

## Homework Assignment 7

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```
In [1]: % First we call the Wavelab toolbox
WavePath;
```

```
Welcome to WaveLab v 850
Setting Global Variables
```

```
WAVELABPATH =
```

```
/Applications/MATLAB_R2014a.app/toolbox/Wavelab850/
```

```
[Warning: Function realpow has the same name as a MATLAB builtin. We
suggest you
```

```
rename the function to avoid a potential name conflict.]
```

```
[> In path at 33
```

```
    In WavePath at 132
```

```
    In pymat_eval at 31
```

```
    In matlabserver at 24]
```

```
Pathnames Successfully Set
```

```
global WAVELABPATH = "/Applications/MATLAB_R2014a.app/toolbox/Wavela
b850/"
```

```
global PATHNAMESEPARATOR = "/"; global MATLABVERSION = 8.3
```

```
global PREFERIMAGEGRAPHICS = 1
```

```
WaveLab v 802 Setup Complete
```

```
Available Demos - Figures from the following papers:
```

```
    AdaptDemo - ``Adapting to Unknown Smoothness via Wavelet Shrinkag
e''
```

```
    AsympDemo - ``Wavelet Shrinkage: Asymptopia?''
```

```
    BlockyDemo - ``Smooth Wavelet Decompositions with Blocky Coefficie
nt Kernels''
```

```
    CorrelDemo - ``Wavelet Threshold Estimators for Data with Correlat
ed Noise''
```

```
    IdealDemo - ``Ideal Spatial Adaptation via Wavelet Shrinkage''
```

```
    MESDemo - ``Minimum Entropy Segmentation''
```

```
    MIPTDemo - ``Nonlinear Wavelet Transforms based on Median-Interpo
laton''
```

```
    RiskDemo - ``Exact Risk Analysis of Wavelet Regression''
```

```
    SCDemo - ``Nonlinear Wavelet Methods for Recovery of Signals,
Densities
```

```
and Spectra from Indirect and Noisy Data''
```

```
    CSpinDemo - ``Translation-Invariant De-Noising''
```

```
    TourDemo - ``Wavelet Shrinkage and W.V.D. -- A Ten-Minute Tour''
```

```
    VdLDemo - ``WaveLab and Reproducible Research''
```

```
Available Workouts:
```

```
    BBWorkout - Workouts for Best Basis
```

```
    MPWorkout - Workouts for Matching Pursuit
```

```
    MultiFrac - Workouts for Continuous Wavelet Transform
```

```
    Toons - The Cartoon Guide to Wavelets
```

```
Available Book(s):
```

```
    WaveTour - ``WaveLet Tour of Signal Processing''
```

Lets construct a filter of compactly supported Daubechies wavelets

