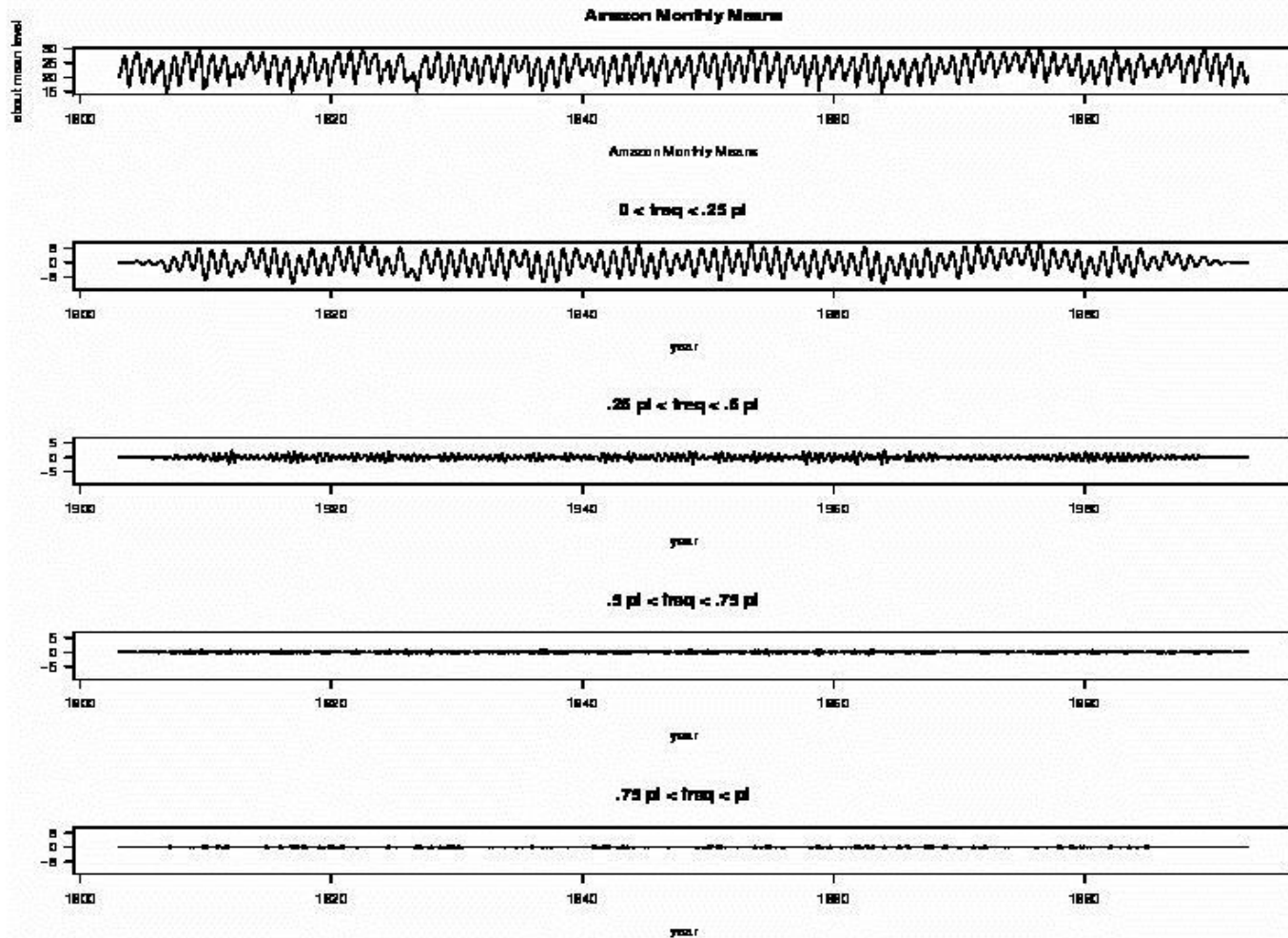


A bank of band-pass filters



```

par(mfrow=c(5,1))
junk<-scan("2003/bymonths"); taxis<-1903+c(1:1080)/12
plot(taxis,junk,main="Amazon Monthly Means",xlab="Amazon Monthly
Means",ylab="about mean
level",type="l",las=1,ylim=c(min(junk),max(junk)))
junkl<-length(junk); junk<-junk-mean(junk)
junk1<-c(spec.taper(junk),rep(0,2048-junkl))
junk2<-fft(junk1); OMEGA<-.25*pi
junkf<-2*pi*c(0:2047)/2048; junk3<-junk2
junk3[junkf>OMEGA&junkf<2*pi-OMEGA]<-0
junk4<-fft(junk3,inv=T)/2048
plot(taxis,Re(junk4[1:junkl]),main="0 < freq <
.25pi",xlab="year",ylab="",type="l",las=1,ylim=c(min(junk),max(junk)))
junk3<-junk2
junk3[junkf>2*OMEGA&junkf<2*pi-2*OMEGA]<-0
junk3[junkf<OMEGA|junkf>2*pi-OMEGA]<-0
junk4<-fft(junk3,inv=T)/2048

```

CONTINUES

```
plot(taxis,Re(junk4[1:junkl]),main=".25 pi < freq  
pi",xlab="year",ylab="",type="l",las=1,ylim=c(min(junk),max(junk)))  
junk3<-junk2  
junk3[junkf>3*OMEGA&junkf<2*pi-3*OMEGA]<-0  
junk3[junkf<2*OMEGA|junkf>2*pi-2*OMEGA]<-0  
junk4<-fft(junk3,inv=T)/2048  
plot(taxis,Re(junk4[1:junkl]),main=".5 pi < freq <  
pi",xlab="year",ylab="",type="l",las=1,ylim=c(min(junk),max(junk)))  
junk3<-junk2  
junk3[junkf<3*OMEGA|junkf>2*pi-3*OMEGA]<-0  
junk4<-fft(junk3,inv=T)/2048  
plot(taxis,Re(junk4[1:junkl]),main=".75 pi < freq  
pi",xlab="year",ylab="",type="l",las=1,ylim=c(min(junk),max(junk)))
```

Complex demodulation

In complex demodulation we first form the pair of real-valued series

$$\begin{aligned} Y_1(t) &= \cos \lambda_0 t X(t) \\ Y_2(t) &= \sin \lambda_0 t X(t) \end{aligned} \quad (2.7.31)$$

for $t = 0, \pm 1, \dots$ and then the pair of series

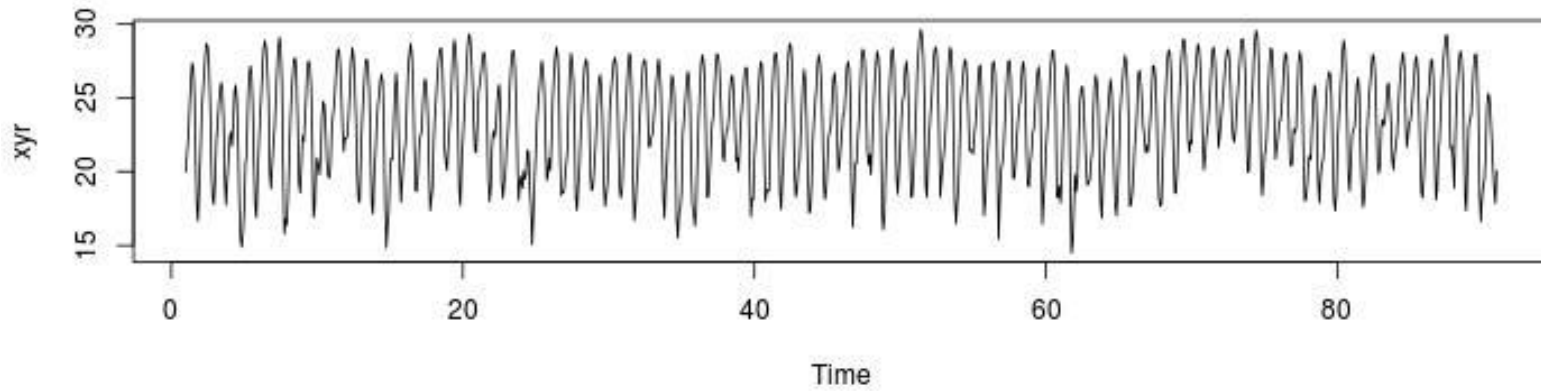
$$\begin{aligned} W_1(t) &= \sum_u a(t-u) Y_1(u) \\ W_2(t) &= \sum_u a(t-u) Y_2(u) \end{aligned} \quad (2.7.32)$$

where $\{a(t)\}$ is a low-pass filter. The series, $W_1(t), W_2(t) - \infty < t < \infty$, are called the **complex demodulates** of the series $X(t)$, $-\infty < t < \infty$. Because $\{a(t)\}$ is a low-pass filter, they will typically be substantially smoother than the series $X(t)$, $-\infty < t < \infty$. If we further form the series

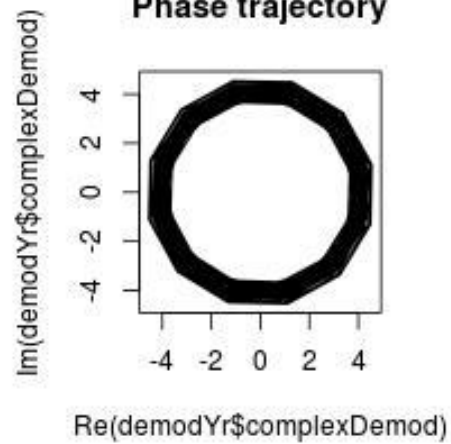
$$\begin{aligned} V_1(t) &= \cos \lambda_0 t W_1(t) + \sin \lambda_0 t W_2(t) \\ V_2(t) &= \sin \lambda_0 t W_1(t) - \cos \lambda_0 t W_2(t) \end{aligned} \quad (2.7.33)$$

for $-\infty < t < \infty$, then the following lemma shows that the series $V_1(t)$ is essentially a band-pass filtered version of the series $X(t)$, while the series $V_2(t)$ is essentially a band-pass filtered version of the series $X^H(t)$.

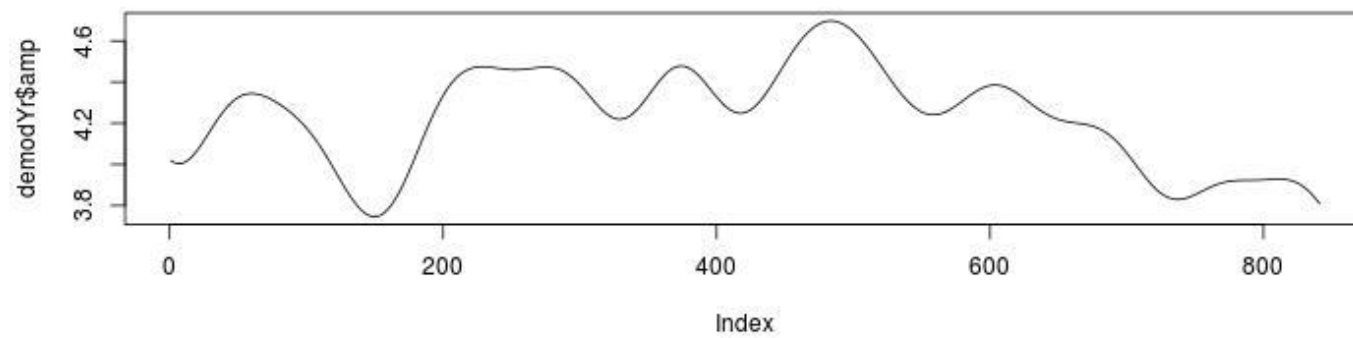
Monthly Manaus stage



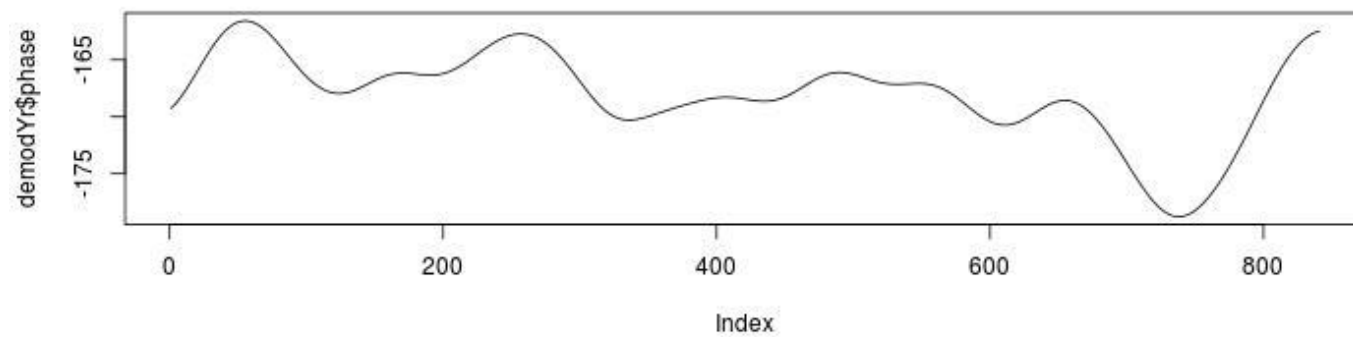
Phase trajectory



amplitude



phase



The spectrogram/dynamic spectrum analysis.

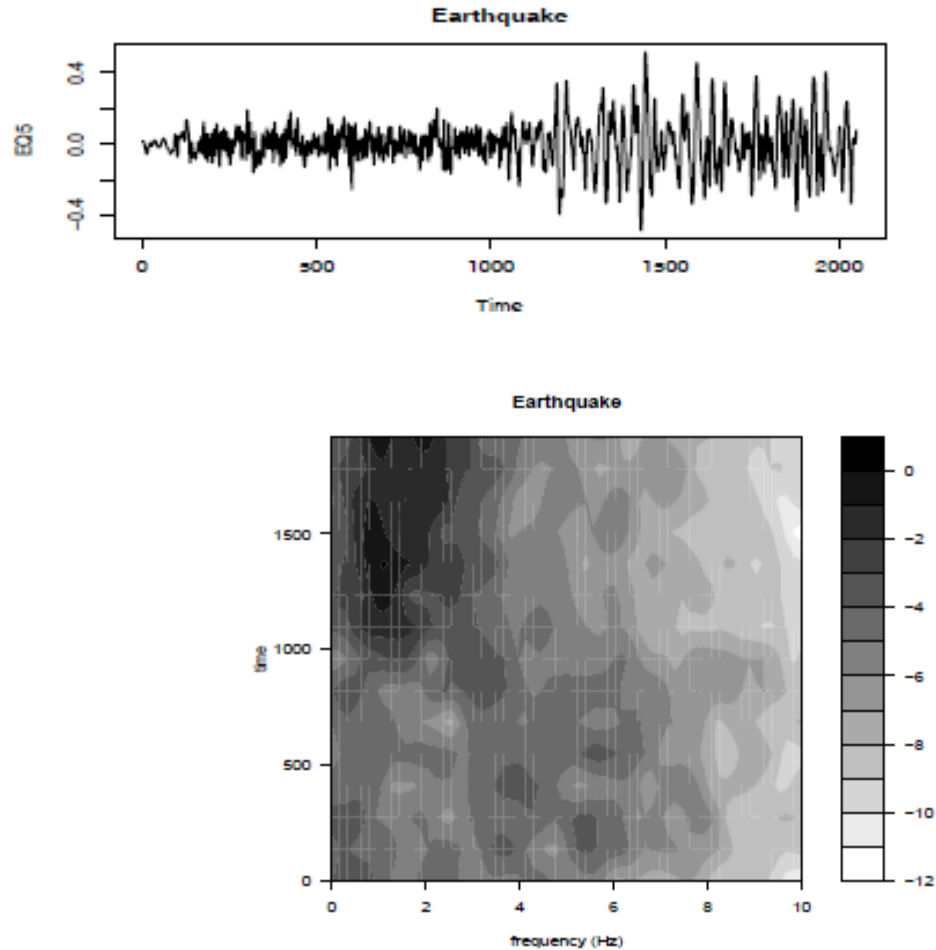


Fig. 4.17. Time-frequency image for the dynamic Fourier analysis of the earthquake series shown in [Figure 1.7](#).

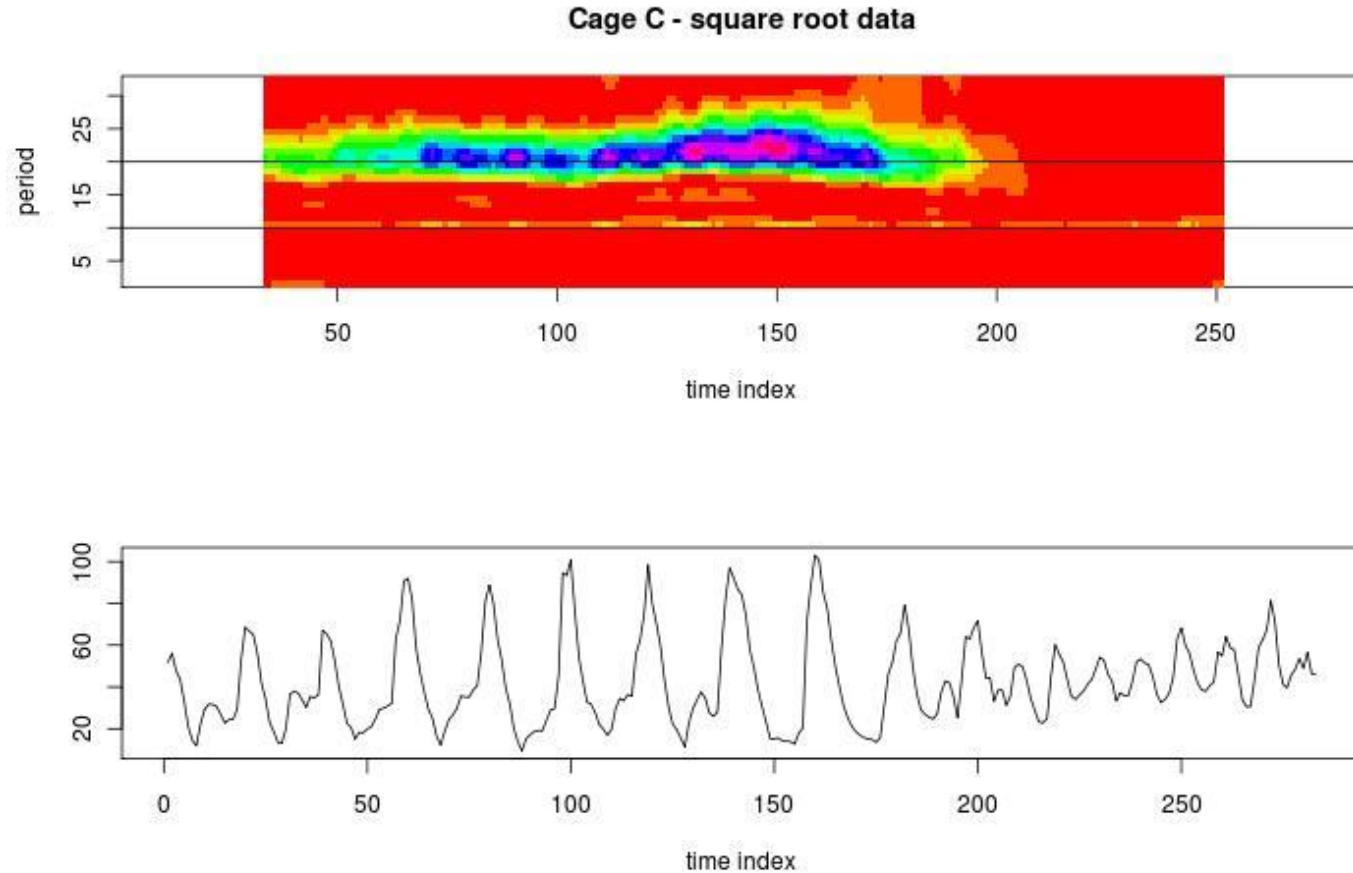
```

1 nobs = length(EXP6) # number of observations
2 wsize = 256 # window size
3 overlap = 128 # overlap
4 ovr = wsize-overlap
5 nseg = floor(nobs/ovr)-1; # number of segments
6 krnl = kernel("daniell", c(1,1)) # kernel
7 ex.spec = matrix(0, wsize/2, nseg)
8 for (k in 1:nseg) {
9   a = ovr*(k-1)+1
10  b = wsize+ovr*(k-1)
11  ex.spec[,k] = spectrum(EXP6[a:b], krnl, taper=.5, plot=F)$spec }
12 x = seq(0, 10, len = nrow(ex.spec)/2)
13 y = seq(0, ovr*nseg, len = ncol(ex.spec))
14 z = ex.spec[1:(nrow(ex.spec)/2),]
15 filled.contour(x , y, log(z), ylab="time",xlab="frequency (Hz)",
16               nlevels=12, col=gray(11:0/11), main="Explosion")
17 persp(x, y, z, zlab="Power",xlab="frequency (Hz)",ylab="time",
18       ticktype="detailed", theta=25,d=2, main="Explosion") # not shown

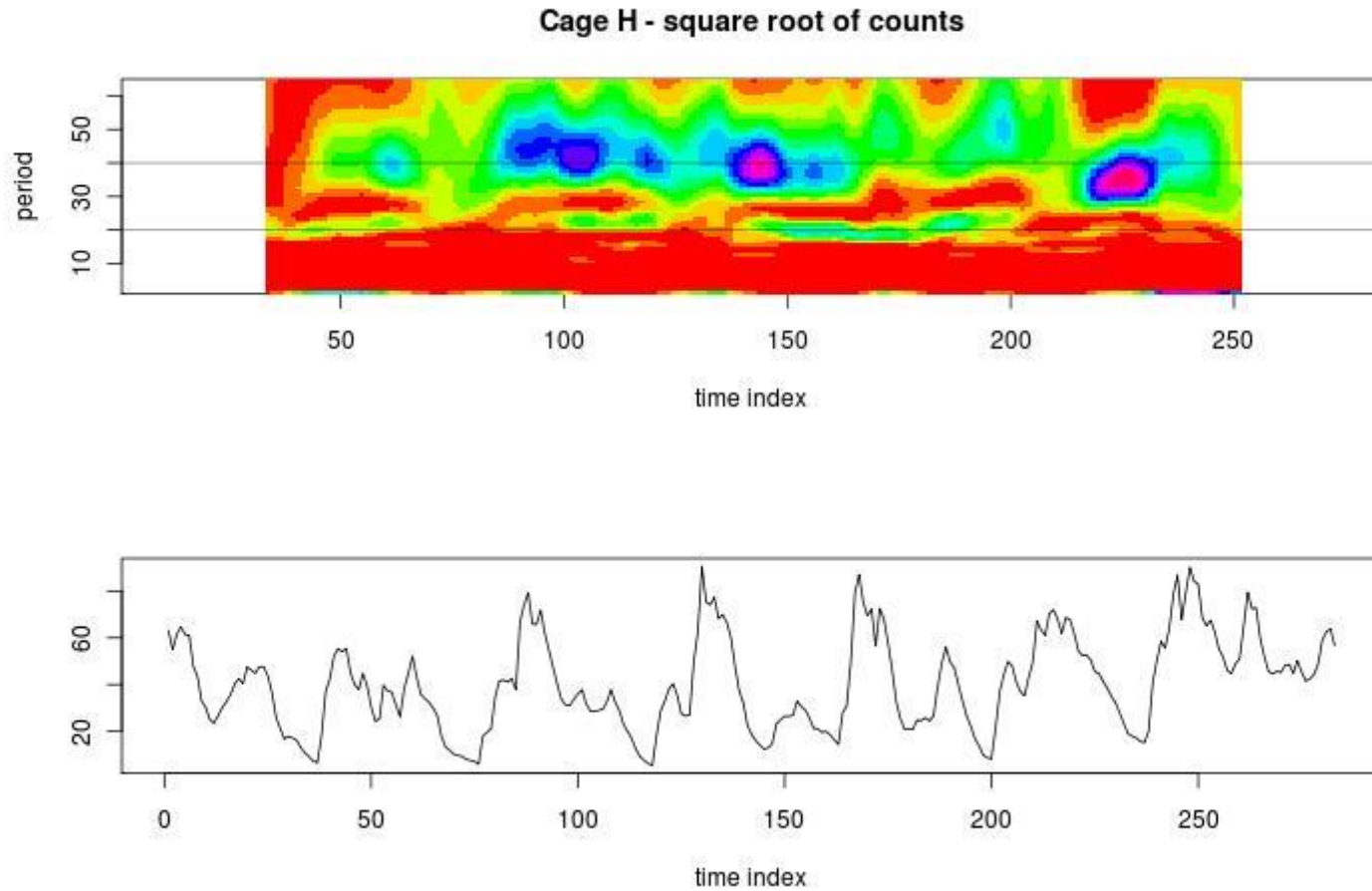
```

