Predicting bird classification by bone lenghts

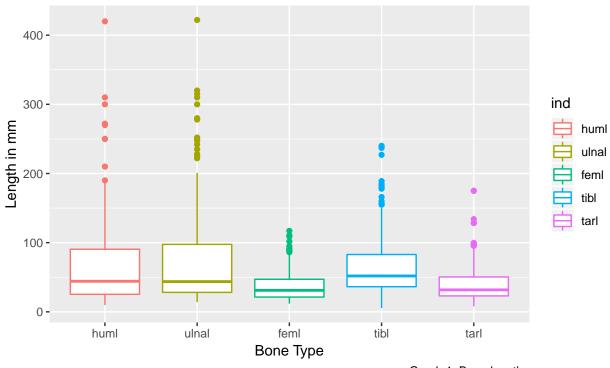
Jouke Profijt October 8, 2018

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Resutls

EDA

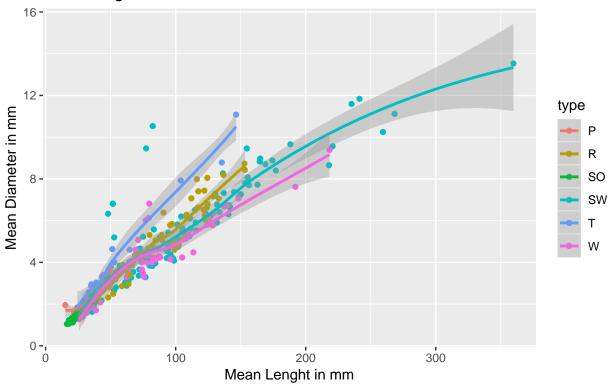
Boxplot for bone lenghts per bone in mm Variation between small and large bones.



Graph 1: Bone lengths

The Femuer as we can see in Graph 1 is the one of the smallest bones in all bird groups, 2nd to the Taesometatarsus. Because these 2 bones are so small there is not a lot of variation measurable between the diffrent bird groups and thus not that important for classification, that is why in this experiment there was chosen to leave these 2 bones out for classification.

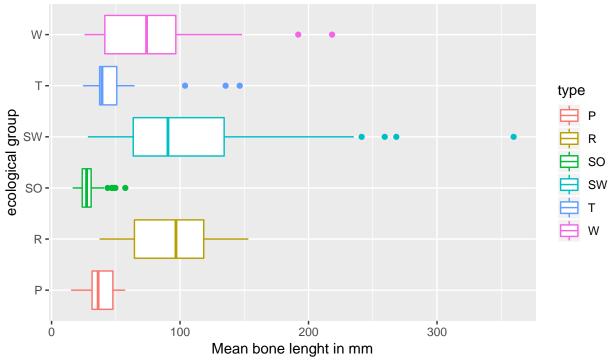
mean lenght vs mean diameter For Humerus, Ulna and Tibiotarsus



graph 2: lenght vs diameter

graph 2 shows us that the total bone length for the Swimming Birds is quite a bit larger than the other birds. Also we can see that the Terrestrial birds have thicker bones than the rest of the birds but are stull quite small. the singing birds are overall very small ans should be easy to classify. The raptors and Wading birds are a bit average and could become diffcult to classify but the raptors are a bit thicker and the Wading birds are a bit smaller.

Boxplot for each ecological group's mean bone lenght For Humerus, Ulna and Tibiotarsus



graph 3: mean bone lenghts

In graph 3 again we see the bigger birds seperated between their bonelenghts, and can quite easely distinguish each group. Altough their is some overlap between lenghts we can conclude that classification via lenght the birds bone lenght is probably the best way of classification.

Weka

In weka the goal was to find a classification algorithm that has the highest possible accuracy because classifying a birds herritage 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.

Zero R

Table 1: Zero R Confusion matrix

	SW	W	Т	R	Р	SO
$\overline{\mathbf{SW}}$	0	0	0	0	0	108
${f W}$	0	0	0	0	0	63
${f T}$	0	0	0	0	0	23
${f R}$	0	0	0	0	0	48
${f P}$	0	0	0	0	0	38
\mathbf{SO}	0	0	0	0	0	124

Table 2: Zero R classification results

	Correct	Incorrect
Instances	124	280
Percentage	30.6931~%	69.3069 %

Zero R uses zero rules as the name suggests, becouse of this it chooses 1 group that is most likely to be classified as seen in table 1.

One R

Table 3: One R Confusion matrix

	SW	W	T	R	Р	SO
SW	69	11	1	10	8	9
\mathbf{W}	35	2	4	4	10	8
${f T}$	4	1	4	2	8	4
${f R}$	29	5	1	6	6	1
P	5	0	7	0	16	10
\mathbf{SO}	2	0	3	0	10	109

Table 4: One R classification results

	Correct	Incorrect
Instances	208	196
Percentage	51.4851~%	48.5149 %

Chosen classifier: Random.Forest

Table 5: Random.Forest Confusion matrix as chosen classifier

	SW	W	Т	R	Р	SO
\mathbf{SW}	87	11	0	4	0	6
${f W}$	17	34	0	2	4	6
${f T}$	2	0	12	2	4	3
${f R}$	7	1	0	37	3	0
P	0	0	1	1	30	6
\mathbf{SO}	0	0	1	1	1	121

Table 6: Random.Forest as chosen classifier

	Correct	Incorrect
Instances Percentage	321 79.4554 %	83 20.5446 %
Percentage	19.4334 70	20.3440 70