```
In [2]: h = MakeONFilter('Daubechies',4);
```

```
In [3]: % Lets define the size of the signal
N=4*64;
```

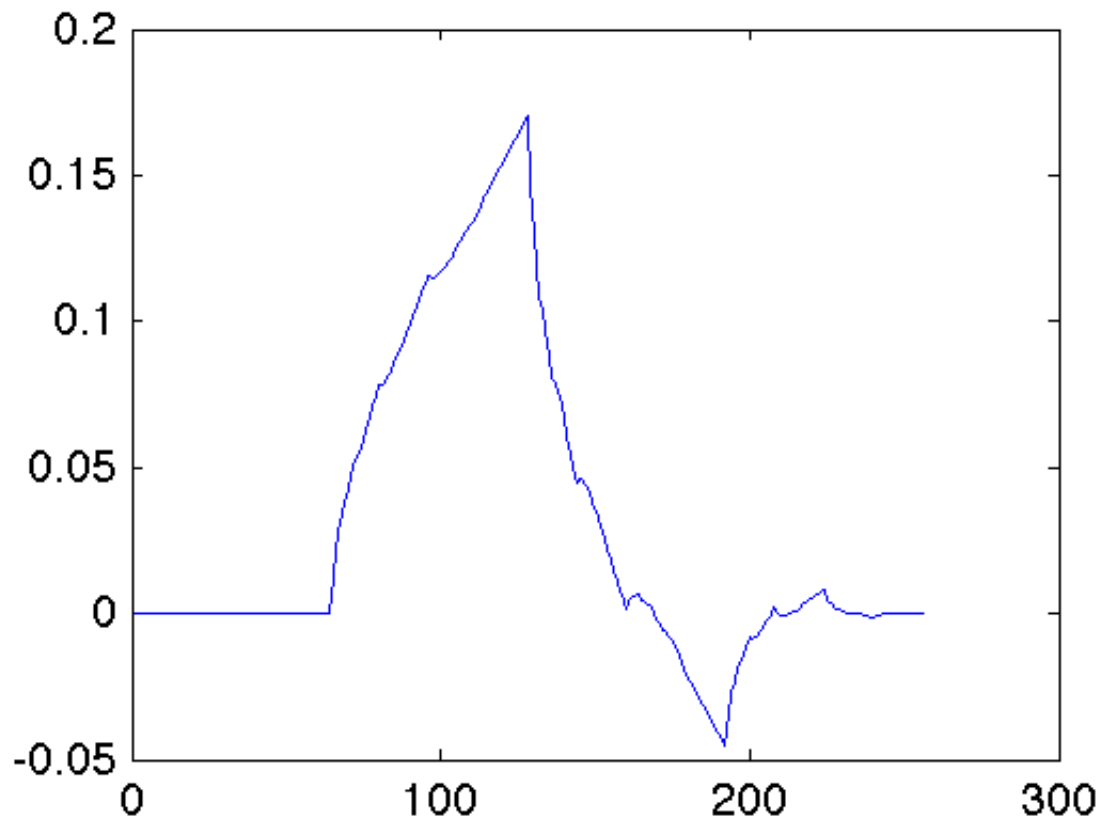
Now let's define  $e_i$ , as the canonical vector of length N with 1 in i

