

cmstatr: An R Package for Statistical Analysis of Composite Material Data

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Summary

A number of statistical techniques are commonly used when analyzing strength data for composite materials used in aerospace applications, such as carbon fiber and fiberglass. Currently, many users use MS Excel spreadsheets for performing this analysis. **cmstatr** is an R package that implements the statistical analysis techniques commonly used for composite material strength data.

The design standards for civil aviation require that the probability of structural failure due to material variability is minimized. To do so, the designer must select Design Values for each material and compare those to the stresses experienced by those materials. These Design Values are selected so that, with 95% confidence, the Design Value is either the 99% or 90% lower confidence bound of the material strength, depending on the type of structure. These one-sided tolerance bounds are referred to as A-Basis and B-Basis values, respectively. Computing these A- and B-Basis values is the main problem that **cmstatr** addresses.

A set of statistical methods are described in a publication called the Composites Materials Handbook, or CMH-17-1G (CMH-17 2012). The use of these methods is widely accepted by industry and civil aviation regulators. The methods described in CMH-17-1G are implemented in **cmstatr**.

The MS Excel spreadsheets typically used, such as **STAT-17** (Materials Sciences Corporation 2008), **ASAP** (Raju and Tomblin 2008) and **CMH17-STATS** (Wichita State University and Keshavanarayana 2012), use password-protected VBA macros to perform the computations. As such, the code cannot be audited by the user. **cmstatr** aims to address this by providing open-source code for performing these computations.

Statement of Need

The purpose of **cmstatr** is to:

- Provide a consistent user interface for computing A- and B-Basis values and performing the related diagnostic tests in the R programming environment
- Allow auditing of the code used to compute A- and B-Basis values
- Enable users to automate computation workflows or to perform simulation studies

Implementation Goals

cmstatr aims give a consistent interface for the user. Most functions are written to work with the **tidyverse** (Wickham et al. 2019) and most functions have similar argument lists. The intent is to make the package easy to learn and use.

The implementation of **cmstatr** also aims to avoid the use of look-up tables and minimize the use of approximations. While this decision leads to increased computation time, the typically small data sets (tens to hundreds of observations) associated with composite material test data, and the speed of modern computers make this practical for interactive programming.

Example Usage

Normally, to use `cmstatr` the user will load `cmstatr` itself as well as the `tidyverse` package.

```
library(cmstatr)
library(tidyverse)
```

`cmstatr` contains some example data sets, which can be used to demonstrate the features of the package. One of those data sets — `carbon.fabric.2` — will be used in the following example. This data set contains results from several mechanical tests of a typical composite material. and contains the typical measurements obtained from a test lab. In the following examples, results from the “warp tension” (WT) test will be used. Part of this data set is shown below.

```
carbon.fabric.2 %>%
  filter(test == "WT") %>%
  head(10)
```

##	test	condition	batch	thickness	nplies	strength	modulus	failure_mode
## 1	WT	CTD	A	0.112	14	142.817	9.285	LAT
## 2	WT	CTD	A	0.113	14	135.901	9.133	LAT
## 3	WT	CTD	A	0.113	14	132.511	9.253	LAT
## 4	WT	CTD	A	0.112	14	135.586	9.150	LAB
## 5	WT	CTD	A	0.113	14	125.145	9.270	LAB
## 6	WT	CTD	A	0.113	14	135.203	9.189	LGM
## 7	WT	CTD	A	0.113	14	128.547	9.088	LAB
## 8	WT	CTD	B	0.113	14	127.709	9.199	LGM
## 9	WT	CTD	B	0.113	14	127.074	9.058	LGM
## 10	WT	CTD	B	0.114	14	126.879	9.306	LGM

One common task is to calculate B-Basis values. Depending on the distribution of the data, this can be done using one of several functions. Assuming that the data from the warp tension (WT) elevated-temperature wet (ETW) strength follows a normal distribution, this can be done as follows:

```
carbon.fabric.2 %>%
  filter(test == "WT") %>%
  filter(condition == "ETW") %>%
  basis_normal(strength, batch)
```

```
## Warning: `anderson_darling_normal` failed: Anderson-Darling test rejects
## hypothesis that data is drawn from a normal distribution
##
## Call:
## basis_normal(data = ., x = strength, batch = batch)
##
## Distribution: Normal ( n = 18 )
## The following diagnostic tests failed:
## `anderson_darling_normal`
## B-Basis: ( p = 0.9 , conf = 0.95 )
## 122.9315
```

All of the various basis functions perform diagnostic tests. If any of the diagnostic tests fail, a warning is emitted and the test failure is also recorded in the returned object (and shown in that object’s `print` method). In the example above, the output shows that the Anderson-Darling test for normality (Lawless 1982) rejects the hypothesis that the data is drawn from a normal distribution. The single-point basis functions perform the following tests: the maximum normed residual test for outliers within a batch (CMH-17 2012), the Anderson-Darling k-Sample test to check if batches are drawn from the same (unspecified) distribution (Scholz

and Stephens 1987), the maximum normed residual test for outliers within the data, and the Anderson-Darling test for a particular distribution (Lawless 1982).

Two non-parametric basis calculations, based on (Guenther 1969) and (Vangel 1994) are also implemented. These functions perform the same diagnostic tests, but skip the Anderson-Darling test for a particular distribution.

The diagnostic test can be run directly using `cmstatr` as well. For example, the failed diagnostic test above can be run as follows:

```
carbon.fabric.2 %>%
  filter(test == "WT") %>%
  filter(condition == "ETW") %>%
  anderson_darling_normal(strength)
```

```
##
## Call:
## anderson_darling_normal(data = ., x = strength)
##
## Distribution: Normal ( n = 18 )
## Test statistic: A = 0.9381665
## Significance: 0.01103075 (assuming unknown parameters)
## Conclusion: Sample is not drawn from a Normal distribution (alpha = 0.05 )
```

If it is decided that the failure of the diagnostic test is acceptable, the test can be overridden to avoid a warning from being emitted by the basis function:

```
carbon.fabric.2 %>%
  filter(test == "WT") %>%
  filter(condition == "ETW") %>%
  basis_normal(strength, batch, override = c("anderson_darling_normal"))
```

```
##
## Call:
## basis_normal(data = ., x = strength, batch = batch, override = c("anderson_darling_normal"))
##
## Distribution: Normal ( n = 18 )
## The following diagnostic tests were overridden:
## `anderson_darling_normal`
## B-Basis: ( p = 0.9 , conf = 0.95 )
## 122.9315
```

`cmstatr` provides functions for calculating basis values using data pooled across environments, as recommended by (CMH-17 2012). These functions use the variance observed in different environmental conditions in the computation, but acknowledge the different mean values under each environmental condition.

Another common statistical technique is to determine if a sample is drawn from a particular population. This is often used to determine if data from a second manufacturing site supports the basis values determined from test data generated at the first manufacturing source. The statistical test often recommended for this application considers both the mean and minimum individual value (Vangel 2002). This statistical test has higher power than some other tests that could be used. `cmstatr` also provides functions for computing limits based on this test. For example:

```
carbon.fabric.2 %>%
  filter(test == "WT") %>%
  filter(condition == "RTD") %>%
  equiv_mean_extremum(strength, n_sample = 8, alpha = 0.05)
```

```
##
```

```
## Call:
## equiv_mean_extremum(df_qual = ., data_qual = strength, n_sample = 8,
##   alpha = 0.05)
##
## For alpha = 0.05 and n = 8
## ( k1 = 2.700045 and k2 = 0.6789966 )
##           Min Individual      Sample Mean
## Thresholds:      121.4921      135.0655
```

Validation and Comparison With Existing Tools

Where possible, `cmstatr` has been verified against the examples given in the articles in which the statistical methods were published. Unit tests have been written so that this verification is re-checked routinely to prevent unintended regressions. Agreement between `cmstatr` and the examples in the original articles is within the expected numeric accuracy.

`cmstatr` has also been verified against existing software, such as STAT-17 (Materials Sciences Corporation 2008), ASAP (Raju and Tomblin 2008) and CMH17-STATS (Wichita State University and Keshavanarayana 2012) using several example data sets. Agreement between `cmstatr` and this other software is generally good, but some results differ slightly, likely due to approximations used in the software. Comparison between `cmstatr` and this other software is performed within various unit tests to guard against future regressions.

The tests are automatically run each time a change is made to the code of `cmstatr` using a continuous integration service. Additionally, CRAN runs R CMD check on each package routinely.

Reproducibility

It is envisioned that many users of `cmstatr` will use it within an R Notebook or a Jupyter Notebook. It is further envisioned that this notebook will be directly converted into the statistical analysis report. If this is done, the reader of the statistical report will be able to verify all of the detailed steps used in the statistical analysis.

References

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