## **HydroGeoSines**

Signal In the Noise Exploration Software for hydrogeological datasets

## **Code demonstration**

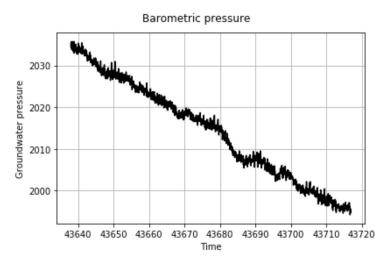
First, we import the sines package and create a new instance of a sines model.

```
In [1]: import hydrogeosines
s = hydrogeosines.model()
```

Next, we import groundwater pressure data, take a look at the first ten values, and then visualise the full dataset.

In addition to the barometric pressure dataset, we can assess one of three groundwater pressure datasets: RN027214, RN039613, or RN039617.

```
In [2]: s.wd = 'hydrogeosines/test_data/port_keats/'
        s.id = "RN039613"
        s.get GW()
        s.print_GW(10)
        s.plot_GW()
              time pressure
         43638.000
                    2035.359
         43638.040 2034.306
         43638.080 2035.704
         43638.130 2034.636
         43638.170 2034.715
         43638.210 2035.108
         43638.250 2033.772
         43638.290 2035.249
         43638.330 2035.075
         43638.380 2035.768
```



Next, we import groundwater pressure data, take a look at the first ten values, and then visualise the full dataset.

```
In [3]: | s.get_BA()
         s.print_BA(10)
         s.plot BA()
               time pressure
          43638.000 1032.010
          43638.040 1032.004
          43638.080 1031.804
          43638.130 1032.470
          43638.170 1032.487
          43638.210
                     1033.031
          43638.250
                     1033.424
          43638.290
                     1033.127
          43638.330
                     1034.121
          43638.380
                     1033.933
            1036
         Barometric pressure
            1034
            1032
            1030
```

Next, we calculate the linear regression of groundwater pressure versus barometric pressure, print the resulting parameters, and then plot the data and the estimated linear function.

43680

Time

1028

43640

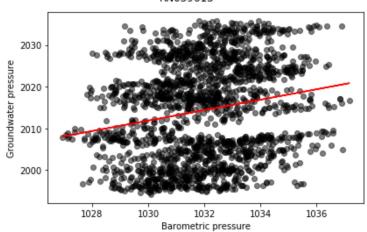
43650 43660 43670

```
In [4]: s.calc_linear_GW()
%matplotlib inline
s.plot_linregress(pname=None)

1.251 723.807

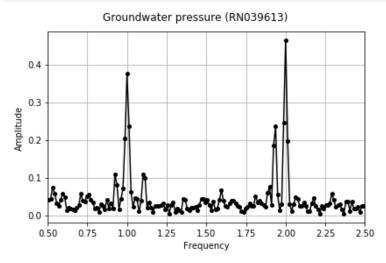
RN039613
```

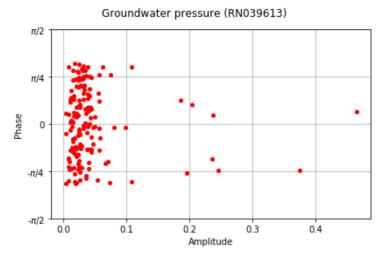
43690 43700 43710 43720



Next, we calculate the discrete Fourier transform of the groundwater pressure dataset and plot the resulting amplitude spectrum, as well as an amplitude versus phase plot, in order to identify dominant component frequencies.

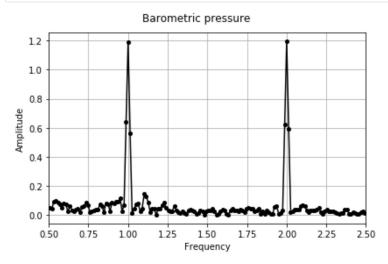
```
In [5]: s.calc_ft_GW()
%matplotlib inline
s.plot_ft_avf_GW(pname=None)
s.plot_ft_pva_GW(pname=None)
```

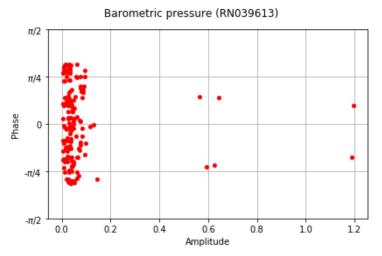




Next, we repeat the analysis for the barometric pressure dataset.

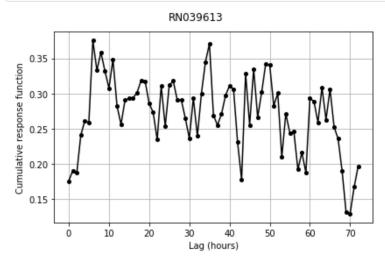
```
In [6]: s.calc_ft_BA()
%matplotlib inline
s.plot_ft_avf_BA(pname=None)
s.plot_ft_pva_BA(pname=None)
```

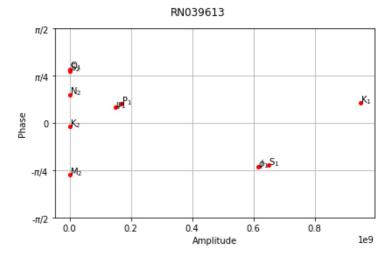




Next, we use regression deconvolution to estimate the amplitudes and phases of specific frequencies of interest, and to estimate a cumulative response function.

```
In [7]: s.calc_regress_deconv()
    s.plot_regress_deconv_crf(pname=None)
    s.plot_regress_deconv_pva(pname=None)
```





In [ ]: