Notes on Improving Spectral Estimates- Sandra Chapman (MPAGS: Time series analysis) Improving spectral estimates (Fourier Transforms)

Our 1...N DFT is a finite sized sample therefore suffers from finite sized effects.

Some <u>practical</u> methods for improving spectral estimates.

i) <u>Summed spectrogram</u>

Sub divide x_k into intervals 1...*p*, p + 1... 2*p*..., obtain DFT of each interval then sum. Spectral accuracy at the expense of frequency resolution (NB one can consider overlapping intervals- depends on application).

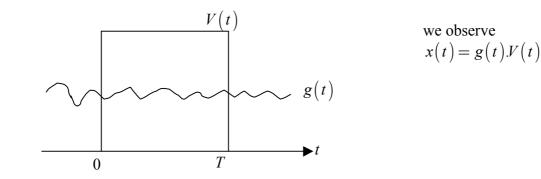
<u>Spectral variability</u> estimate – provided by $\langle S_m^2 \rangle$ over the interval for which there is one value of S_m for each sub-interval.

- ii) <u>Variables</u> chosen to reflect data, eg: power law spectra re sum over <u>log bins</u> in f.
- iii) Window in time domain to reduce spectral leakage (next)
- iv) Use a different transform (eg: wavelets).

Spectral estimates for a finite data interval (T)

We perform a DFT over N datapoints and $T = N\Delta t$. How does this affect the power spectrum?

Consider a process g(t) observed from $0 \le t \le T$.



recall the convolution theorem:

$$g(t) * h(t) = \int_{-\infty}^{\infty} g(\tau) h(t-\tau) d\tau$$

and

$$\int_{-\infty}^{\infty} g(t) * h(t) e^{-2\pi i f t} dt = G(f) . H(f)$$

also holds that

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$$G(f) * H(f) = \int_{-\infty}^{\infty} G(f') H(f - f') df'$$

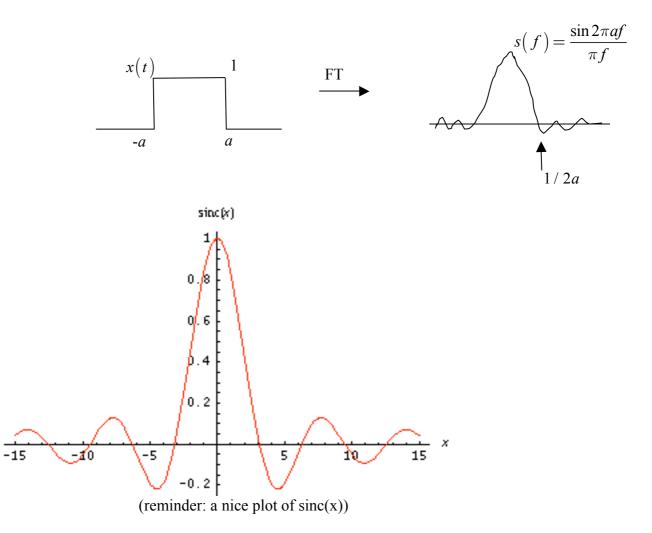
and

$$\int_{-\infty}^{\infty} G(f) * H(f) e^{2\pi i f t} df = g(t) h(t)$$

Then FT of x(t) sampled over interval T will be a <u>convolution – in f space</u> of the FT of g(t) with FT of V(t), this "spreads out" the local value of power spectrum.

We can see this as follows. Here V(t) = 1 for $0 \le t \le T$ and zero elsewhere.

We have already seen that:



[We can always "shift" the time base here by T/2, answer differs by phase $e^{-i\pi fT}$ and amplitude spectrum is the same.]

Then

$$V(f) = \frac{\sin(\pi fT)}{\pi f}$$

and our observed power spectrum will be

$$S(f) = V(f) * G(f)$$

where G(f) is the 'original' Power spectrum if sampled over arbitrarily long *T*- spectral "leakage" over $f \simeq \frac{1}{T}$.

No problem if

- 1) $T \to \infty$ since sinc(πfT) $\to \delta(f)$.
- 2) Signal is periodic fundamental period $T_s = 2T$.

<u>Otherwise</u> what to do – reduce leakage in f, by applying a time domain window to make signal "look like" case (2)

There are many of these! (see Matlab, in particular signal processing toolbox).

