

Normalization of Power Spectral Density

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Abstract

Normalizations come in many forms, but here we outline the normalization used by `rlpSpec`: power spectral density.

Contents

1	Background	1
1.1	<code>stats::spectrum</code>	1
1.2	<code>multitaper::spec.mtm</code>	3
1.3	<code>SDF::sapa</code>	3

1 Background

There can often be confusion about the different quantities used in spectral analysis ¹, partly due to myriad nomenclature within the incredibly vast literature on the subject. Regarding nomenclature: Phrases including “amplitude spectrum”, “energy spectral density”, “power”, “power spectra”, and even “spectra” all mean *something*, but are rarely equivalent.

Let us, for the sake of brevity, assume we have a stationary signal $f(x)$ having Fourier Transform (FT) $F(s)$, which is complex. The of this signal is simply the real component of the FT, $\arg\{F\}$, which we’ll denote as $^{(A)}S$. The corresponding is thus $\text{mod}F$, or $^{(\phi)}S$. These quantities are, however, meaningless.

The latter assumes that the spectrum has the normalization used in power – which is that the Nyquist is assumed to be 1/0, always (power doesn’t ask for a sample interval). But `psd` assumes that the interval is 1 (unless you specify otherwise) so that the Nyquist is 0.5. So say you have a white noise with variance 1: power will return a flat spectrum with level 0 db (=1) but `psd` will return a level of 2 (=3 db), both so that the level times the Nyquist will be 1. But `logsmoo`, told that the interval is 1, will multiply the spectrum by 2 to covert from $lc^2/\text{Nyquist}$ to lc^2/Hz : giving 3 dB for the spectrum from power, but 6 dB (=4) for the spectrum from `psd`.

First load the package into the namespace:

```
> library(rlpSpec)
```

1.1 `stats::spectrum`

Included in the core distribution of R is `stats::spectrum`, which accesses `stats::spec.ar` or `stats::spec.pgram` for either parametric and non-parametric estimation, respectively. The user can optionally apply a single cosine taper, and/or a smoothing kernel. Our method is non-parametric; hence, we will compare to the latter.

¹ This post to R-help very eloquently describes the problem, and provides some guidance: <http://r.789695.n4.nabble.com/Re-How-do-I-normalize-a-PSD-td792902.html>

```
> spec.pgram(X, pad=1, taper=0.2, detrend=FALSE, demean=FALSE, plot=FALSE)
```

However, the logical arguments `detrend` and `demean` to `psdcore` are passed to `spec.pgram`; they are, by default, both `TRUE`.

As a matter of bookkeeping, we must deal with the working environment accessed by `rlpSpec` functions. Specifically, we should ensure `psdcore` does not access any inappropriate information by setting `refresh=TRUE`. We can then re-calculate the multitaper PSD and the raw periodogram with `plotpsd=TRUE`. The results are shown in Figure 1.1.

```
> data(magsat)
> psdcore(magsat$clean, ntaper=10, refresh=TRUE, plotpsd=TRUE)
```

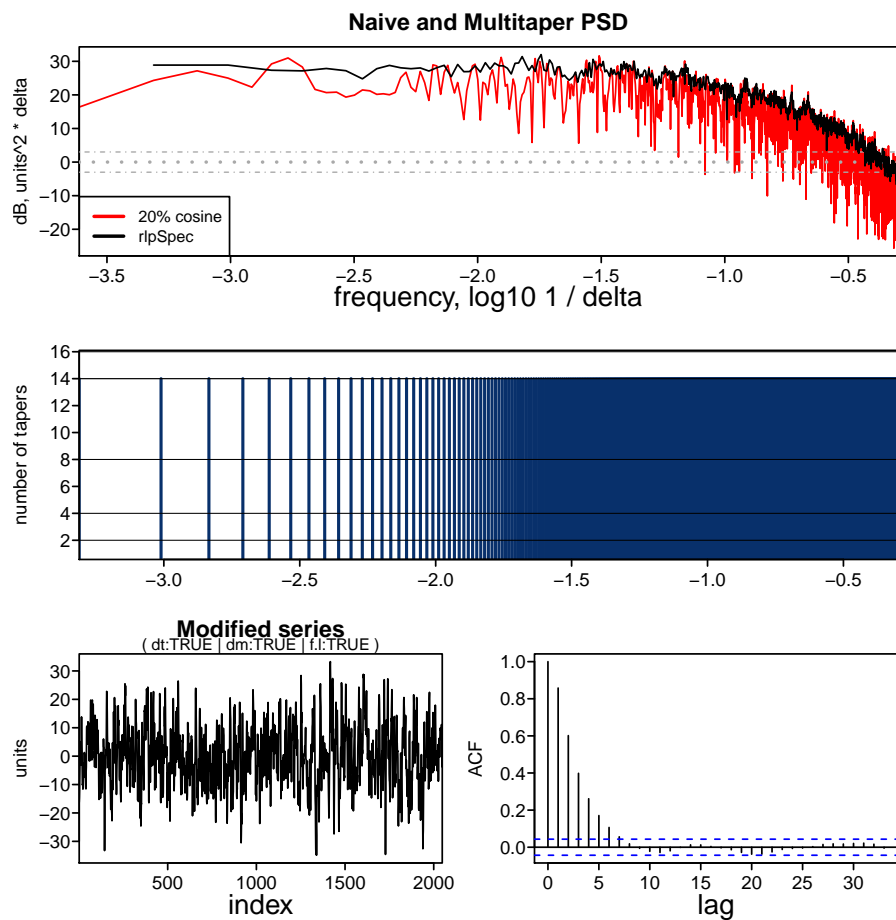


Figure 1: Top: Comparison between naïve and multitaper PSD estimators for the clean MAGSAT data. The frequency axis is in units of $\log_{10} \text{ km}^{-1}$, and power axis is in decibels. Bottom: The spatial series used to estimate the PSDs.

1.2 multitaper::spec.mtm

1.3 SDF::sapa

Index

amplitude spectrum, 1

phase spectrum, 1