```
In [4]: i=2;
e_i=zeros(N,1)';
e_i(i)=1.0;
```

Lets check how the scaling function and wavelet function looks like

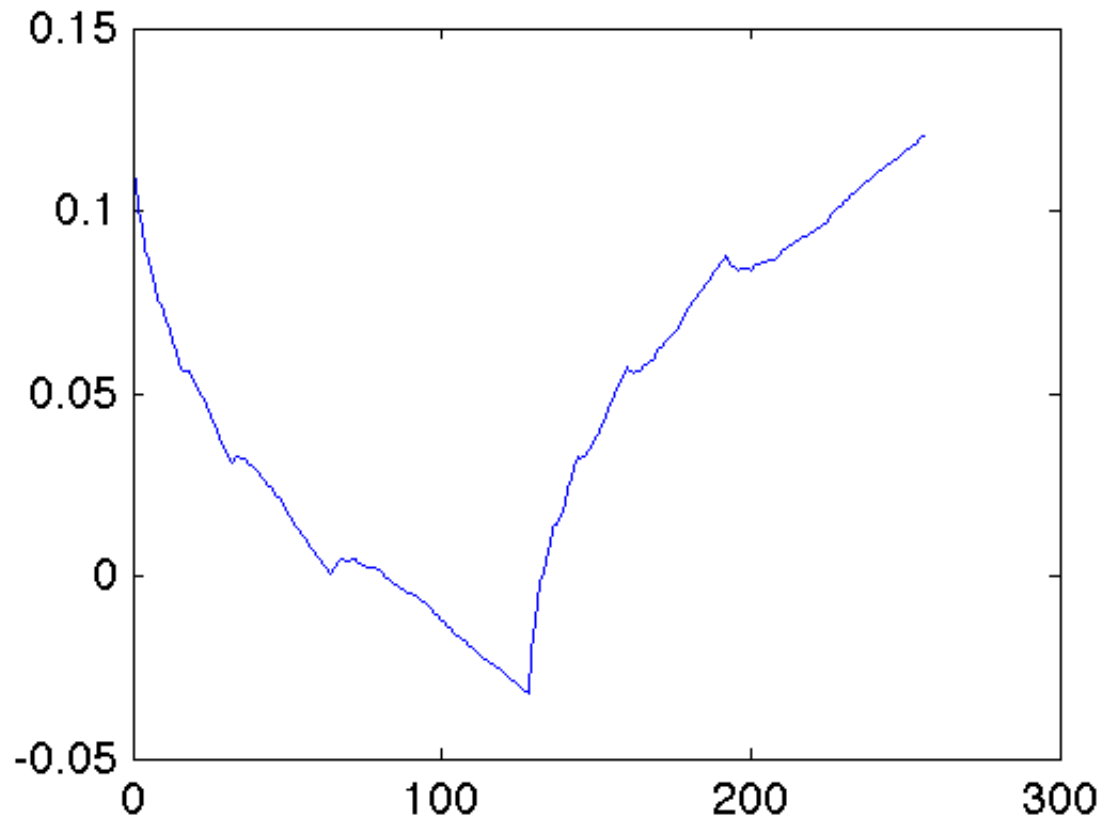
Scaling function:

