

APPLICATION

pavo: an R package for the perceptual analysis, visualization and organization of spectral colour data

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Summary

1. Animal colouration has been a powerful model for the study of a diverse range of topics in biology. Recent technical, methodological and empirical advances have led to a dramatic increase in the use of spectrophotometry to quantify reflectance properties of biological material and of models to determine how these colours are perceived by the animals themselves, providing important insights to ecological and evolutionary aspects of animal visual communication.
2. Despite this growing interest, a unified framework for analyzing spectral data has not been available. We introduce **pavo**, an R package that facilitates the organization, visualization, and analysis of spectral data in a cohesive framework. **pavo** is highly flexible: it allows the user to (a) organize and manipulate data from a variety of sources, (b) visualize data with easy-to-use tools, and (c) analyze data using visual system modeling for a broad range of taxa.
3. In this paper, we present a summary of the functions implemented in **pavo**, suggest a workflow to explore spectral data, and model colours using user-defined visual systems. We also present an exact solution for the calculation of colour volume overlap in colourspace, thus expanding previously published methodologies.
4. As an example of **pavo**'s capabilities, we compare the colour patterns of three African Glossy Starling species, two of which have diverged very recently. Using spectral data and avian visual models, we show that colour disparity between the recently diverged species is as great as that between them and the more distantly related species. The flexibility and streamlined workflow of **pavo** is demonstrated through use of different visual models and several plotting capabilities.
5. **pavo** provides a unique environment capable of addressing complex sensory ecology questions that previously required the use of multiple, sometimes restricted, software and programs.

Key-words: animal communication, colourspace, just noticeable differences, receptor noise, sensory ecology, spectrophotometry, visual model

10 Introduction

11 The role of colouration and colour vision in animal communication has been a fundamental question in evolutionary
12 biology for centuries (Darwin, 1859, 1896; Poulton, 1890; Bennett & Théry, 2007). Visual communication has provided
13 a unique opportunity to investigate aspects of natural (Chittka & Menzel, 1992) and sexual selection (Hill, 2002), and
14 how these interact (Kemp *et al.*, 2009). It is also an ideal system for truly integrative biological research, spanning from
15 the optical processes generating colour (Shawkey *et al.*, 2009), hormonal and genetic mechanisms regulating phenotype
16 (Muller & Eens, 2009), physiological processes involved in perceiving the signal (Hart, 2001), and its adaptive and
17 evolutionary patterns (Badyaev & Hill, 2003; Darst *et al.*, 2006).

18 However, “colour” refers to a sensory experience, not an objective quantity, and the realization that animals can
19 vary quite considerably in their visual systems and how they process this information prompted two important
20 methodological advances. First, it highlighted the need for an objective quantification of the colour reflected by surfaces,
21 as a first approximation of a “receiver-independent” measure of an organism or object’s colour (Endler, 1993; Eaton
22 & Lanyon, 2003; Bennett & Théry, 2007). Over the last 20 years, the falling cost and rising popularity of portable
23 spectrophotometers have made objective quantification of the spectral properties of animal and plant integuments
24 commonplace (Endler, 1990; Eaton & Lanyon, 2003; Andersson & Prager, 2006). Second, advances in the understanding
25 of perception and processing of colour have allowed analysis of reflectance data using visual models that estimate how
26 animals see and differentiate these colours (Goldsmith, 1990; Tovee, 1995; Vorobyev & Osorio, 1998; Vorobyev *et al.*,
27 1998).

28 Because of these advances, a cohesive framework for working with and analyzing colour from reflectance data is
29 needed. Output file types from spectrophotometer manufacturers are not standardized, and software programs for
30 analyses are limited either in the methods they implement, the types of data they import and process, or the platforms
31 in which they are available. Moreover, many are proprietary and/or closed, making customization impossible (for a
32 review of available software, see Montgomerie, 2006). Colour data often require additional conversion and export into
33 statistical software for analyses, restricting protocol standardization across labs, and batch processing automation of
34 workflows. Combined with published descriptions that are often poorly detailed or inconsistent, the current procedures
35 and implementations severely hinder the reproducibility of research and the reimplementations of methods.

