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The repository principally contains the compiled products rather than the source for size reasons.

- ▶ Associated Python code (as Jupyter notebooks mostly) will be held on the same repository. The source data however might be big, so I am going to be naughty and possibly just refer you to where you might get the data if that is the case (e.g. JRA-55 data). I know I should make properly reproducible binders etc., but I didn't...
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# OCES 3301 : basic Data Analysis in ocean sciences

Session 8: fairly basic time-series analysis

# Outline

(Just overview here; for actual content see Jupyter notebooks)

- ▶ Fourier analysis
  - background and theory
  - power spectrum
  - Fourier analysis as a filtering tool

# Signal analysis

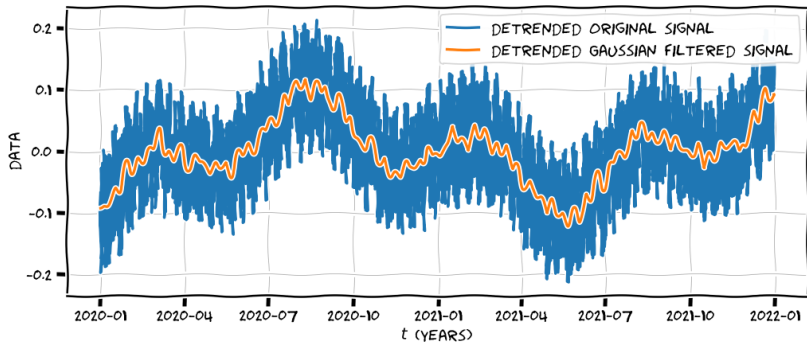


Figure: A detrended signal.

- ▶ low-passed signal still has some oscillations
- ▶ how to pick these out though?
  - can we say which period of oscillation is larger definitively?

## Recall: waves

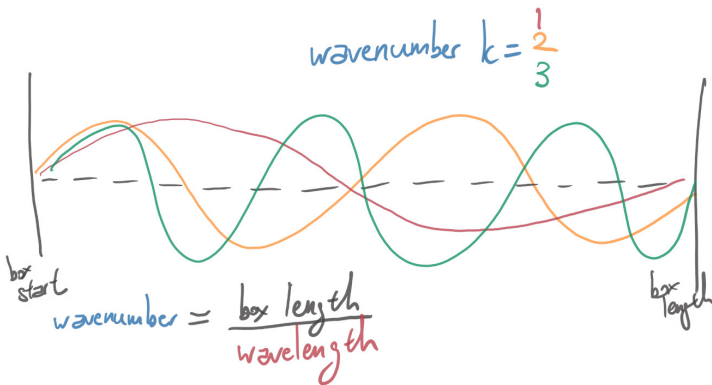


Figure: Schematic of wave features. Box length  $L = 2\pi$  for simplicity.

- displacement  $\eta$  described by (could also be sine)

$$\eta \sim A \cos(2x), \quad \eta \sim A \cos(1x), \quad \eta \sim A \cos(3x)$$

## Recall: waves

$$\gamma = \frac{v}{\lambda}, \quad k = \frac{2\pi}{\lambda}$$

- ▶  $\gamma$  the **frequency** (units:  $\text{s}^{-1} = \text{Hz}$ )  
→ how quickly the wave oscillates
- ▶  $v = c_p$  the **phase velocity**  
→ how fast the wave itself moves around
- ▶  $\lambda$  the **wavelength**  
→ how long the wave is
- ▶  $k$  the **wavenumber**  
→ intuitively how many waves can you fit in a box (so  $k \sim \lambda^{-1}$ )  
→ does not necessarily have to be an integer