```
In [5]: plot(IWT_PO(e_i,2,h));
```



Wavelet function:

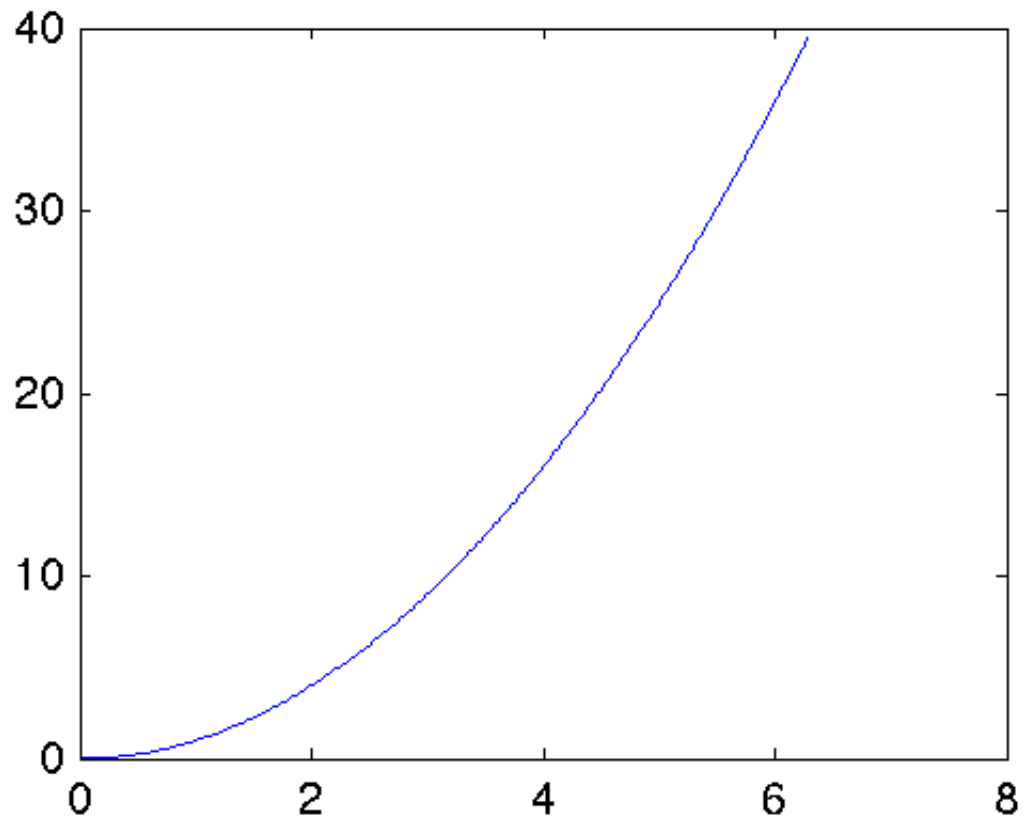
```
In [6]: plot(IWT_PO(e_i,1,h));
```



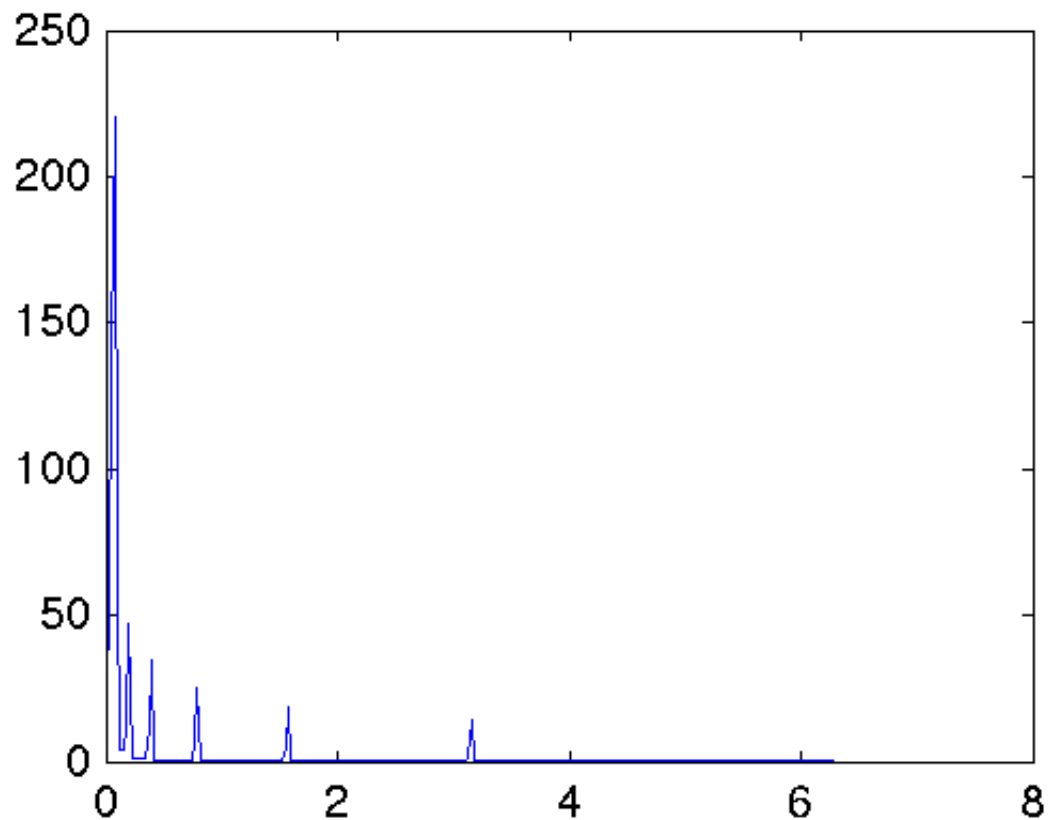
Lets define a smooth function  $f$  and plot its wavelet coefficients.

