

# Predicting bird classification by bone lengths

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## Classification of differing ecological bird populations

### Introduction

### Materials & Methods

During this research we used data from an external source, used weka 3.8.0 for classification and R version 3.5.1 in combination with Rstudio for data analysis.

### Data

Birds' Bones and Living Habits, Kaggle dataset

Bone measurements were measured from a skeleton collection of Natural History Museum of Los Angeles County, provided by Dr. D. Liu of Beijing Museum of Natural History

### R and Rstudio

R was used for the exploration of the data, cleaning of the data and statistical analysis. In our EDA(Exploratory Data Analysis) we looked at the structure of the data and the distribution of the data using basic graphs with the ggplot2 library.

Cleaning was done based on missing values which we all removed & on outliers which we also removed,

And lastly we compared different classification algorithms again using plots from ggplot2 to choose the correct algorithm.

### weka 3.8.0

Weka (Waikato Environment for Knowledge Analysis) is a free to use datamining software written in Java. It is a Java application that is capable of doing lots of things applicable to datamining. We will be using it to determine what is the best classification algorithm for our usecase. As we will be creating a Java application of our own that can be used to classify the different bird groups.

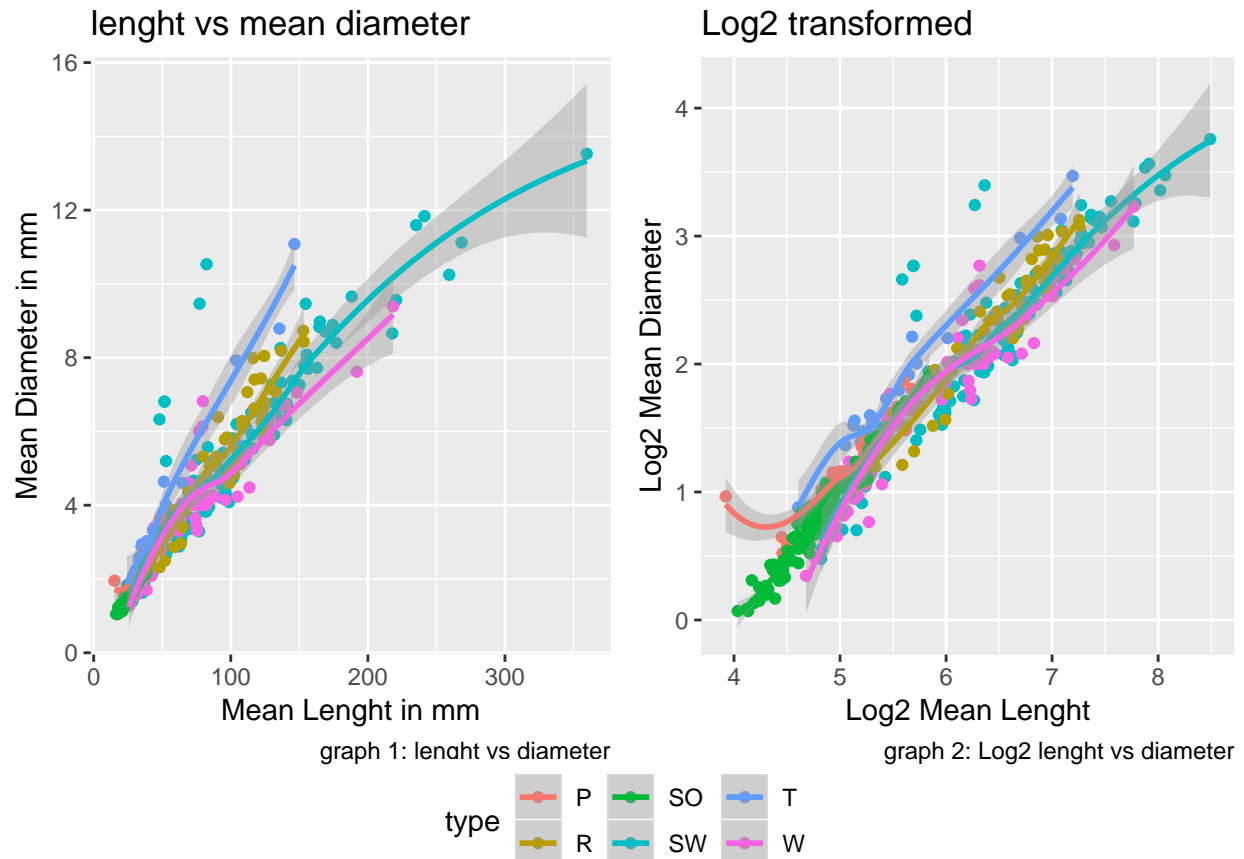
- First we use the explorer to make our datafiles usable in weka by removing duplicate id columns
- when the data is ready we can use the classify module in weka to test different classification algorithms that have reasonable accuracy's.
- at last when we found some interesting algorithms we can use the experimenter to try different settings to find optimal ones.