# Fourier analysis

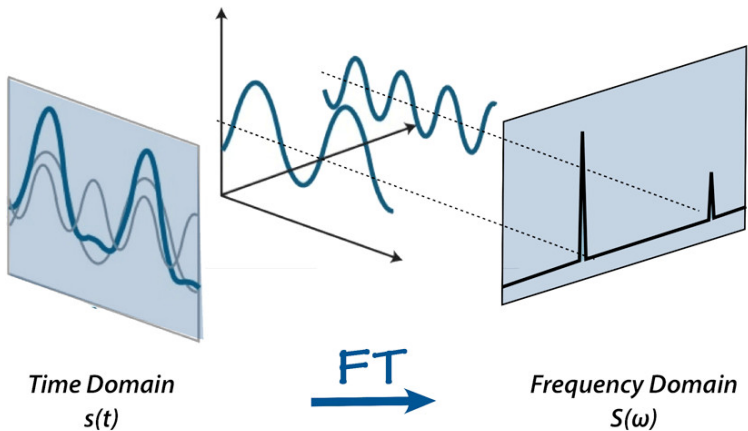


Figure: Schematic (?) of a Fourier transform.

# Fourier analysis

- generically (!) a signal can **uniquely** be written as

$$\begin{aligned} f(t) &= a_0 + a_1 \cos(t) + a_2 \cos(2t) + \dots + b_1 \sin(t) + b_2 \sin(2t) + \dots \\ &= a_0 + \sum_{k=1}^{\infty} a_k \cos(kt) + \sum_{l=1}^{\infty} b_l \sin(lt), \quad f \in L^2(P). \end{aligned}$$

- $(k, l)$  are the **wavenumbers**  
→ really **angular frequencies** here (since dealing with time)
- $(a_k, b_l)$  are the **amplitudes** associated with the oscillations at  $(k, l)$

**magnitudes of  $(a_k, b_l)$  give you power of the oscillations at frequencies/wavenumbers at  $(k, l)$**



# Fourier analysis: background

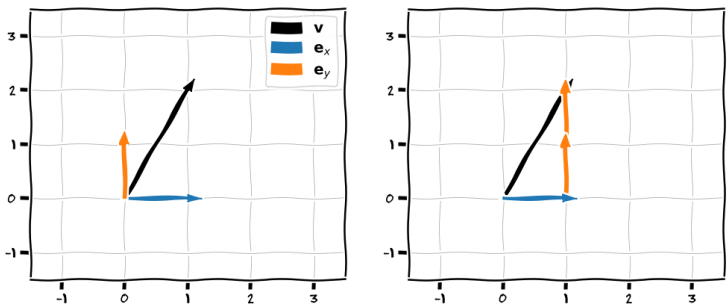


Figure: Demonstration of canonical basis in  $\mathbb{R}^2$ .

$$\mathbf{v} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} = 1\mathbf{e}_x + 2\mathbf{e}_y = 1 \begin{pmatrix} 1 \\ 0 \end{pmatrix} + 2 \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

# Fourier analysis: background

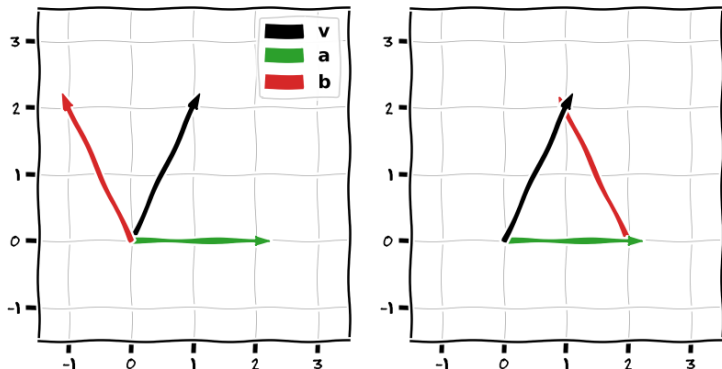


Figure: Demonstration of alternative basis in  $\mathbb{R}^2$ .

$$\mathbf{v} = \begin{pmatrix} 1 \\ 2 \end{pmatrix} = 1 \begin{pmatrix} 2 \\ 0 \end{pmatrix} + 1 \begin{pmatrix} -1 \\ 2 \end{pmatrix} = 1\mathbf{a} + 1\mathbf{b}$$

# Fourier analysis: background

- ▶ cf. change of units (e.g. 1 inch  $\approx$  2.5 cm)
  - ▶ cf. papillon vs. Schmetterling vs. فراشة
  - ▶ cf. change of co-ordinates (e.g. PCA<sub>in 04</sub>)
  - ▶ cf. uses different Lego combination to make the same thing
  - ▶ cf. below?
- 
- ▶ water, 35 L; carbon, 20 kg;  
ammonia, 4 L; lime, 1.5 kg;  
phosphorus, 800 g; salt, 250 g;  
saltpeter, 100 g; sulfur, 80 g;  
fluorine, 7.5 g; iron, 5 g;  
silicon, 3 g; trace amounts of  
15 other elements



Figure: "Ed...ward..." (sorry not sorry)

# Fourier analysis: background

A **basis** of (say)  $\mathbb{R}^2$  is a set  $\{\mathbf{v}_1, \mathbf{v}_2\}$ ,  $\mathbf{v}_i \in \mathbb{R}^2$  where

- ▶ the elements themselves are **linear independent** of each other, i.e.

$$\sum_i a_i \mathbf{v}_i = \mathbf{0} \quad \Leftrightarrow \quad a_i = 0 \quad \forall i$$

- ▶ the set **spans**  $\mathbb{R}^2$ , i.e.,

$$\forall \mathbf{u} \in \mathbb{R}^2, \exists \{b_i\} : \mathbf{u} = \sum_i b_i \mathbf{v}_i.$$

- ▶ note **uniqueness** follows from the two definitions
- ▶ no criteria about  $\mathbf{v}_i$  being **orthogonal** to each other  
→ elements of canonical basis  $\{\mathbf{e}_i\}$  are mutually orthogonal and unit length (i.e. **orthonormal**)

# Fourier analysis

Claim:

- ▶  $\{\cos(kt), \sin(lt)\}$  is a basis of the space of functions  $L^2(P)$ , where the space is equipped with the **inner product**

$$\langle f, g \rangle = \int_I f(t)g(t) \, dt$$

- ▶ with respect to that inner product, the Fourier basis is an **orthogonal basis**, i.e.,

$$\int_I \sin(kt) \cos(lt) \, dt = \int_I \sin(kt) \sin(lt) \, dt = \int_I \cos(kt) \cos(lt) \, dt = 0.$$

# Fourier analysis

- ▶ a **Fourier transform** takes  $f(t) \in L^2(P)$  into the **co-ordinates**  $(a_k, b_l)$   
→ e.g. by orthogonality,

$$\begin{aligned}\int_I f(t) \cos(t) \, dt &= \int_I \cos(t) \left[ a_0 + \sum_{k=1}^{\infty} a_k \cos(kt) + \sum_{l=1}^{\infty} b_l \sin(lt) \right] dt \\ &= a_1 \int_I \cos^2(t) \, dt,\end{aligned}$$

where we can in principle compute the integrals  
(numerically or otherwise)

→  $f(t)$  is in **time-domain** or **real space**

→  $(a_k, b_l)$  is in **frequency-domain** or **spectral space**

- ▶ `scipy.fft.fft/rfft` uses the **Fast Fourier Transform**  
to do it (Cooley–Tukey 1965 algorithm, although was actually known to Gauss 1876)

# Fourier analysis

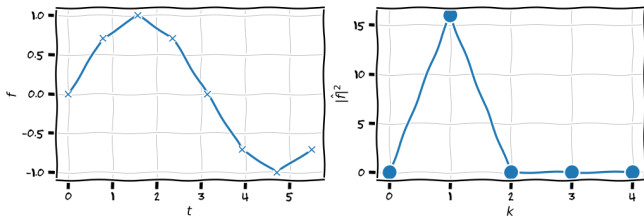
- ▶ a trivial example: taking  $I = [0, 2\pi]$ ,  $f(t) = \sin(t)$ 
  - choice of  $I$  means wavenumbers  $(k, l) \in \mathbb{Z}$
  - $b_1 = 1$ , everything else zero
- ▶ transform is (up to **normalisation**)

$$f(t) = (f(t_0), f(t_1), \dots) \leftrightarrow \hat{f} \sim (a_0, a_1 + ib_1, a_2 + ib_2, \dots),$$

