



Fourier noise reduction methods



Antares data-set

Looking for periodic sources in Antares data

All data is used

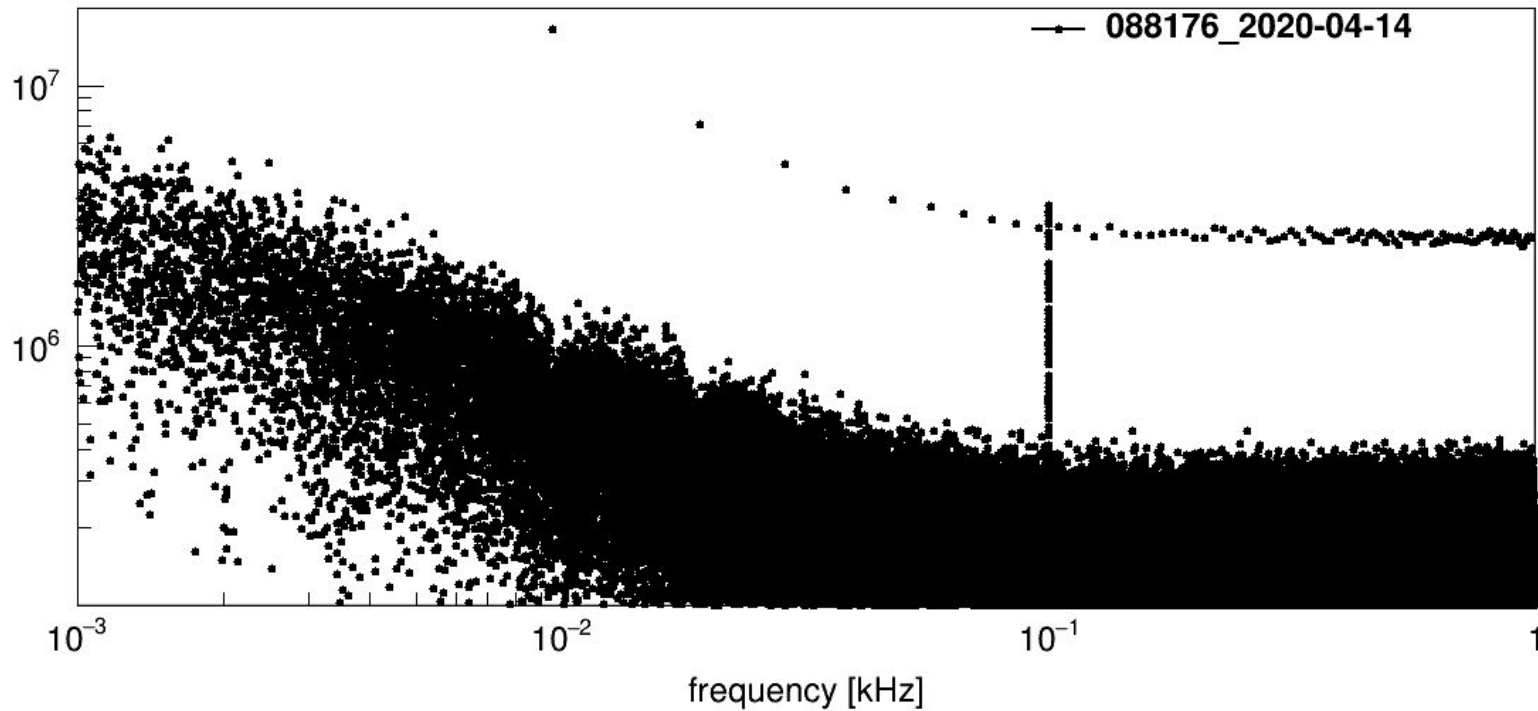
Energy \sim MeV

Histogram into (standard) fast fourier transform (FFT)

Noise increases near lower frequencies

Applied transform

power [a.u.]