Every algorithm their initial output is a confusion matrix. This matrix is the measurement that we use to compare different algorithms. they are also inspected using a learning & ROC curve's.

- A Learning curve shows the false positive rate over how much data is provided to the learning algorithm
- A ROC curve shows the Sensitivity over the Specificity.

## Results

### EDA

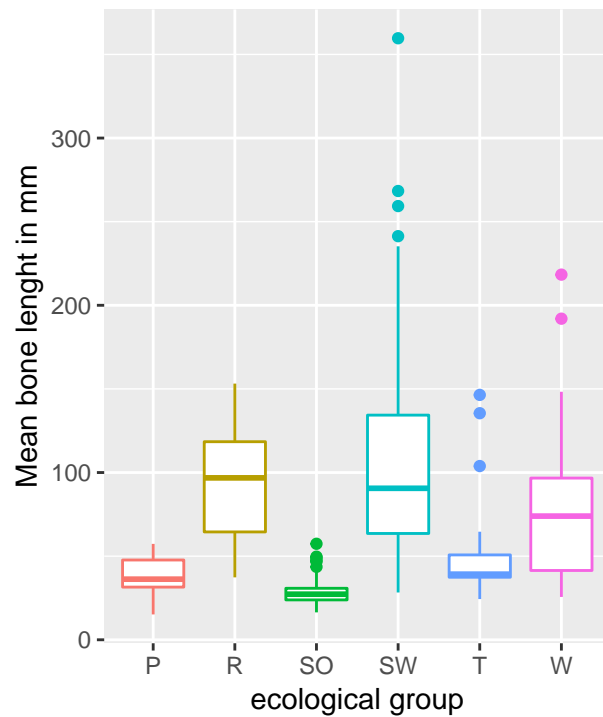


Graph 1 & 2 show the bone length distribution. Graph 1 is the distribution without any changes. Here we see that the Swimming birds have quite large bones, which could be because of abnormally big bird samples. as you look closer at the swimming birds in graph 1 and 2 we see that there might be a lot of noise in these samples because there are a lot of points outside the main distribution. For the other groups we also see that the Scansorial Birds & the Sininging birds are the smaller groups.

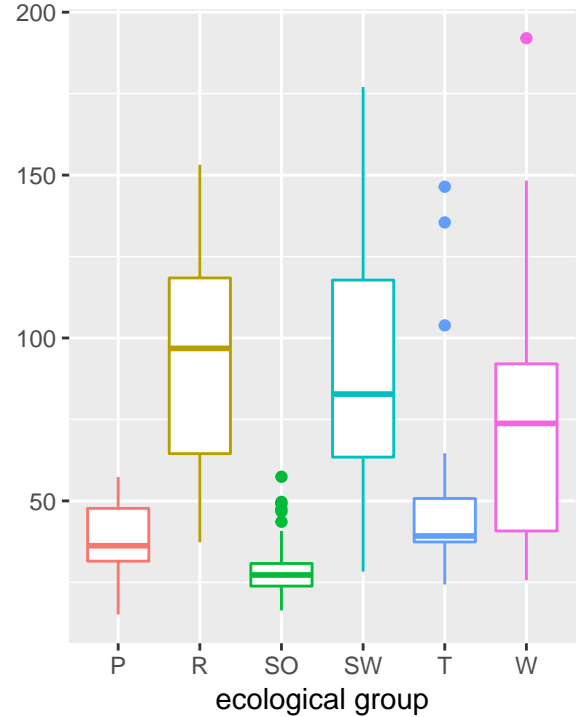
Terrestrial Birds, Raptors and Wading Birds seem to be in the middle of the rest with the Terrestrial Birds having notisable thicker bones that the others.