36 We introduce **pavo** – a package for R (R Core Team, 2012) that addresses the aforementioned problems by providing
37 a flexible, yet cohesive, environment in which researchers can organize, analyze and visualize colour data generated
38 by spectrophotometry. R is also open source and multi- platform, and is rapidly becoming the working language for
39 scientific programming and data analysis, particularly in ecology and evolution (e.g. Paradis *et al.*, 2004; Bolker, 2008).
40 **pavo** incorporates R’s flexibility by using object classes that can seamlessly interpret each other, providing functions
41 that can be used to intuitively import, explore, process and analyze spectral colour data under a variety of user- defined
42 models. We propose that combining these procedures under a coherent framework not only streamlines workflow, but
43 also allows data to be explored and manipulated in new ways that can be used to visualize patterns, obtain information
44 and develop and test hypotheses (Figure 1).

45

Figure 1 about here

46 The pavo package

47 The stable release of **pavo** is available from CRAN (<http://cran.r-project.org/>) for direct installation from R,
48 and the development version is available from github (<https://github.com/rmaia/pavo>). **pavo** was developed with

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three main workflow stages in mind (Figure 1): **organization** of spectral data by inputting raw files and processing their spectral content; **visualization** of the output, including exploratory capacities to identify further required manipulations and previously unconsidered patterns; and **analysis** of data from the spectral shape of reflectance curves or by incorporating receiver psychophysiology in visual models (Figure 1). As noted by Bennett & Théry (2007) and others (Andersson & Prager, 2006; Montgomerie, 2006), though spectral data have become commonplace in studies of animal colouration, it is easy to obtain poor-quality or inaccurate data. Therefore a workflow for spectral colour data analysis has to go beyond a “plug and chug” implementation, requiring thorough exploratory investigation. With this in mind, **pavo** takes advantage of R’s object-oriented programming environment, as outlined below.

ORGANIZATION

Spectral data are stored in **pavo** and recognized for its functions by use of a new object class, “**rspec**”, which inherits methods from **data.frame**. Objects of class **rspec** are characterized by having individual reflectance spectra as columns of the data frame, with a first column containing the wavelength identification data. Raw spectral data can be imported using the function **getspec**, which currently supports data from a variety of spectrophotometer and microspectrophotometer models (Ocean Optics OOIBase and SpectraSuite files, Avantes AvaSpec, and CRAIC). In addition, previously compiled data frames can be imported into R and converted to **rspec** objects using the **as.rspec** function.

The use of dedicated R object classes allows generic functions like **plot** and **summary** to identify the object as a particular type of data frame and interpret it accordingly (see below). The class “**vismodel**” is used to interpret spectral data that has been processed through one of the visual models implemented, and also stores information on how it was generated (e.g. the visual phenotype, background and illuminant used, and any transformations applied; see below). Additionally, the “**tcs**” class refers to tetrahedral colourspace models (Endler & Mielke, 2005; Stoddard & Prum, 2008), and the **summary** function can be used to extract summary variables, like the colour volume or hue span (Stoddard & Prum, 2008) for groups and subsets of points.

It is common when collecting spectral data to take multiple measurements from the same sample, averaging these to avoid sampling error (Quesada & Senar, 2006). **pavo** provides the **agggspec** function for this purpose, as well as the **procspec** function for noise removal via smoothing, and transformations to standardize and clean spectral data.