where  $i = \sqrt{-1}$ , and  $(a_k, b_l) \in \mathbb{R}$

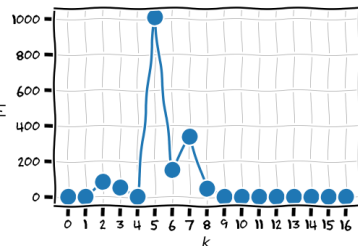
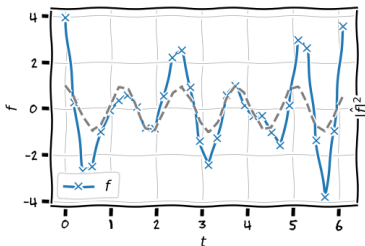
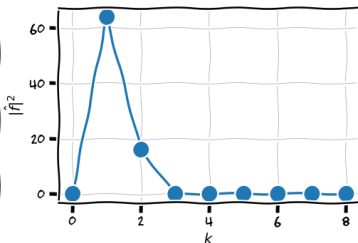
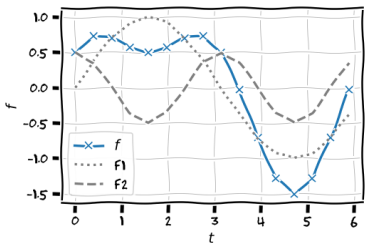
- ▶ **power spectrum** as a function of  $k$  is

$$|\hat{f}|^2 \sim (a_0^2, |a_1|^2 + |b_1|^2, |a_2|^2 + |b_2|^2, \dots)$$



# Fourier analysis

- still pure waves (sines and cosines) on  $I = [0, 2\pi]$



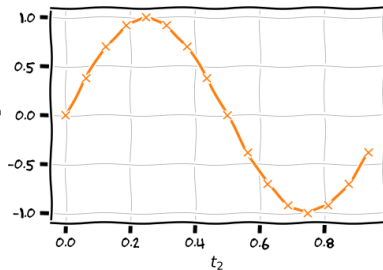
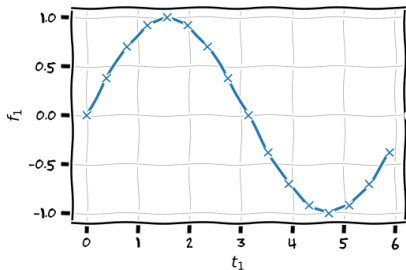


# Fourier analysis: change of domain

- ▶ still pure waves (sines and cosines) but now on  $I = [0, L]$
- ▶ take instead

$$\{\cos(kt), \sin(lt)\} \leftrightarrow \left\{ \cos\left(\frac{2\pi k}{L}t\right), \sin\left(\frac{2\pi l}{L}t\right) \right\}$$

- ▶ below is when  $L = 1$ , showing  $f_2(t) = \sin(2\pi t)$  (so  $l = 1$ )



## Fourier analysis: change of domain

- ▶ still pure waves (sines and cosines) but now on  $I = [0, L]$
- ▶ it's just a stretch/contraction, so just need to re-scale the definition of  $(k, l)$ , as

$$k(\in \mathbb{Z}) \rightarrow \frac{2\pi k}{L}$$

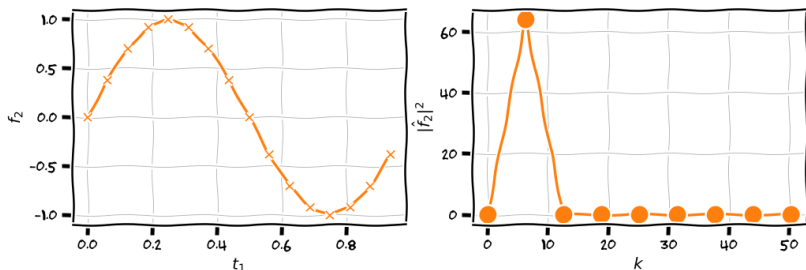


Figure: For  $I = [0, 1]$ . Compare with result on  $I = [0, 2\pi]$  four slides ago.