## Data Cleaning

Boxplot for each ecological group:  
For Humerus, Ulna and Tibiotarsus



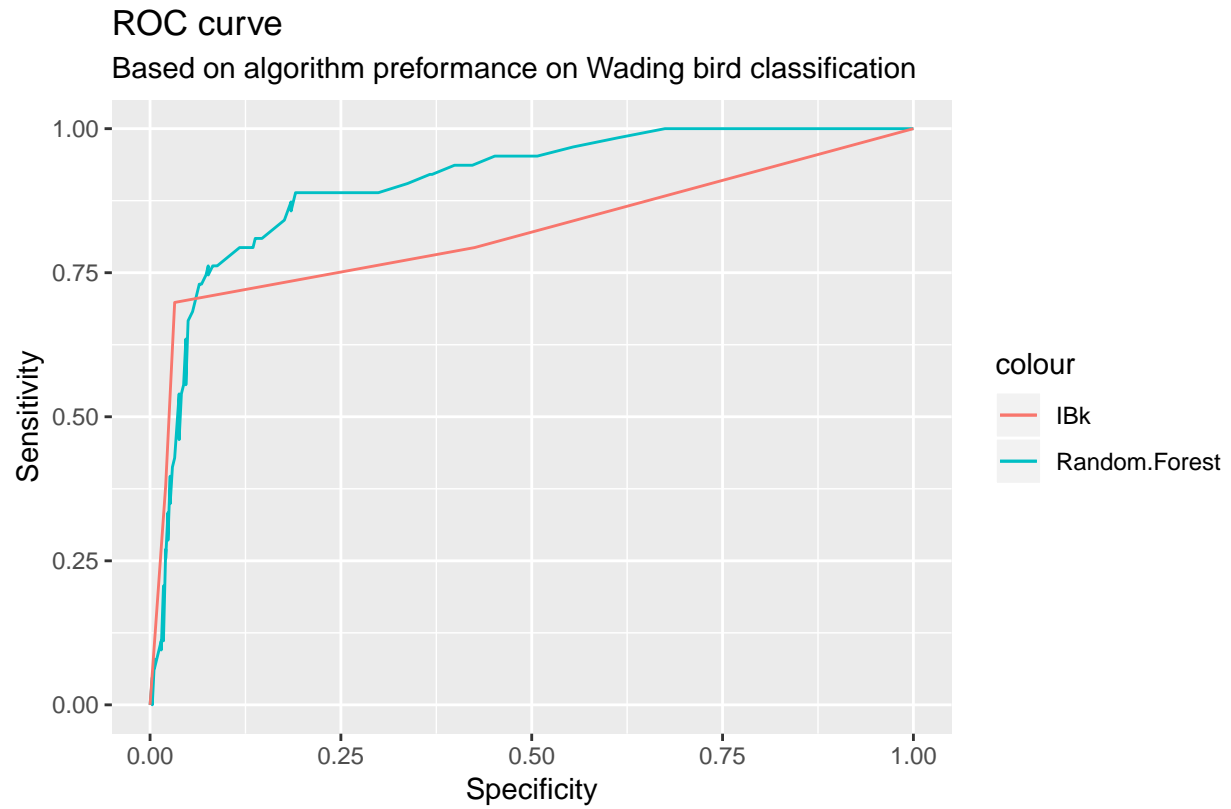
Boxplot for each ecological group:  
For Humerus, Ulna and Tibiotarsus



For our cleaning of data we removed all rows with NA's which cost us 7 datapoints. After that we calculated the 1st and 3rd quartile outliers from the humerus which can be seen in graph 3 & 4, and removed 9 rows. After all cleaning we left with 404 datapoints to use for Weka Analyses and classification.