VISUALIZATION

With **pavo** installed and loaded, the **plot** function recognizes **rspec** objects and plots them accordingly – interpreting the first column as wavelengths (usually using it as the x axis) and the remaining columns as reflectance values (y values) for individual spectra. Several plotting options for multiple spectra are implemented (Figure 1). In addition, the **aggplot** function provides plotting capabilities for among-spectra summary statistics plotting (Figure 1), with a similar syntax to **agggspec**. **pavo** also offers exploratory plotting capabilities that can be combined with data processing and formatting, such as **explorespec** (for visualizing groups of spectra) and **smoothplot** (for choosing smoothening parameters; Figure 1).

Finally, **pavo** offers plotting capabilities for the avian tetrachromatic colourspace model (Stoddard & Prum, 2008; Endler & Mielke, 2005) through the **tcsplot** and **projplot** functions (Figure 1, see below).

ANALYSIS

The **summary** function can be applied to **rspec** objects to produce several tristimulus (hue, saturation and brightness) colour variables extracted from the spectral shape of the reflectance curve. This function calculates variables previously described in the literature (summarized in Montgomerie, 2006) for the spectra contained in the **rspec** input object (description and formulae for the variables can be found in the help file, by typing **?summary.rspec**, and in the package Vignette). Additionally, the function **peakshape** provides descriptors of spectral peaks, such as the wavelength of maximum reflectance and the full width at half maximum, and can be fine-tuned to extract information from specific

areas of the curve. This implementation can be useful when the spectral curve has multiple peaks or a complex shape (e.g. describing the UV peak of carotenoid curves, Figure 1).

pavo also allows the easy production of models that incorporate the visual system of the receiver through the **vismodel** function. Models can be calculated incorporating the visual phenotype (cone absorbance), background colour, and ambient illuminant (Vorobyev & Osorio, 1998). Several avian receptor phenotypes (Hart, 2001; Endler & Mielke, 2005) are implemented as options, but user-defined receptor data from any taxon can be used as model input. Further, the **sensmodel** function implements the calculation of cone absorbance curves based on peak sensitivity information (available from the literature, e.g. Hart, 2001), and can also include oil droplet and ocular transmission information in the calculations (Govardovskii *et al.*, 2000; Hart & Vorobyev, 2005).

Visual models can be calculated in terms of absolute photon catches, in which case the receptor noise model can be used to infer contrast between colours (implemented in the function **colordist** Vorobyev & Osorio, 1998) or in relative cone stimulation, in which case the model reduces to a colourspace model represented in $n - 1$ dimensions (where n is the number of different receptors involved in colour vision Goldsmith, 1990; Endler & Mielke, 2005; Stoddard & Prum, 2008). Absolute or relative cone stimulation can be selected by the logical argument **relative** from the **vismodel** function. In the case of the avian tetrahedral colourspace, several additional variables can be calculated based on spherical coordinates which represent the hue based on the angles and saturation based on the distance from the achromatic center (see Stoddard & Prum, 2008) by calling the **tcs** function. This function generates an object of class **tcs**; a **summary** call from a **tcs** object will return summary statistics described in Stoddard & Prum (2008) for sets of points (see below).

pavo also builds upon previously described visual model methods. For example, Stoddard & Stevens (2011) present the useful technique of calculating the overlap between the volumes defined by two sets of points in colourspace. They used this metric to quantify mimicry (Stoddard & Stevens, 2011; Stoddard, 2012), such that a greater volume overlap would indicate greater overall colour similarity. Given the complexity of calculating the intersection of three-dimensional convex hulls, Stoddard & Stevens (2011) used a Monte Carlo approach to estimate the degree of volume overlap. **pavo**, instead, provides the exact solution for the calculation of the intersection of colour volumes, using a method originally implemented to calculate the overlap between multidimensional niches (Villéger *et al.*, 2011) through the computational geometry capabilities available from the **rcdd** package (Geyer *et al.*, 2012) (For performance and precision comparison, see Supplemental Information)

Example: colourspace divergence in glossy starlings (*Sturnidae*)