Nicholson's blowfly experiments

Cage C. Known period 20 (Near natural period) lines at 10,20



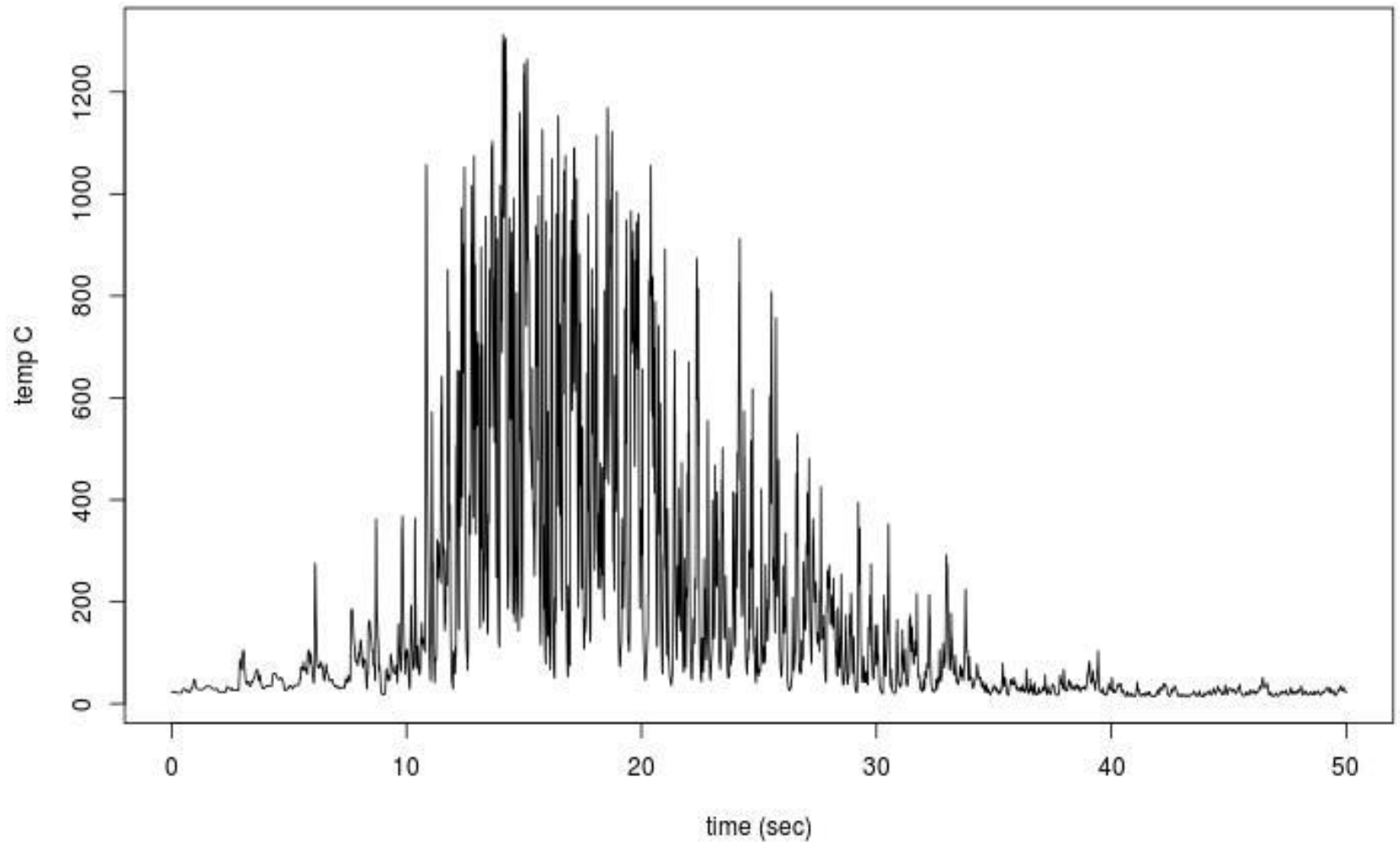
Cage H. Known period 40 Square root {counts}



USFS experimental fire



Experiment 13 thermocouple 1 temperatures



Dynamic spectrum