## Weka

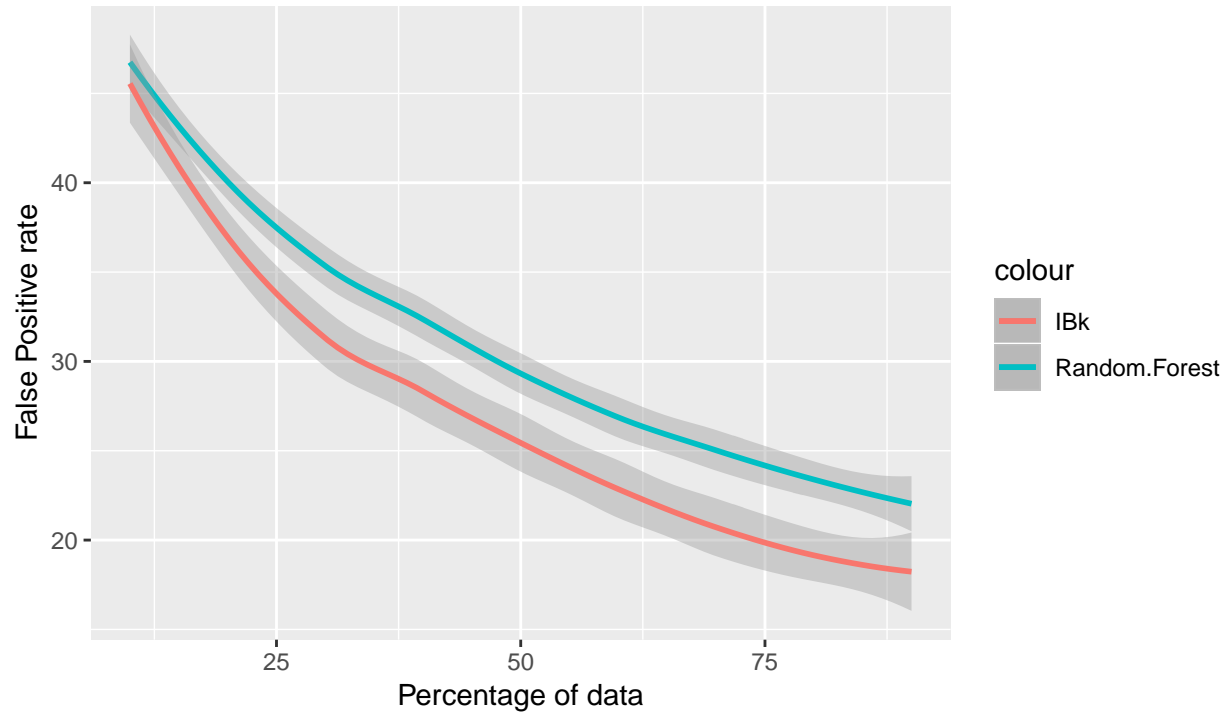
In weka the goal was to find a classification algorithm that has the highest possible accuracy because classifying a birds heritage it does not matter that much if there is a False positive, the goal was to keep false negatives low and keep true positives high.



```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
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```

## Learning curve

For Random.Forest & IBk



graph 7: Learning curve inspected algorithm's

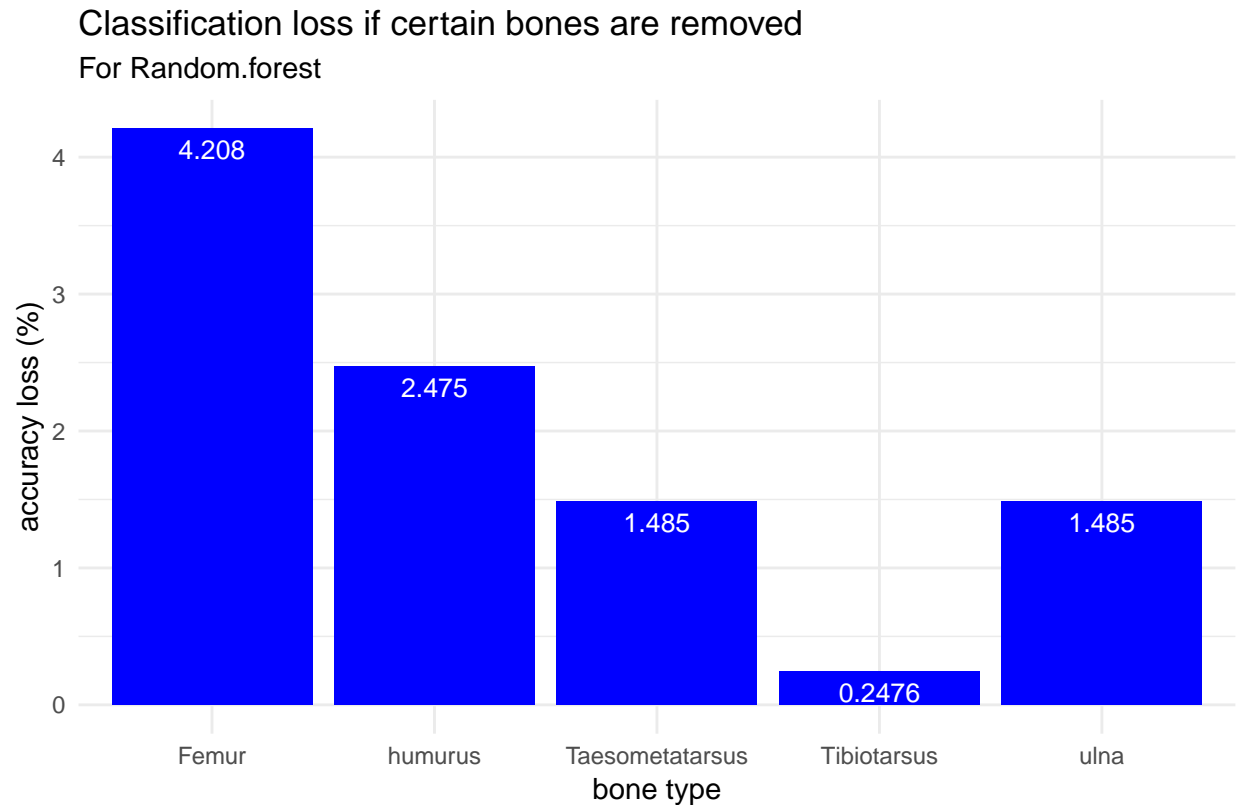
Chosen classifier: Random.Forest

Table 1: Random.Forest Confusion matrix as chosen classifier

	SW	W	T	R	P	SO
SW	87	11	0	4	0	6
W	17	34	0	2	4	6
T	2	0	12	2	4	3
R	7	1	0	37	3	0
P	0	0	1	1	30	6
SO	0	0	1	1	1	121

Table 2: Random.Forest as chosen classifier

	Correct	Incorrect
Instances	321	83
Percentage	79.4554 %	20.5446 %



graph 5: accuracy loss per bone

in graph 5 we can see what the importance is from certain bones if we are talking about classification. the differences we can see are because of their importance for the functionality of the bird groups.

## Conclusion & discussion

What is the most important bone for each ecological group their function? we can conclude that (At least for classification) the Femur is the most important as when we remove this from our classification algorithms the accuracy loss is great. we can see this in graph 5 where the loss per bone is displayed. while researching the subject we made the assumption that the Femur and Taesometatarsus were the least important for classification. Because of this the first classification algorithms were done only using the longer bones but further in reverted this decision. The classification algorithm that was chosen however wasn't changed. In the future we should first collect some data about the importance for some bones before making such rational decisions.

We do want to think about removing some bones as when we have less data to be put in the more unknown fossils with missing bones could possibly be classified

## Minor Proposal

## References