Here we exemplify some of **pavo**'s capabilities and workflow by comparing the colour patterns of a monophyletic clade of three glossy starling species. Glossy starlings from the African clade have bright and diverse iridescent colours likely used in sexual selection and social competition (Rubenstein & Lovette, 2009). The lesser blue-eared starling (*Lamprolornis chloropterus*) and the southern blue-eared starling (*L. elisabeth*) form a recently diverged clade that some (including the International Ornithological Committee; Gill & Donsker, 2012) have considered full species, whereas others have classified them as subspecies of *L. chloropterus* (Craig *et al.*, 1998; Lovette & Rubenstein, 2007). We compared the extent to which these two species have diverged in their colours and compared each to their most recently diverged sister species, the sharp-tailed starling (*L. acuticaudus*).

STEP 1: ORGANIZATION AND PROCESSING

The data consist of reflectance spectra (in Avantes ".ttt" output format) taken from museum specimens of the three species. We measured three reflectance spectra from 11 plumage patches (see Figure 2 for list) of four males per species. We loaded, averaged and removed the electrical noise arising from the spectrometer (using local polynomial regression fitting, or loess) from these 396 raw spectral data files using the following annotated lines of code in **pavo**:

```

> specs <- getspec(when = '/Desktop/glossystarlings', ext = 'tiff', lim = c(300, 700)) #get raw data
> specs <- aggspec(specs, by = 3, FUN = mean) #average by groups of 3 spectral curves
> specs <- procspec(specs, opt='smooth') #remove electrical noise

```

134 STEP 2: VISUALIZATION

135 Next, we plotted spectra contained in this `rspec` object. We used `aggplot` to visualize the mean reflectance curves for
 136 each body part from each species. Below we show the code for one of these plots, as well as the results for the “belly”
 137 and “wing” body patches: (Figure 2):

```

> specs.belly <- specs[,1:13] # subset wavelength column and the 12 spectra from the belly patch
> spp <- substr(names(specs.belly),1,4)[-1] # 4 first characters from column names (species labels)
> aggplot(specs.belly, by = spp) # average spectral data by species

```

138 STEP 3: ANALYSIS

139 To understand how these colours are perceived by birds, we used the `vismodel` function to compare colours taking into
 140 account avian visual sensitivities (sensory phenotype). `pavo` incorporates available data on the retinal sensitivities of
 141 numerous taxa, including the European starling (*Sturnus vulgaris*, `visual='star'`), which we incorporated here (Hart
 142 *et al.*, 1998). We used both the `relative=FALSE` and `relative=TRUE` options of the `vismodel` function to measure
 143 colour attributes using the receptor noise and colourspace models, respectively.

144 Colour distances

145 First, `relative=FALSE` was used to calculate raw photon catch values for the four avian photoreceptor classes (`{usml}`),
 146 suitable for calculating chromatic distances (ΔS , Vorobyev *et al.*, 1998). We used relative cone abundances (arguments
 147 `n1`, `n2`, `n3`, and `n4`) for the European starling (Hart *et al.*, 1998), which can be easily passed as arguments to the
 148 `coldist` function. The `v` argument was set to 0.1 to give a $\{l\}$ cone Weber fraction of approximately 0.05 (Vorobyev
 149 & Osorio, 1998; Vorobyev *et al.*, 1998). The results of `coldist` give colour distances (chromatic, ΔS , Figure 2; and
 150 achromatic, ΔL , not shown) between all possible combinations of plumage patches in the `tcs` object, and can be further
 151 subset to only analyse comparisons of interest (e.g., colour difference between homologous patches of two species). For
 152 example:

```

> vm.star1 <- vismodel(specs, visual = 'star', relative = FALSE)
> delta.S <- coldist(vm.star1, qcatch = 'fi', n1 = 1, n2 = 1.38, n3 = 3.34, n4 = 3.46, v = 0.1)
> # subset only comparisons between wing spectra
> head(delta.S[intersect(grep('wing', delta.S$patch1), grep('wing', delta.S$patch2)),])