Methods of noise reduction

Spectrum Subtraction (SS)

Subtracting estimate of background from signal

Short time fourier transform (STFT)

Cuts histogram into smaller pieces

Difficult to achieve compatibility between t and f



Non Harmonic Analysis

Fourier coefficient is estimated by least squares method

Repeatedly applies steepest descent and Newton's method



Non Harmonic Analysis

Accurate estimation of the frequency f , amplitude A , and initial phase ϕ , avoiding the dominance of the analysis window length

Higher accuracy than theoretical upper limit of DFT

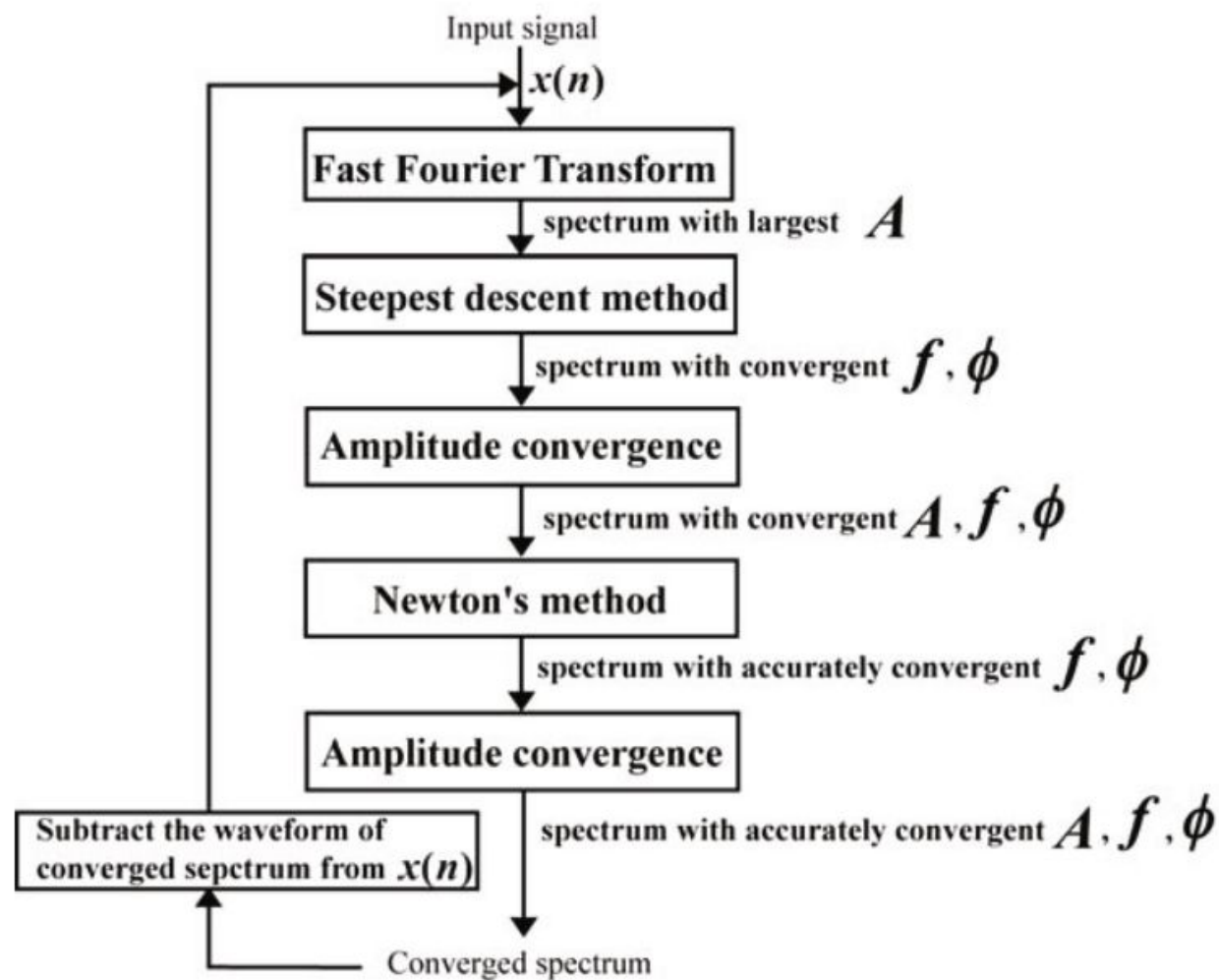


Figure 2 NHA algorithm.

$$F(A, f, \varphi) = \frac{1}{N} \sum_{n=0}^{N-1} \{x(t) - A \cos(2\pi ft + \varphi)\}^2,$$