```
In [7]: x = linspace(0,2*pi,N);  
f = x.^2;
```

```
In [8]: plot(x,f);
```



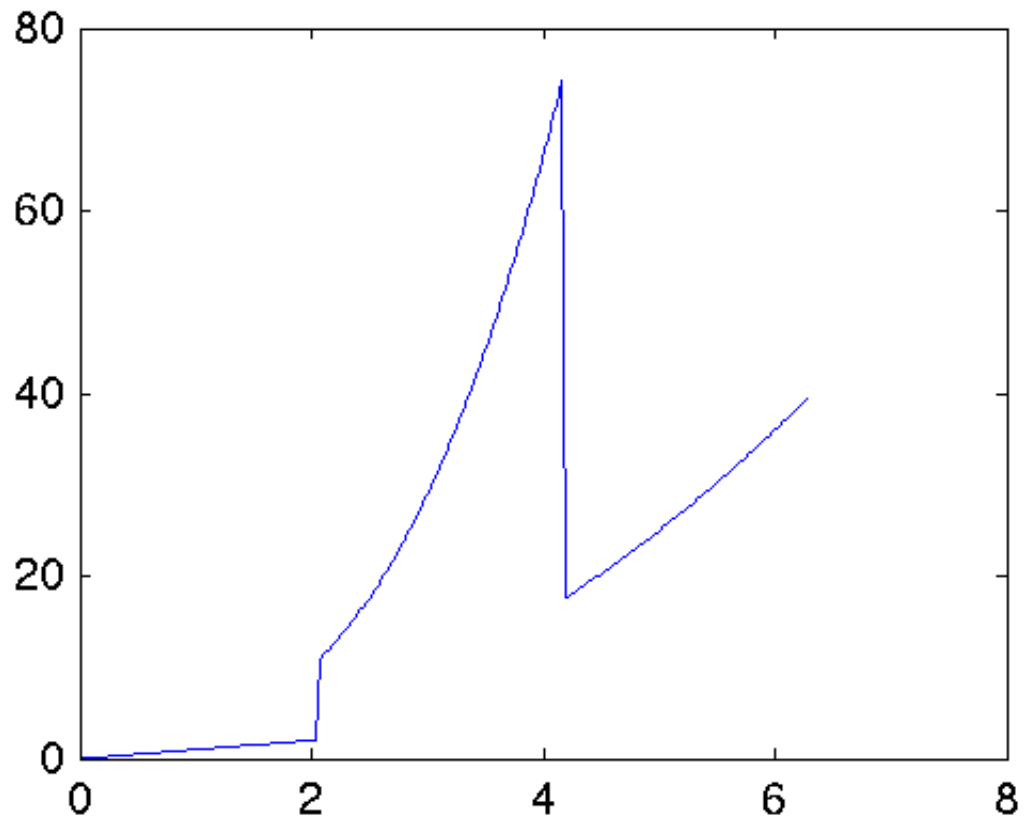
```
In [40]: fwtf=FWT_PO(f,2,h);  
plot(x,abs(fwtf))
```



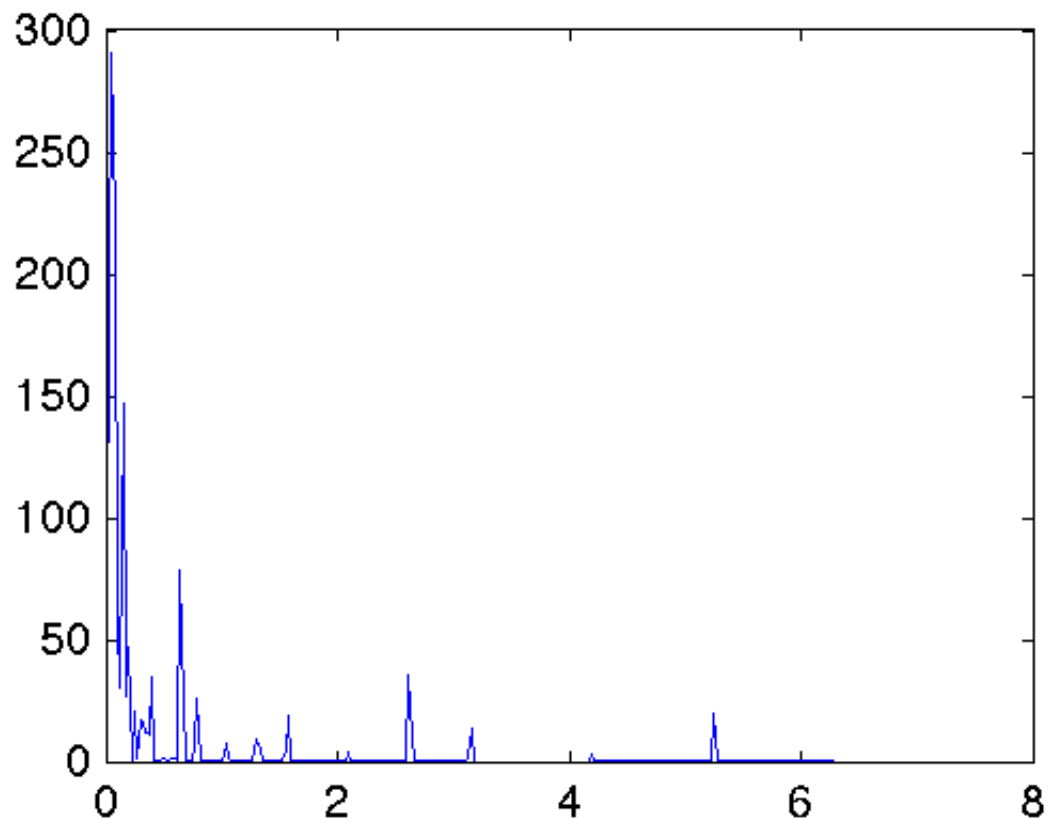
Lets define a smooth function  $g$  and plot its wavelet coefficients.