```

	patch1	patch2	tetra.dS	dL
1031	LAEL.668721.MAL.wing	LAEL.668723.MAL.wing	4.1937305	5.427964
1042	LAEL.668721.MAL.wing	LAEL.668724.MAL.wing	0.6092773	2.042986
1053	LAEL.668721.MAL.wing	LAEL.668727.MAL.wing	1.3653852	1.085405
1064	LAEL.668721.MAL.wing	LACL.162946.MAL.wing	2.5464770	6.832917
1075	LAEL.668721.MAL.wing	LACL.668619.MAL.wing	0.6138824	2.187837
1086	LAEL.668721.MAL.wing	LACL.668623.MAL.wing	1.7393595	2.252061

153 The calculation of both chromatic and achromatic distances of these 8646 pairs of colors took about 0.9 seconds
 154 in a MacBook Pro running R version 2.15.2. We can see from Figure 2 that the recently divergent *L. chloropterus*
 155 and *L. elisabeth* have accumulated similar levels of color disparity among them as they have to their sister species, *L.*
 156 *acuticaudus*. Considering a value of 1 as a threshold for Just Noticeable Differences (JNDs, Figure 2A dashed lines)
 157 nearly all plumage patch comparisons yield colors that are discernible to the starling visual system, both within the

158 *L. chloropterus*-*L. elisabeth* subclade as well as compared to the *L. acuticaudus* outgroup. Given the high saturation
 159 of iridescent colors, small differences in the spectral shape can yield noticeably different colors. This becomes clear
 160 when comparing the results for the most and least contrasting body patches (“belly” and “wing”, respectively) to their
 161 spectral curves (see Figure 2B,C). Even groups of spectra that considerably overlap in their average reflectance, as
 162 in the case of the “wing” patch (Figure 2B), the pairwise comparisons of individual spectra yield values above the
 163 threshold of JND=1 (Figure 2A). Even within-species comparisons yield discernible colors for the “wing” patch, as
 164 evidenced by the output from `coldist` (above).

165 Figure 2 about here

166 *Avian tetrahedral colourspace*

167 Finally, to produce a tetrahedral colourspace plot, we used `relative=TRUE` in the `vismodel` input to indicate that
 168 `{usml}` values should be scaled to sum of unity for each patch. These results indicate that the expansion of occupied
 169 colourspace for both subspecies is largely due to divergence in the colour of the “belly” patch (see circles labeled ‘bel’
 170 at edges of red and blue polyhedrons in Figure 3a). Interestingly, the “cheek” plumage patch of *L. elisabeth* is more
 171 similar to *L. acuticaudus* than *L. chloropterus*, to which it is more closely related (circles labeled ‘ch’ in Figure 3A).

172 We also used the `voloverlap` function to calculate the volumes occupied by each species’ plumage patches, as well
 173 as their overlaps. As an example, the code below shows the result for the overlap between the two recently diverged
 174 species (see Figure 3B):

```
> vm.star2 <- vismodel(specs, visual = 'star', relative = TRUE)
> tcs.star <- tcs(vm.star2)
> #subset points from the two species
> tcs.lael <- tcs[grep('LAEL',row.names(tcs.star))]
> tcs.lacl <- tcs[grep('LACL',row.names(tcs.star))]
> voloverlap(tcs.lael, tcs.lacl, plot = TRUE)
```

	vol1	vol2	overlapvol	vsmallest	vboth
1	1.916209e-05	4.05405e-05	4.673727e-06	0.2439049	0.08493229

175 Figure 3 about here

176 The `plot=TRUE` option provides a useful graphical representation of the overlap in colourspace (Figure 3b). Again, the
 177 calculated volume in `pavo` represents the exact solution to the geometric problem (i.e. not an approximation obtained
 178 through Monte Carlo simulation, Stoddard & Stevens, 2011).