Steepest descent

Sum over histogram

Guess a wave, any wave!

f and ϕ first, A after

$$\hat{f}_{m+1} = \hat{f}_m - \mu_m \frac{\partial F}{\partial f},$$

$$\hat{\varphi}_{m+1} = \hat{\varphi}_m - \mu_m \frac{\partial F}{\partial \varphi}.$$

$$\hat{A}_{m+1} = \hat{A}_m - \mu_m \frac{\partial F}{\partial A},$$

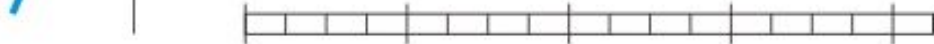


Newton's method

Find root for cost function

Since if $F = 0$ we have found a perfect fit for our wave

second order

y  x 

 Funktion
 Tangente

Second order

$$f(x + \Delta x, p + \Delta p) = f(x, p) + f_x \Delta x + \frac{1}{2} f_{xx} \Delta^2 x + f_p \Delta p + \frac{1}{2} f_{pp} \Delta^2 p + f_{xp} \Delta p \Delta x \quad (8)$$

$$\Delta x = - \frac{f_p \Delta p + \frac{1}{2} f_{pp} \Delta^2 p}{f_x + f_{xp} \Delta p} \left[1 + \frac{f_p \Delta p + \frac{1}{2} f_{pp} \Delta^2 p}{(f_x + f_{xp} \Delta p)^2} f_{xx} \right] \quad (9)$$



Applied to the cost function

$$\hat{f}_{m+1} = \hat{f}_m - \frac{v_m}{J} \begin{vmatrix} \frac{\partial F}{\partial f} & \frac{\partial^2 F}{\partial f \partial \varphi} \\ \frac{\partial F}{\partial \varphi} & \frac{\partial^2 F}{\partial \varphi^2} \end{vmatrix},$$

$$\hat{\varphi}_{m+1} = \hat{\varphi}_m - \frac{v_m}{J} \begin{vmatrix} \frac{\partial^2 F}{\partial f^2} & \frac{\partial F}{\partial f} \\ \frac{\partial^2 F}{\partial f \partial \varphi} & \frac{\partial F}{\partial \varphi} \end{vmatrix},$$

where

$$J = \begin{vmatrix} \frac{\partial^2 F}{\partial f^2} & \frac{\partial^2 F}{\partial f \partial \varphi} \\ \frac{\partial^2 F}{\partial f \partial \varphi} & \frac{\partial^2 F}{\partial \varphi^2} \end{vmatrix}.$$