```
In [10]: g=zeros(1,N);  
for i=1:int64(N/3)  
    g(i)=x(i);  
end  
for i=int64(N/3):int64(2*N/3)  
    g(i)=x(int64(N/3))+x(i)^3;  
end  
for i=int64(2*N/3):N  
    g(i)=x(i)^2;  
end
```

```
In [11]: plot(x,g)
```



```
In [30]: fwtg=FWT_PO(g,2,h);  
plot(x,abs(fwtg))
```



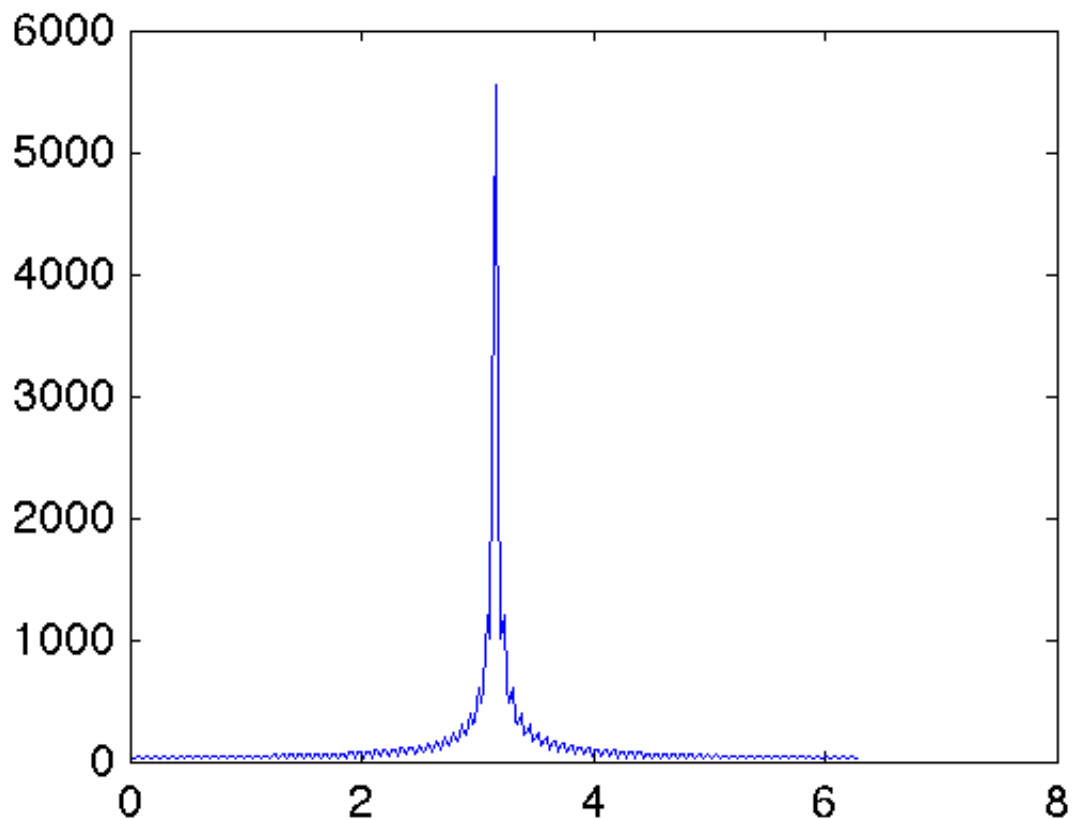
The wavelet coefficients in the smooth function decay rapidly and the wavelet coefficients in the non-smooth function have important coefficients even for  $N$  big (does not decay rapidly), regular functions are more compressible.