179 **Citation of methods implemented in pavo**

180 Most of the methods implemented in `pavo` have been thoroughly described in their original publications, to which users
 181 should refer for details and interpretation. For summary tristimulus variables and spectral shape analysis of reflectance
 182 curves, see Montgomerie (2006). For visual models based on photon catches and receptor noise, see Vorobyev & Osorio
 183 (1998) and Vorobyev *et al.* (1998). For visual phenotype sensitivity curve estimation, see Govardovskii *et al.* (2000)
 184 and Hart & Vorobyev (2005). For tetrahedral colourspace models, see Endler & Mielke (2005) and Stoddard & Prum
 185 (2008), and for colour volume overlap see Stoddard & Stevens (2011) and Stoddard (2012). Users of the functions that
 186 apply these methods must cite the original sources along with `pavo`.

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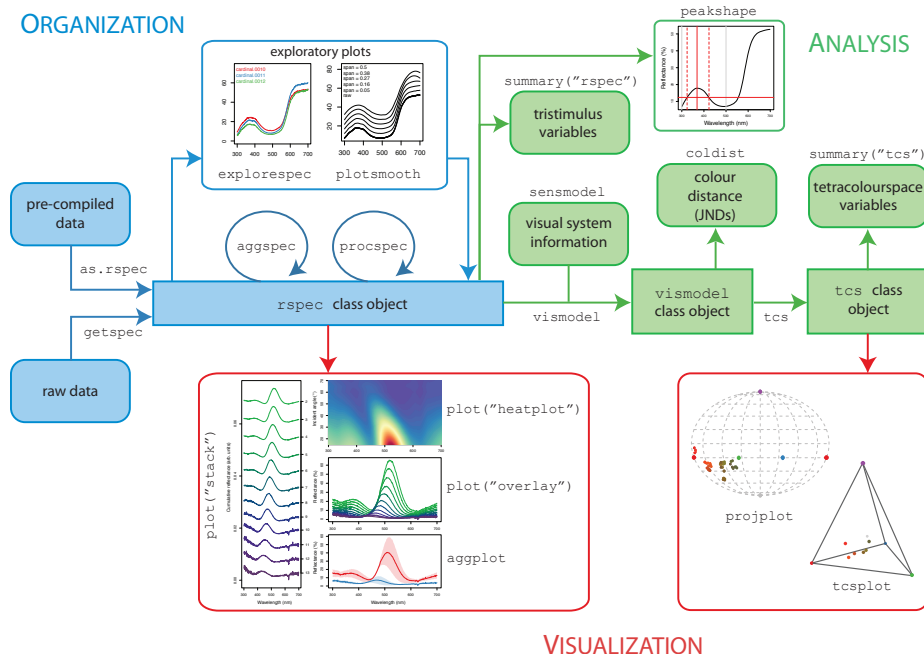


Fig. 1. Example pavo workflow, highlighting its main functions and plotting capabilities