# Fourier analysis: change of units

Taking  $t$  to have units of seconds:

- ▶  $k$  has units of radians per second (angular frequency here)
- ▶  $k = 2\pi\nu$ , so **frequency**  $\nu = k/(2\pi)$  and has units of per second, or **Hertz** (Hz)
- ▶  $\nu = 1/T$ , so **period**  $T = 2\pi/k$ , here having units of seconds
- ▶ again, start with  $k \in \mathbb{Z}$  and re-scale accordingly

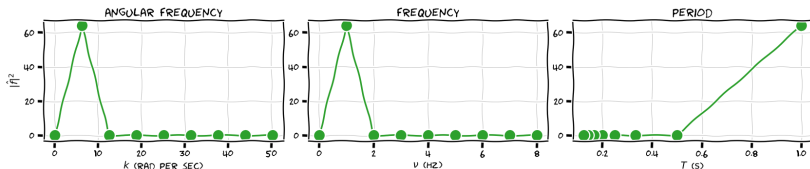
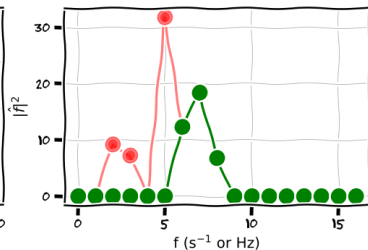
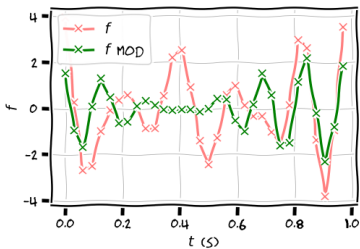
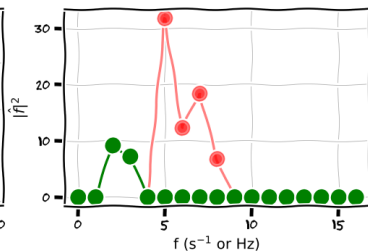
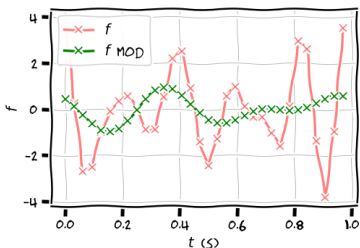


Figure: For  $I = [0, 1]$ , as previous slide, but in different units.

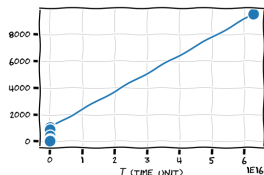
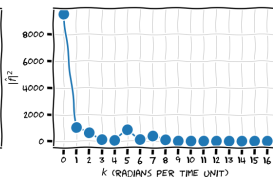
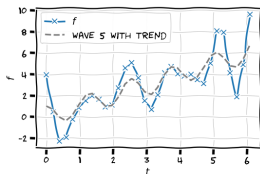
# Fourier analysis: filtering

- modify the spectrum and do inverse Fourier transform



# Things to be careful: signals with trends / non-zero

- ▶ a signal with some trend

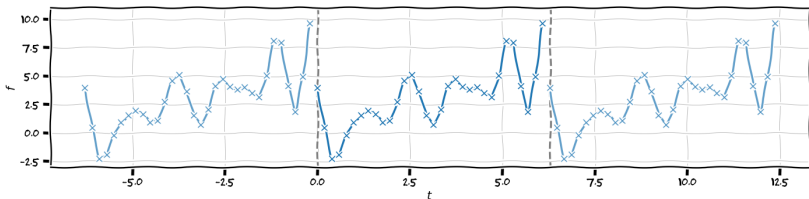


→ massive power in the  $k = 0$  mode and in one with massive period

- ▶ non-zero mean so  $a_0 \neq 0$ , but this thing has no period
  - if you know there is a trend then detrend it
  - you could also get rid of the mean
  - or ignore / don't plot it, because the mean has no defined period of oscillation anyway