The signal  $g$  is piecewise, let's recover the signals, with wavelet coefficients and Fourier coefficients, setting to zero the coefficients with value less than 10% of the largest coefficient.

Let's compute first, the FFT coefficients



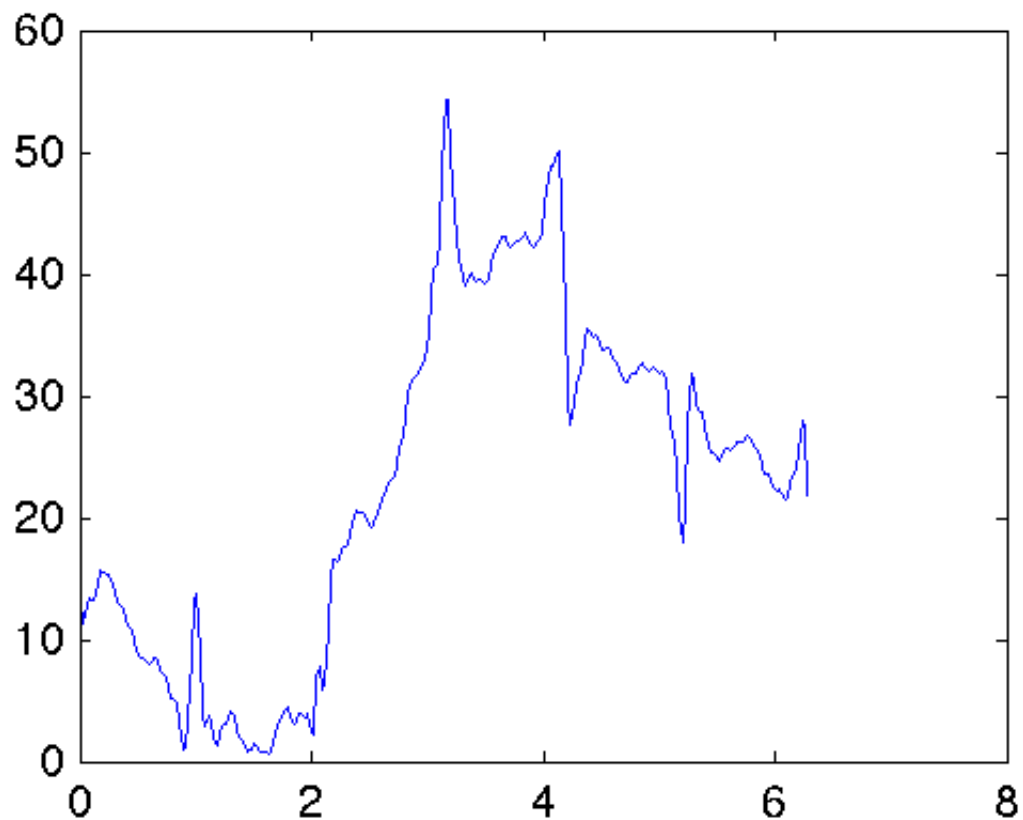
```
In [31]: fftg=ifftshift(fft2(fftshift(g)));  
plot(x,abs(fftg))
```



Lets set to zero the values less than the 10% of the maximum value

