Project/out/logtrans_Weight_Leng

Weight and Width

'geom_smooth()' using formula 'y ~ x'



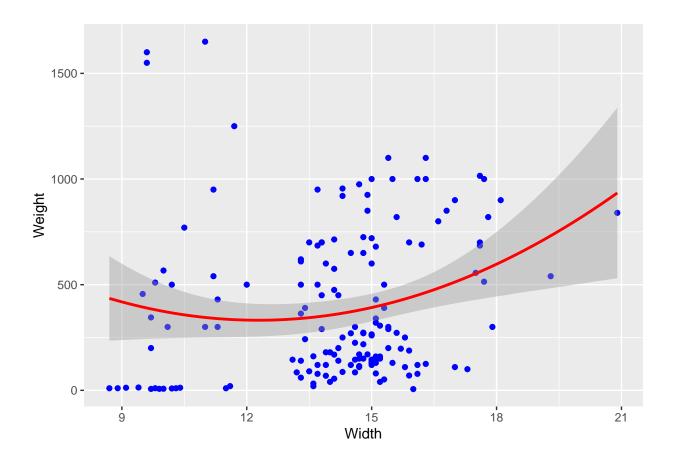
'geom_smooth()' using formula 'y ~ x'

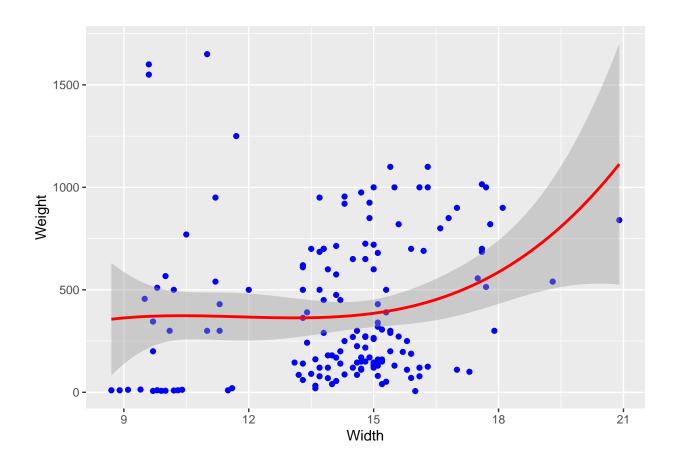


```
##
## studentized Breusch-Pagan test
##
## data: log_lvl
## BP = 40.838, df = 1, p-value = 1.654e-10

##
## studentized Breusch-Pagan test
##
## data: log_log
## BP = 42.731, df = 1, p-value = 6.282e-11
```

The table was written to the file 'D:/CEU/Data_Analysis_2/DA2_Final_Project/out/logtrans_Weight_Widt





	reg1	reg2	reg5	reg8
(Intercept)	-5.27 ***	-5.58 ***	-6.46 ***	-5.87 ***
	(0.29)	(0.17)	(0.09)	(0.13)
ln_Final_Length	3.17 ***	3.03 ***	3.00 ***	3.02 ***
	(0.08)	(0.05)	(0.02)	(0.04)
Height		0.03 ***	0.02 ***	0.02 ***
		(0.00)	(0.00)	(0.01)
Width			0.09 ***	0.04 ***
			(0.00)	(0.01)
Species1				-0.28 **
				(0.10)
Species2				0.02
				(0.03)
Species3				-0.10 ***
				(0.02)
Species4				-0.13
				(0.10)
Species5				-0.33 ***
				(0.05)
Species6				-0.24 ***
				(0.07)
nobs	157	157	157	157
r.squared	0.95	0.97	0.99	1.00
adj.r.squared	0.95	0.97	0.99	1.00
statistic	1399.67	2186.99	6642.68	4658.18
p.value	0.00	0.00	0.00	0.00
df.residual	155.00	154.00	153.00	147.00
nobs.1	157.00	157.00	157.00	157.00
se_type	HC2.00	HC2.00	HC2.00	HC2.00

^{***} p < 0.001; ** p < 0.01; * p < 0.05.