# Things to be careful: non-periodic signal

- signal is not periodic (i.e. not strictly in  $L^2(P)$ )



- the FFT routines don't actually care, and will go ahead and return a spectrum

# Things to be careful: non-periodic signal

- ▶ can force signal to be periodic through a **window function**

(cf. lec 07)

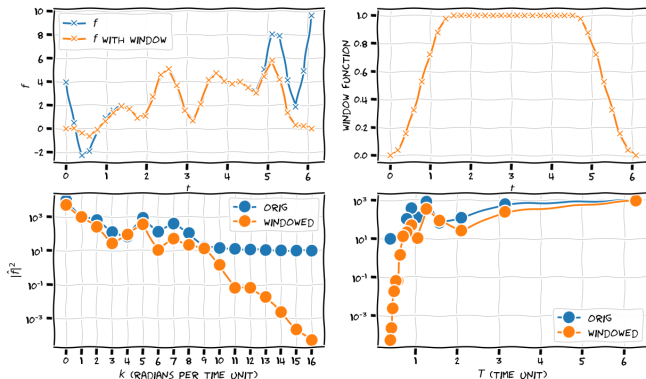


Figure: Example with Tukey window.

- ▶ different window functions look different in real space, and have different responses in spectral space

# Things to be careful: non-smooth signal

- Fourier transform actually still works

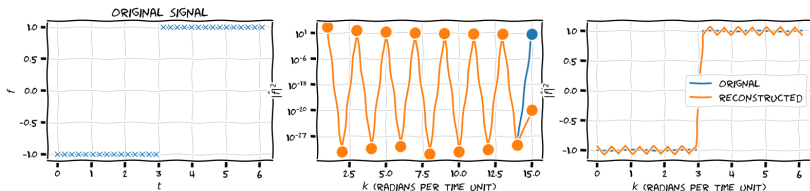


Figure: Classic example of square wave with Gibbs phenomenon.

- power is now everywhere and decays slowly with increasing wavenumber
- minor modification in spectrum leads to saw-tooth pattern  
→ Gibbs phenomenon or Gibbs oscillations



# Things to be careful: non-smooth signal

- signal with noise

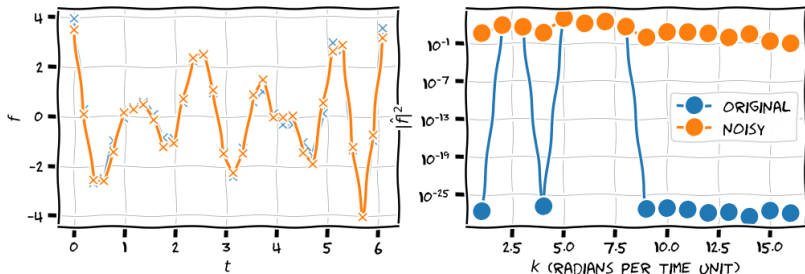


Figure: Power spectrum of signal, and signal with a bit of noise.

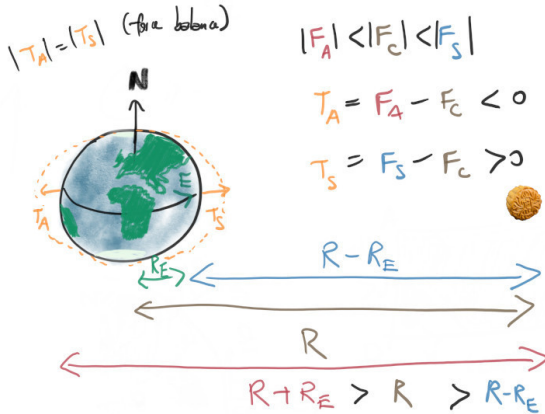
- power is now everywhere and decays slowly with increasing wavenumber

# A real example: tides (see OCES 2003, lec 18)



**Figure:** High (or flood) and low (or ebb) tide at Tobermory, Isle of Mull, Scotland, using the pastel pink and red house as references. Modified images from [www.thechaoticscot.com](http://www.thechaoticscot.com) (left) and from myself (right).