```
In [32]: threshold=max(fftg)*0.01;  
for i=1:N  
    if fftg(i)<=threshold  
        fftg(i)=0;  
    end  
end
```

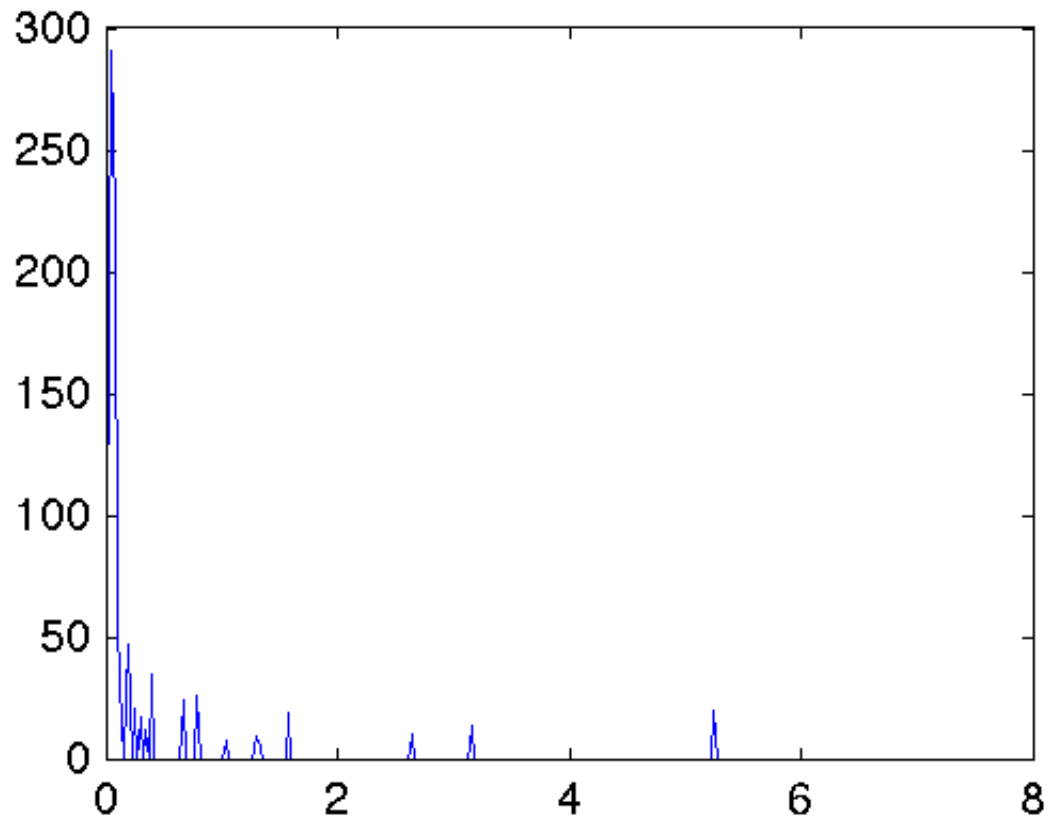
```
In [33]: % Recovery of g with fft  
plot(x,abs(fftshift(iff2(iff2shift(fftg))))))
```



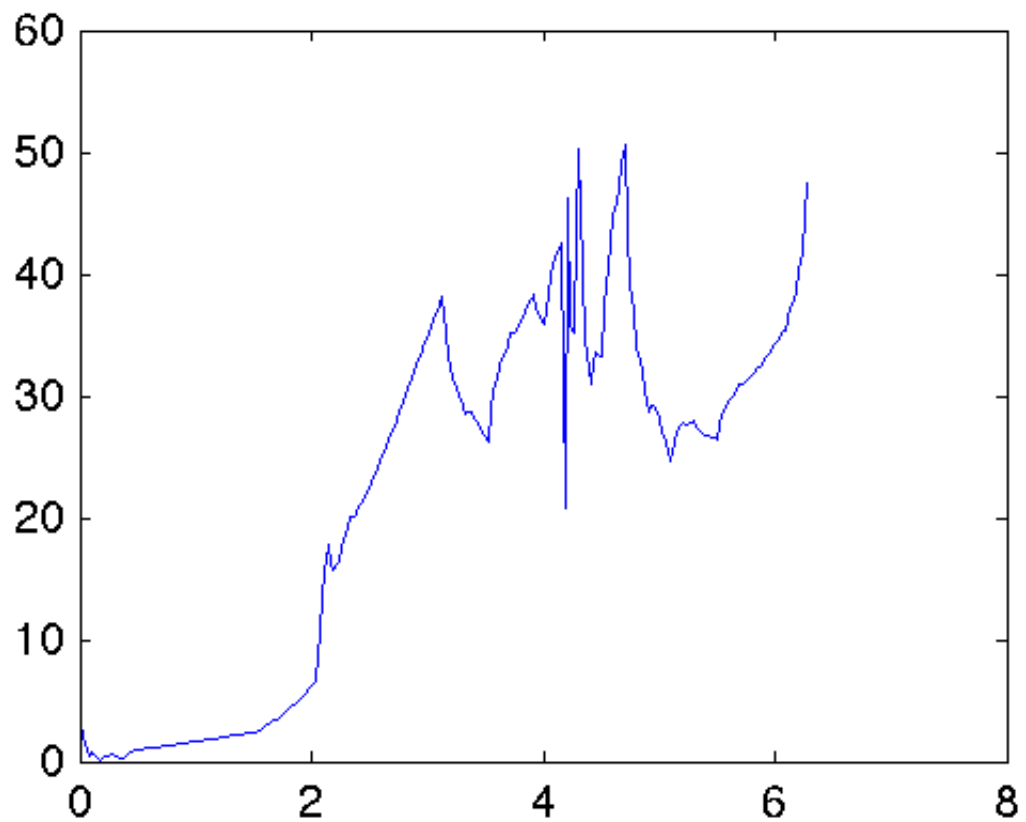
Now with wavelets

```
In [34]: threshold=max(fwtg)*0.01;  
for i=1:N  
    if fwtg(i)<=threshold  
        fwtg(i)=0;  
    end  
end
```

```
In [35]: plot(x,fwtg)
```



```
In [41]: % Recovery of g with wavelets  
plot(x,abs(IWT_PO(fwtg,2,h)))
```



The recovery via wavelets is more accurate