Most common:

a) Hanning
$$(\cos^2)$$
 window

$$U_{H}(t) = \cos^{2}\left(\frac{\pi t}{T}\right) - T / 2 \le t \le T / 2$$

= 0 otherwise

it's IFT is:

$$U_H(f) = T / 2 \cdot \frac{\operatorname{sinc}(\pi fT)}{1 - T^2 f^2}$$
$$= \frac{T}{2} \left[\operatorname{sinc}(\pi fT) + \frac{1}{2} \operatorname{sinc}(\pi fT - \pi) + \frac{1}{2} \operatorname{sinc}(\pi fT + \pi) \right]$$

this is "adding" power at the zeros of V(f), ie: at $f = \frac{1}{T}$ and suppressing at $f > \frac{1}{T}$.

We can then correct for the Hanning window as follows:

- consider a DFT – discrete $f_m = \frac{m}{N\Delta t} = \frac{m}{T}$

$$U_H(f_m) = \frac{T}{2} \left[\operatorname{sinc}(m\pi) + \frac{1}{2} \operatorname{sinc}((m-1)\pi) + \frac{1}{2} \operatorname{sinc}((m+1)\pi) \right]$$

Notes on Improving Spectral Estimates- Sandra Chapman (MPAGS: Time series analysis) or, with $Tf_m = m$ we have:

$$U_{H}(f_{m}) = \frac{T}{2} \left[\operatorname{sinc}(\pi f_{m}T) + \frac{1}{2}\operatorname{sinc}(\pi f_{m-1}T) + \frac{1}{2}\operatorname{sinc}(\pi f_{m+1}T) \right]$$

If we apply this to a white noise process S

- all the $S(f_m)$ are uncorrelated
- the power spectrum is uniform

then

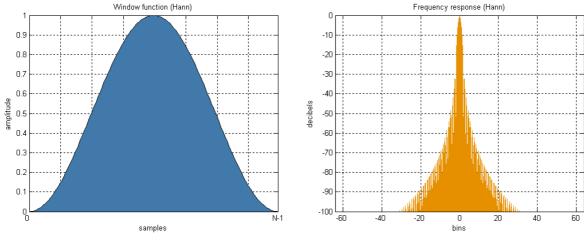
$$\left\langle S_m^2 \right\rangle = \left(\frac{1}{2}\right)^2 \left[1^2 + \left(\frac{1}{2}\right)^2 + \left(\frac{1}{2}\right)^2\right] \cdot G_m^2$$
$$= \frac{3}{8}G_m^2$$

then the "correct" $S_m \rightarrow \sqrt{\frac{8}{3}} \times S_m$ windowed.

Finally, the procedure is

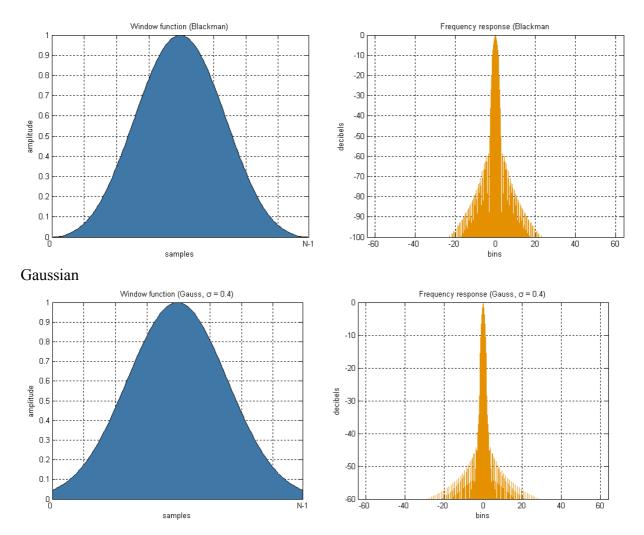
- 1) divide data into sub intervals T_0 in time domain; p of them $p = \frac{T}{T_0}$ these can be <u>overlapped (Welch method)</u>
- 2) Hanning window over T_0 .
- 3) correct by $\sqrt{\frac{8}{3}}$
- 4) average over the *p* spectra to obtain $\langle S_m \rangle_p$ - with uncertainty $\langle S_m^2 \rangle_p$.

<u>Other windows</u> (there is a nice list on Wikipedia-NB these are all in dB ie a logarithmic scale in power) :



Hann(shifted cosine)

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Flat top (sum of shifted cosines, period 2T, 4T,6T,8T)