# A real example: tides (see OCES 2003, lec 18)



**Figure:** Schematic of tidal forcing by an astronomical body. Assume instantaneous response (“equilibrium theory”). No rotation is assumed here.

## A real example: tides (see OCES 2003, lec 18)

symbol	period (in solar hrs)	rel. amp (to $M_2$ )	name
$M_2$	12.42	1	principal lunar (semi-diurnal)
$K_1$	23.93	0.58	luni-solar (diurnal)
$S_2$	12.00	0.47	principal solar (semi-diurnal)
$O_1$	25.82	0.42	principal lunar (diurnal)
$N_2$	12.66	0.19	larger lunar elliptic (semi-diurnal)
$\vdots$	$\vdots$	$\vdots$	$\vdots$
Mf	327.85 ( $\approx 14$ days)	0.09	lunar fortnightly
Mm	661.30 ( $\approx 28$ days)	0.05	lunar monthly
SSa	4382.86	0.04	solar semi-annual

**Table:** Some sample tidal forcings sorted by relative amplitude to the  $M_2$  tide (which is the largest forcing for Earth). Subset of Table 6.2 given in Wunsch (2015). The last few entries are weak and long term but they are there.

- ▶  $M_2$  and  $K_1$  the dominant ones  
→ usually do include these two in **numerical models**
- ▶ notice the periods are close to multiples of 12 hours

## A real example: tides (see OCES 2003, lec 18)

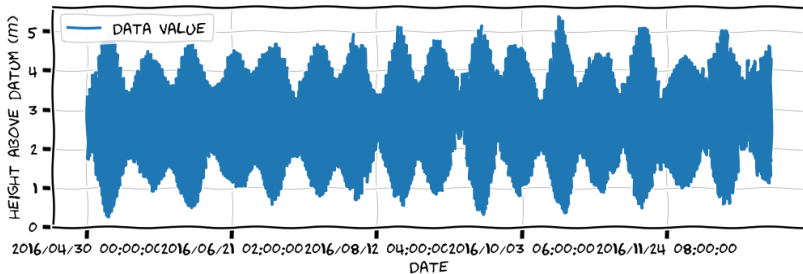
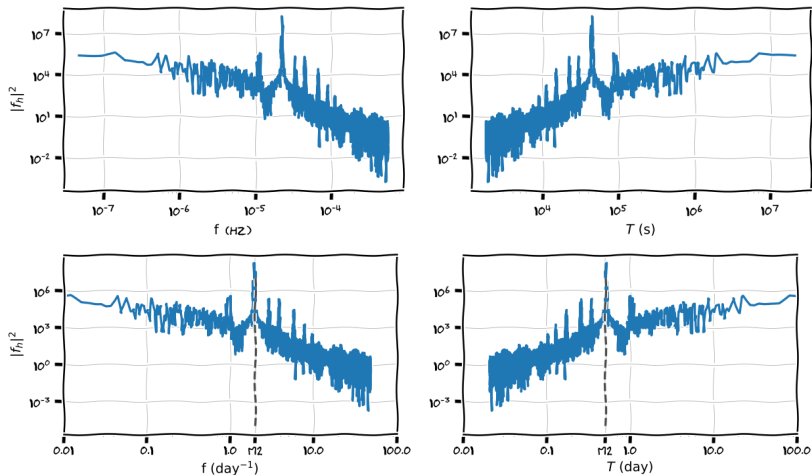


Figure: Data segment from BODC of sea level above datum at Tobermory.

- ▶ short-ish period, ignore trend with rising sea level  
→ going to be ignoring  $k = 0$  mode anyway

# A real example: tides (see OCES 2003, lec 18)



**Figure:** Power spectrum with respect to different quantities in different units, and denoting the M2 tide.

# Jupyter notebook

go to 08 Jupyter notebook to get some code practise

- ▶ try something similar for the El-Niño 3.4 data
  - be careful of trends
  - be careful of units (time units is in **years**)

**Note:** none of the content I introduced in ‘times series’ are exclusive to ‘time’, and works just as well for ‘space’ too (see lec 09 and 10)