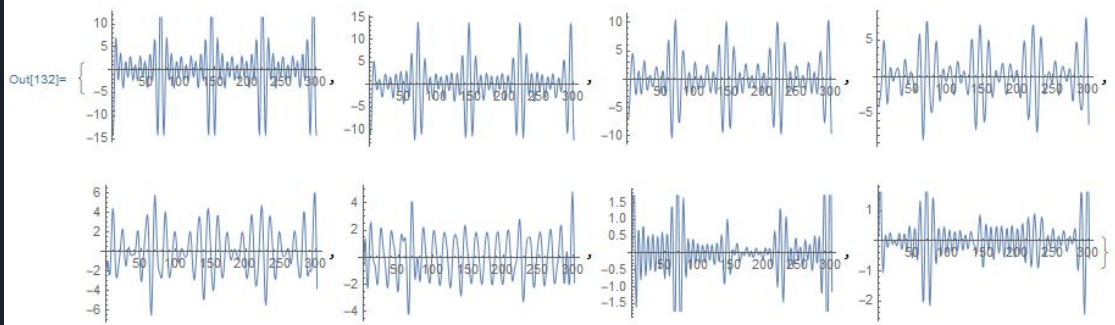


Idealized data

Searching iteratively

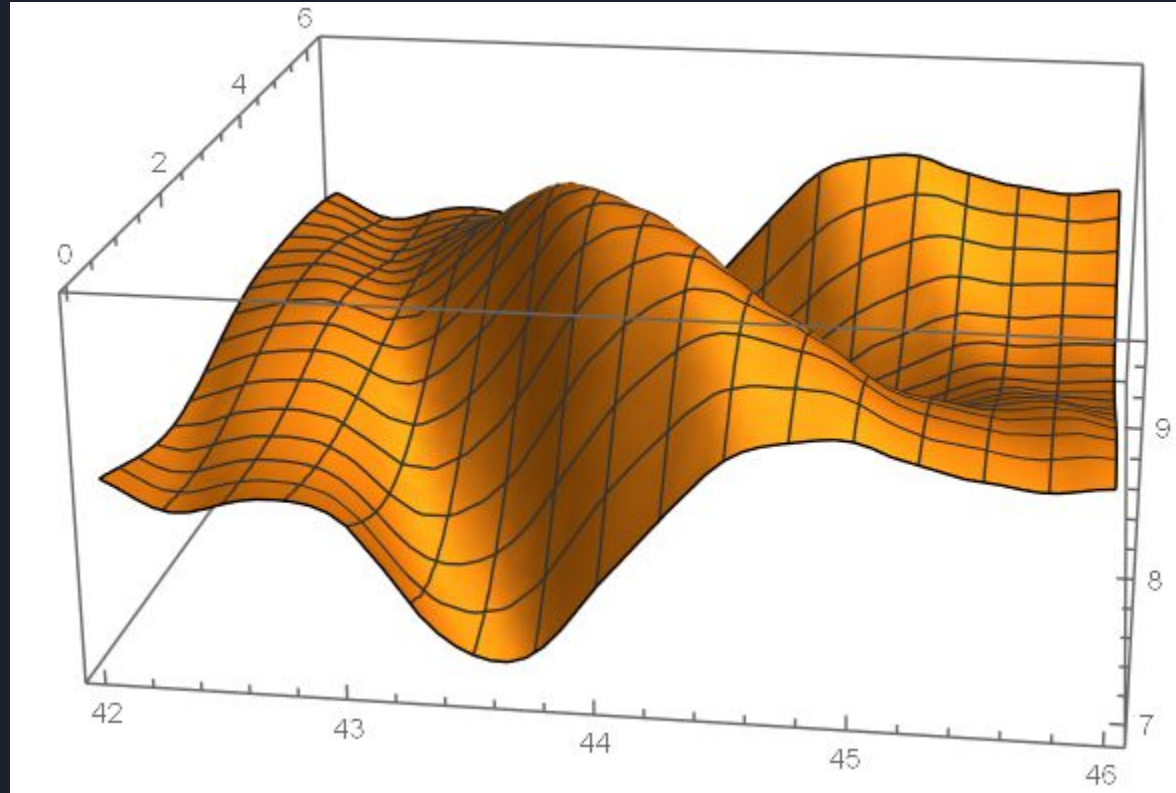
```
In[128]:= SpecifiedTestFunction = Sum[.4 k Cos[4 k 2 Pi var + .4 k], {k, 5, 10}];  
Plotlist = {};  
Guesslist = {};  
For[s = 0, s < 8, s++,  
  ourtestdata = testdata[300, 1/300, SpecifiedTestFunction];  
  ourplot = ListLinePlot[ourtestdata];  
  Print[s];  
  Plotlist = Append[Plotlist, ourplot];  
  {FOut, PhiOut, AOut} = CompleteFunction[200, 1/200, 300, 0.05, SpecifiedTestFunction, 10^(-6)];  
  Guesslist = Append[Guesslist, {FOut, PhiOut, AOut}];  
  guess = AOut Cos[FOut 2 Pi var + PhiOut];  
  SpecifiedTestFunction = SpecifiedTestFunction - guess  
]  
Plotlist  
Guesslist
```

0
1
2
3
4
5
6
7



```
Out[133]= {{40.0849, -2.59294, 4.12941}, {36.0899, -3.00513, 3.64459}, {32.0901, -3.4029, 3.18481}, {28.0819, 2.50439, 2.74158},  
{24.0595, 2.16869, 2.32578}, {20.0019, 1.94435, 1.94196}, {40.7223, -1.63267, 0.579088}, {37.521, -0.0320461, -0.0134654}}
```

Optimization space



Applied to real Antares data

```
In[ ]:= ourproducts = Timing[CompleteFunction[10^4, 10^(-4), 100, 0.05, 10^(-4)]]
```

```
Out[ ]:= {38.1406, {{64841.3, -0.834105, 0.64555},
```

