

PHY408

Lecture 11: Power Spectrum Estimation

March 29, 2023

Final report: Due April 26th (no extensions)

- 1 Maximum of 8 pages single spaced, with a 12 pt font size.
- 2 You can use extra pages for large figures; additional math references can also be on additional pages.
- 3 You can use any format for the references, and there is no lower or upper limit for references.
- 4 5% for overall formatting of the report.
- 5 20% for introduction and laying out the question you want to address clearly.
- 6 50% for analysis and results.
- 7 25% for final discussion of the results, discuss caveats, how this may be applicable more broadly, etc.
- 8 See the syllabus for links to possible sources of data.
- 9 We will use the University's plagiarism detection tool to provide to a similarity assessment of the final reports.

Cross-correlation

Cross-correlation of two signals can be used to measure the similarity of two signals under time translation. It is given by

$$C_{fg}(\tau) = \int_{-\infty}^{\infty} f^*(t)g(t + \tau)dt. \quad (1)$$

- ❶ shift $g(t)$ by τ with respect to $f(t)$, multiply it to $f^*(t)$ and integrate over time
- ❷ $C_{fg}(\tau) \neq C_{gf}(\tau)$, $C_{fg}(\tau) = C_{gf}(-\tau)$
- ❸ The Fourier transform of C_{fg} is

$$C(\omega) = F^*(\omega)G(\omega) \quad (2)$$

Auto-correlation

If $f(t) = g(t)$, then cross-correlation is simplified to the auto-correlation of $f(t)$ itself:

$$a(\tau) = \int_{-\infty}^{\infty} f^*(t)f(t + \tau)dt \quad (3)$$

- ① provides information on how similar the signal is to itself under time translation.
- ② $C(\omega) = F^*(\omega)F(\omega) = |F(\omega)|^2$, i.e. proportional to energy spectrum
- ③ sometimes for periodic signal or ergodic signal, auto-correlation is defined as:

$$a(\tau) = \frac{1}{T} \int_0^T f^*(t)f(t + \tau)dt \quad (4)$$

then $A(\omega) = \lim_{T \rightarrow \infty} \frac{1}{T} |F(\omega)|^2$, is the **power spectrum**.

Power spectrum of a stationary process

Recall that

$$F_T(\omega) = \int_{-T/2}^{T/2} x(t) e^{-i\omega t} dt. \quad (5)$$

Therefore, the power spectrum can be expressed as follows:

$$\begin{aligned} P(\omega) &= \lim_{T \rightarrow \infty} \frac{1}{T} \langle |F_T(\omega)|^2 \rangle \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \langle \int_{-T/2}^{T/2} \int_{-T/2}^{T/2} x(t) x(t') e^{-i\omega t} e^{i\omega t'} dt' dt \rangle \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \int_{-T/2}^{T/2} \langle x(t) x(t') \rangle e^{-i\omega(t-t')} dt' dt \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \int_{-T/2}^{T/2} A(\tau) e^{-i\omega\tau} dt' d\tau \\ &= \lim_{T \rightarrow \infty} \int_{-T/2}^{T/2} A(\tau) e^{-i\omega\tau} d\tau = \int_{-\infty}^{\infty} A(\tau) e^{-i\omega\tau} d\tau \end{aligned} \quad (6)$$

Power spectrum of a stationary process

- Sometimes for a periodic signal or ergodic signal, $A(\omega) = \lim_{T \rightarrow \infty} \frac{1}{T} |F(\omega)|^2$, is the **power spectrum**, where

$$A(\tau) = \frac{1}{T} \int_0^T f^*(t) f(t + \tau) dt \quad (7)$$

- We define Power spectrum of a stationary process as the averaged power spectrum of realizations

$$P(\omega) = \lim_{T \rightarrow \infty} \frac{1}{T} \langle |F_T(\omega)|^2 \rangle \quad (8)$$

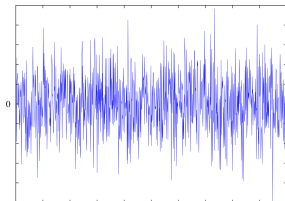
Stationary Process and LTI system

- ① At different frequencies

$$\begin{aligned} & \lim_{T \rightarrow \infty} \frac{1}{T} \langle F_T(\omega_1) F_T(\omega_2) \rangle \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} A(\tau) e^{-i\omega_1 \tau} d\tau \int_{-T/2}^{T/2} e^{-i(\omega_1 - \omega_2)t'} dt' = 0 \quad (9) \end{aligned}$$

i.e. $F(\omega_1)$ and $F(\omega_2)$ are uncorrelated if $\omega_1 \neq \omega_2$, i.e. a stochastic process consists of an infinite set of uncorrelated oscillators.