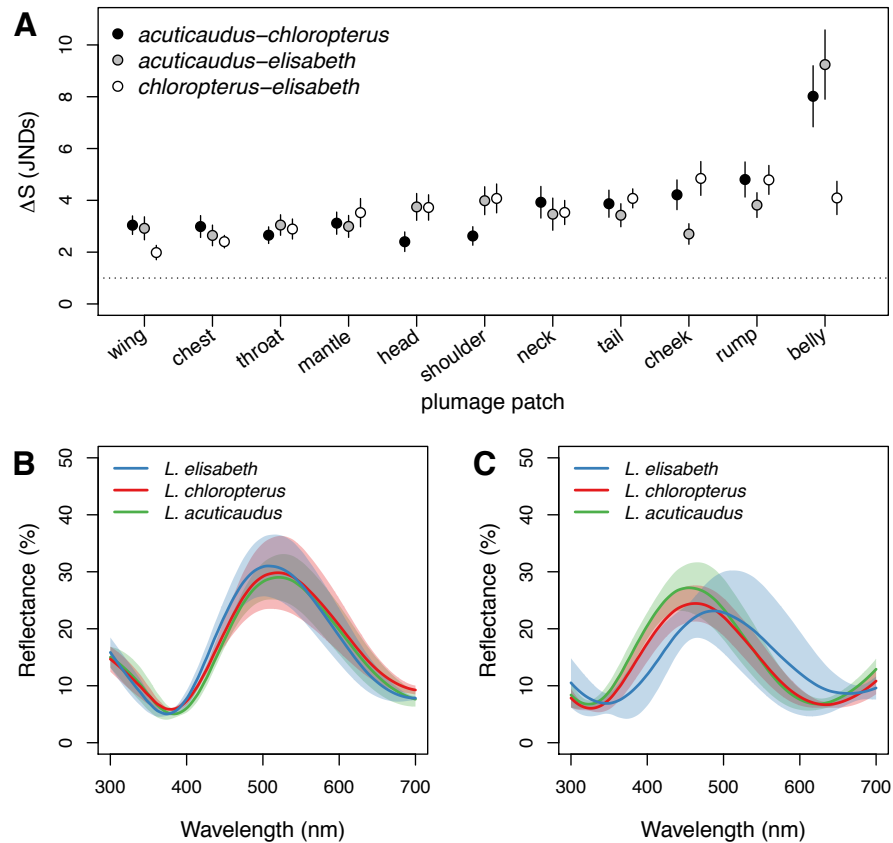


Fig. 2. **A:** Dotplot showing color distances (in units of just noticeable differences, JNDs) by patch (y-axis) for three pairs of African starling species (Sturnidae). Dotted vertical lines indicate JND=1, above which the pair of colour patches is considered to be distinguishable by birds. Points indicate chromatic distances between different pairs of species: *Lamprolornis chloropterus* and *L. elisabeth* (open circles); *L. acuticaudus* and *L. elisabeth* (grey circles); and *L. acuticaudus* and *L. chloropterus* (black circles). Error bars show \pm one standard error. **B,C:** Plots of mean smoothed spectra for different body patches (B: wing, C: belly). Line colors indicate species (green: *L. acuticaudus*, red: *L. chloropterus*, blue: *L. elisabeth*) and shaded areas indicate the standard deviation of the spectral data.

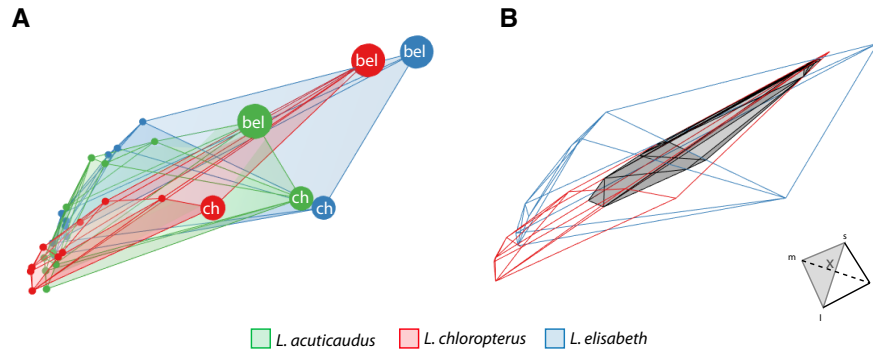


Fig. 3. **A:** Tetrahedral colourspace plot showing minimum convex polygons (convex hulls) containing reflectance spectra for plumage patches from *Lamprolornis acuticaudus* (green), *L. chloropterus* (blue) and *L. elisabeth* (red). Labeled points show representative plumage patches belly (bel) and cheek (ch). **B:** Colourspace occupied by *L. chloropterus* (blue) and *L. elisabeth* (red). Convex hulls volume overlap (similar colourspace occupied) is shown in grey. For reference, tetrahedron in lower right corner shows $\{usml\}$ colour vertices and location of points in colourspace as indicated by an 'x'.