



Some of the more “traditional” methods of interpretation

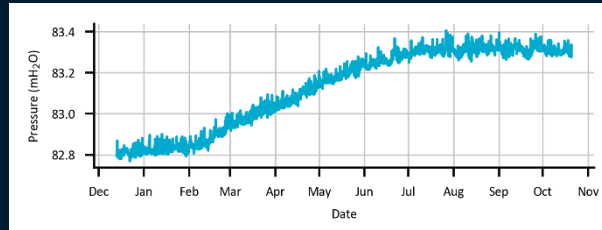
Using these to:

1. Identify **Earth tide** influences
2. Identify **atmospheric** influences

And derive conceptual and quantitative insights from these

1. Are my data affected by Earth tides?

$$P_{GW}(t)$$



1. Forcing:

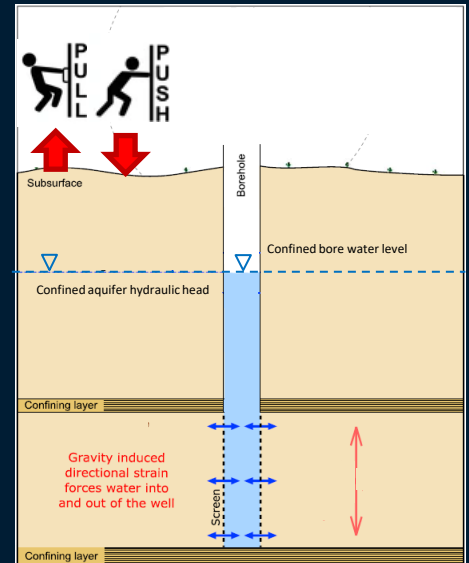
a) Earth tide (gravity) variation at nm to μm scale

2. Responses: (i.e. do we see a pressure change?)

a) Confined aquifer: yes

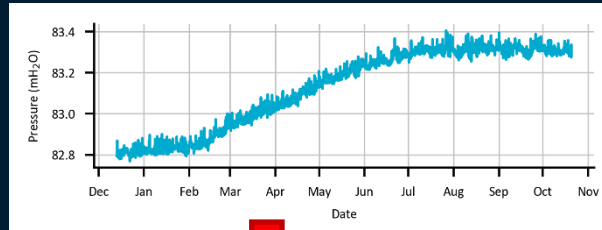
b) Open borehole: no

c) Unconfined aquifer: no

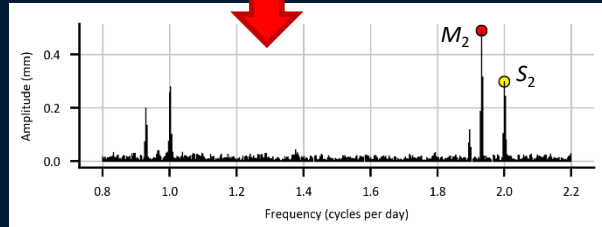


1. Are my data affected by Earth tides?

$$P_{GW}(t)$$



$$A(f)$$



Method:
Fourier transform

Result:
Amplitude spectrum for a
continuous range of frequencies

Regular Python:

- 1 frequencies = numpy.fft.rfftfreq(GWdata, 1./24.)
- 2 amplitudes = numpy.abs(numpy.fft.rfft(GWdata))
- 3 matplotlib.pyplot.plot(frequencies, amplitudes)

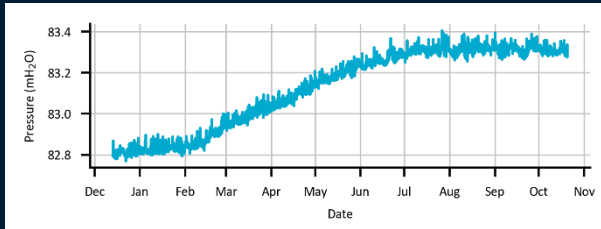
HydroGeoSines package:

- 1 output = hgs_analysis.Freq_domain.fft_comp(1./24., GWdata)
- 2 frequencies = output['freq']
- 3 amplitudes = numpy.abs(output['complex'])
- 4 matplotlib.pyplot.plot(frequencies, amplitudes)

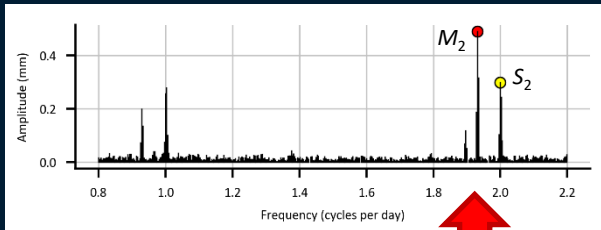
1. Are my data affected by Earth tides?

Interpretation:

$$P_{GW}(t)$$



$$A(f)$$

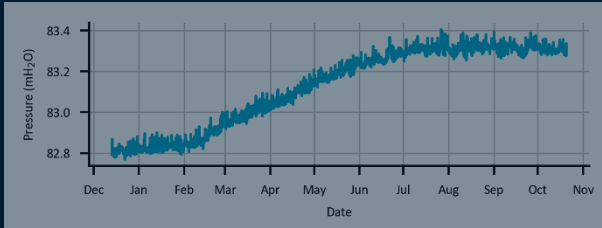


- M_2 amplitude is **dominant**
- Indicates **confined** conditions

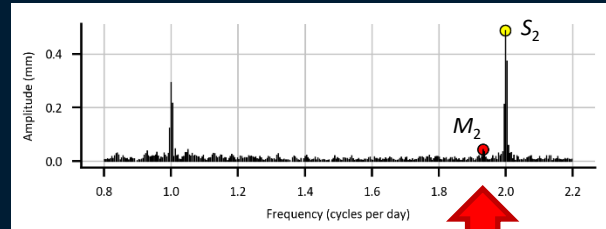
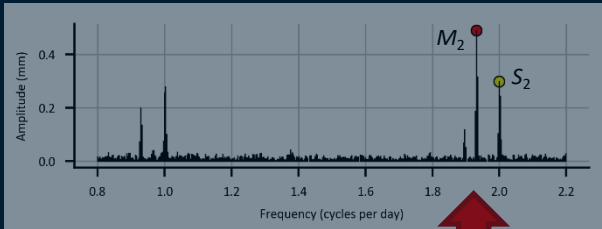
1. Are my data affected by Earth tides?

Interpretation:

$P_{GW}(t)$



$A(f)$

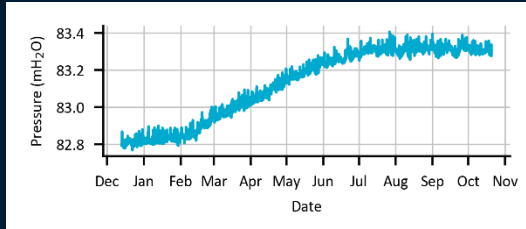


- M_2 amplitude is **dominant**
- Indicates **confined** conditions

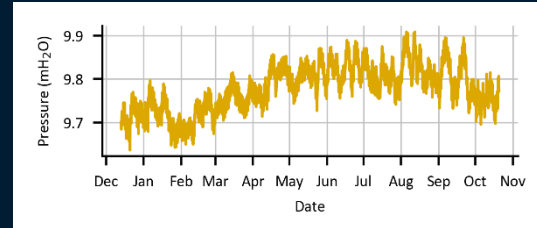
- M_2 amplitude is **negligible**
- Indicates **unconfined** conditions

2. Are my data affected by atmospheric influences?

$P_{GW}(t)$



$P_{BA}(t)$

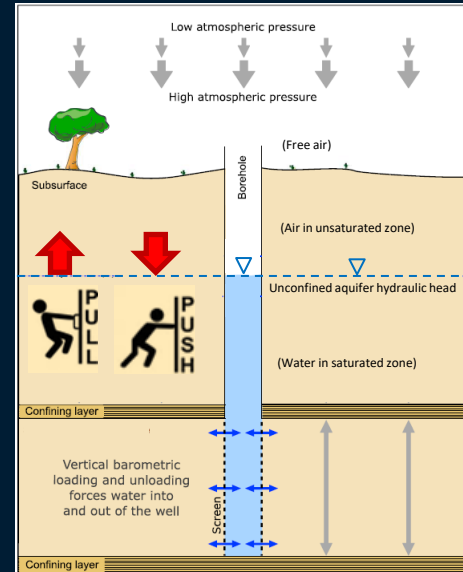


1. Forcing:

a) **Barometric pressure** variation at mm to cm scale

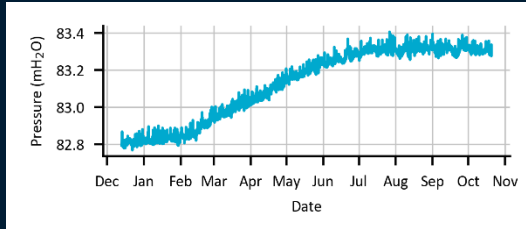
2. Responses: (i.e. to a change in load)