- ② White noise X_i : X_i 's are indep., $\langle X_i \rangle = 0$ and $\text{Var}(X_i) = \sigma^2$



$a(t) = \delta(t)$ and its power spectrum $P(\omega) = 1$

Stationary Process

- 1 Stationary process can be considered as a convolution of a white noise with a LTI system. $P_s(\omega) = P_{LTI}(\omega)$, where $P_{LTI}(\omega)$ is the power spectrum of the deterministic system function.
- 2 How do we estimate the Power spectrum?

$$P(\omega) = \lim_{T \rightarrow \infty} \frac{1}{T} \langle |F_T(\omega)|^2 \rangle \quad (10)$$

- 3 **Periodogram** estimates: for one realization $x_T(t) \Leftrightarrow F_T(\omega)$:

$$\hat{P}(\omega) = \frac{1}{T} |F_T(\omega)|^2 \quad (11)$$

- 4 Increasing T , or increasing number of points n will give better (finer) frequency resolution $1/T$. However, taking a long time series does not accomplish the averaging. Average over different realizations is required.

Periodogram

- ① We know that

$$A(\tau) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T x(t)x(t+\tau)dt \Leftrightarrow P(\omega). \quad (12)$$

However a finite sample of $x(t)$, $0 \leq t \leq T$ makes the integration less accurate as $t + \tau$ falls out of $[0, T]$, which will in turn affect the accuracy of values at every single ω of $\hat{P}(\omega)$ (fluctuations at all f). Indeed,

$$\text{Var}(\hat{P}(\omega)) = P^2(\omega). \quad (13)$$

- ② How to reduce this scatter? Averaging over realizations!
- ③ Make multiple measurements: split data into M sub-series, and average over periodogram estimates of all, then $\text{Var}(\hat{P}_M(\omega)) = \frac{1}{M} P^2(\omega)$, however, frequency spacing becomes M times of the entire series.
- ④ Or sum M neighbouring frequencies of $\hat{P}(\omega)$.

Importance of window function

- 1 What is the effect of estimating the autocorrelation function of $x(t)$ over the finite interval $0 \leq t \leq T$?

$$a(\tau) = \frac{1}{T} \int_0^T x(t)x(t+\tau) dt \Leftrightarrow \hat{P}(\omega) \quad (14)$$

- 2 $a(\tau)$ values are less accurate as τ approaches T .
- 3 Introduces leakage in the frequency components.
- 4 This can be corrected by applying window functions to $a(\tau)$ to reduce the effect of inaccuracy: $w(\tau)a(\tau) \Leftrightarrow \hat{P}_w(\omega)$
- 5 What is the effect of windowing?

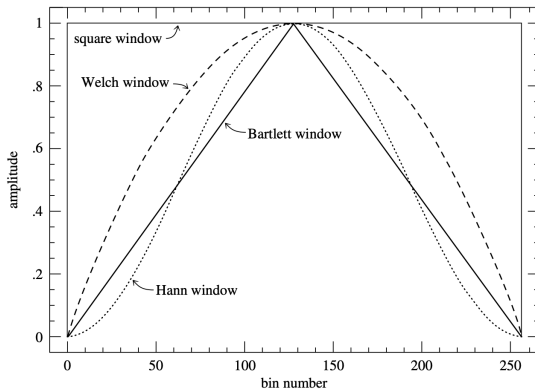
$$\hat{P}_w(\omega) = \int_{\theta} W(\omega - \theta) P(\theta) d\theta \quad (15)$$

local average of the true spectrum.

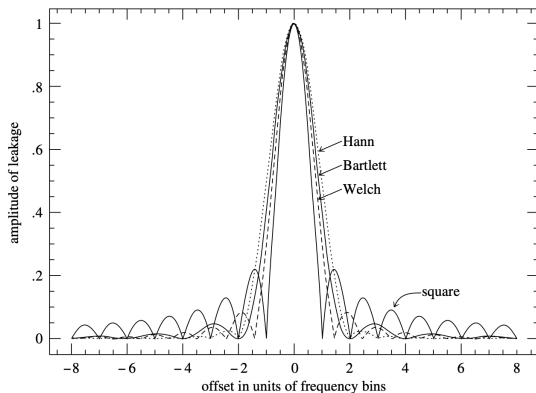
Window Functions

Hann/Hanning/Cosine bell window

$$w(t) = \frac{1}{2} \left(1 - \cos \frac{2\pi t}{T} \right) \quad (16)$$



Response of window functions



Effect of windowing:

- wider center bands, lower frequency resolution
- subdued side bands, less spectral leakage, less variance.