- a) Open borehole: none
- b) Confined aquifer: directly proportional
- c) Leaky (semi-confined) aquifer: variably proportional
- d) Unconfined aquifer: none

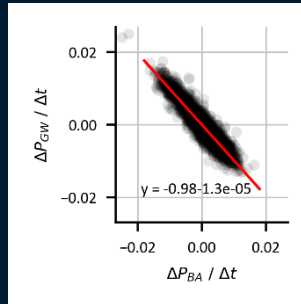
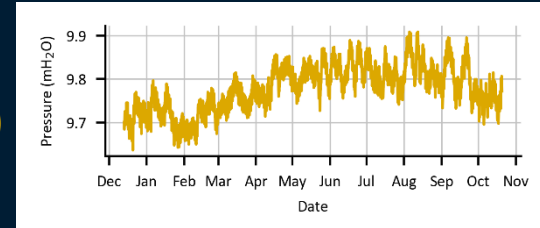


2. Are my data affected by atmospheric influences?

$P_{GW}(t)$



$P_{BA}(t)$



Method:
Linear regression

Result:
Barometric efficiency
from slope

Regular Python:

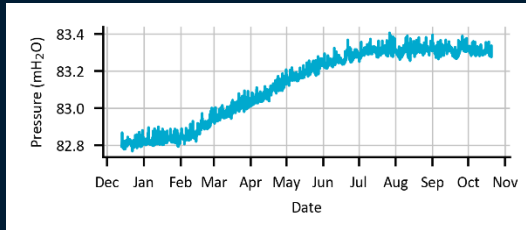
```
BE = scipy.stats.linregress (dBP_dt, dGW_dt)[0]
```

HydroGeoSines package:

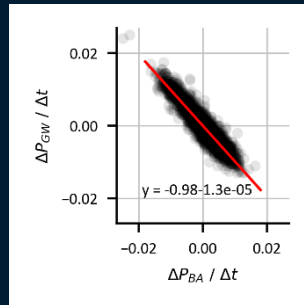
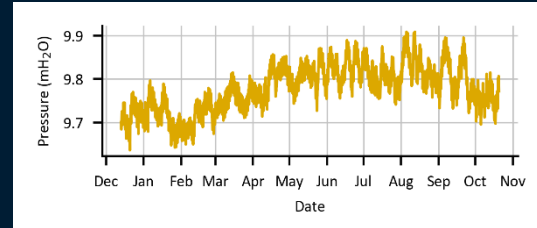
```
BE = hgs_analysis.Time_domain.BE_linear_regression(dBP_dt, dGW_dt)
```

2. Are my data affected by atmospheric influences?

$P_{GW}(t)$



$P_{BA}(t)$



Interpretation:

- $BE \approx 1$: confined aquifer
- $BE \approx 0$: unconfined aquifer
- $0 < BE < 1$: semi-confined / leaky aquifer

2. Are my data affected by **atmospheric influences**?

Deriving further benefit from barometric efficiency estimates:

- Aquifer specific storage is commonly calculated as:

$$S_s = \rho_w g (C_f + C_w \eta_e)$$

- It can instead be defined as a function of barometric efficiency, as:

$$S_s = \rho_w g C_w \eta_e BE^{-1}$$

Density of water,
 $\approx 997 \text{ kg/m}^3$

Gravitational acceleration,
 $\approx 9.8065 \text{ m/s}^2$

Compressibility of water,
 $\approx 4.58 \times 10^{-10} \text{ Pa}^{-1}$

Barometric efficiency

Aquifer effective porosity



Recap of “traditional” methods of interpretation:

1. Identifying **Earth tide** influences via:
 - Fourier transform
 - Aquifer confinement status
2. Identifying **atmospheric** influences via:
 - Linear regression
 - Aquifer confinement status
 - Confined aquifer specific storage