## Peru Lima Temperature

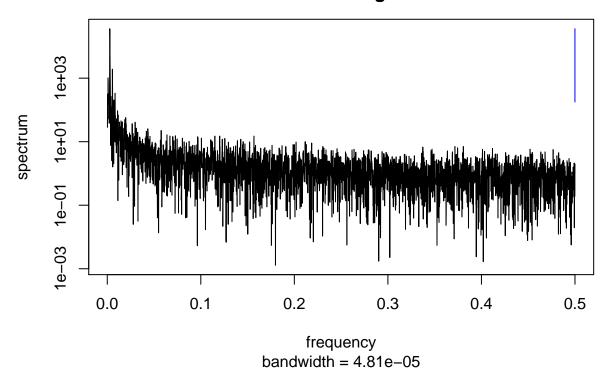
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```
setwd('C:/Users/zmxj/Documents/stocasticProject')
data<-read.csv("Peru3.csv")
y<-data[,3]
y<-as.numeric(as.character(y))
t<-1:length(y)
data$date<-as.Date(data$date,"%m/%d/%Y")
date<-data$date</pre>
```

#### Linear fitting for Peru [1]

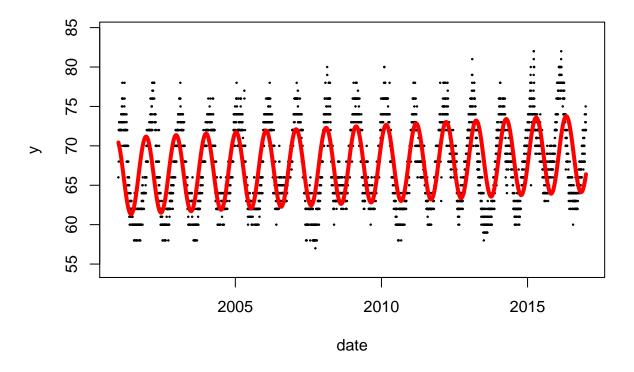
```
ssp <- spectrum(y)</pre>
```

# Series: x Raw Periodogram



```
per <- 1/ssp$freq[ssp$spec==max(ssp$spec)]
reslm <- lm(y ~ sin(2*pi/per*t)+cos(2*pi/per*t))
summary(reslm)</pre>
```

```
##
## Call:
## lm(formula = y \sim sin(2 * pi/per * t) + cos(2 * pi/per * t))
## Residuals:
                                     3Q
##
        Min
                  1Q
                      Median
                                             Max
## -11.2680 -2.5709 -0.0755
                                 2.3899 12.1618
##
## Coefficients:
##
                        Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                        67.57438
                                    0.04907 1377.06
                                                       <2e-16 ***
## sin(2 * pi/per * t) -2.23322
                                    0.06955
                                             -32.11
                                                       <2e-16 ***
## cos(2 * pi/per * t) 4.31729
                                    0.06922
                                              62.37
                                                       <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.75 on 5841 degrees of freedom
## Multiple R-squared: 0.4567, Adjusted R-squared: 0.4565
## F-statistic: 2455 on 2 and 5841 DF, p-value: < 2.2e-16
rg <- diff(range(y))
plot(y \sim \text{date}, y = \text{lim} = c(\min(y) - 0.1 \times rg, \max(y) + 0.1 \times rg), cex = 0.2)
# including 2nd harmonic really improves the fit
reslm2 \leftarrow lm(y \sim t + sin(2*pi/per*t) + cos(2*pi/per*t)) # + sin(4*pi/per*t) + cos(4*pi/per*t))
summary(reslm2)
##
## Call:
## lm(formula = y \sim t + sin(2 * pi/per * t) + cos(2 * pi/per * t))
##
## Residuals:
##
        Min
                  1Q
                       Median
                                     3Q
## -10.7169 -2.5709
                        0.0368
                                 2.4029 12.8530
##
## Coefficients:
##
                          Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                         6.613e+01 9.568e-02 691.13
                                                         <2e-16 ***
                         4.943e-04 2.835e-05
                                                17.43
## t
                                                         <2e-16 ***
## sin(2 * pi/per * t) -2.229e+00 6.781e-02 -32.87
                                                         <2e-16 ***
## cos(2 * pi/per * t) 4.333e+00 6.750e-02
                                                64.20
                                                         <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.656 on 5840 degrees of freedom
## Multiple R-squared: 0.4836, Adjusted R-squared: 0.4833
## F-statistic: 1823 on 3 and 5840 DF, p-value: < 2.2e-16
```

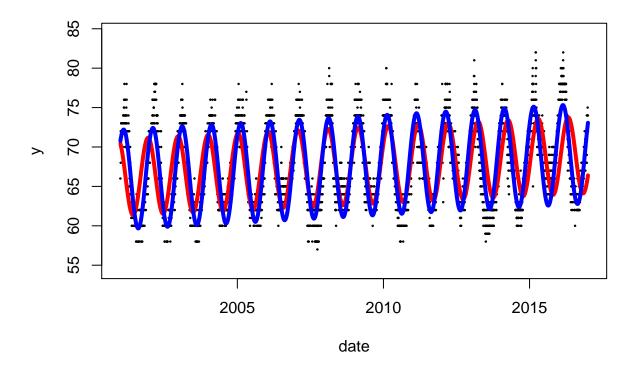


#### # Nonlinear fitting

```
\label{eq:weights} $W = 2*pi/per$$ $\operatorname{reslm_1} <-\operatorname{nls}(y \sim \operatorname{cons} + A*t + B*sin(W*t) + C*cos(W*t), start = 1 ist(\operatorname{cons} = 66.13, A = 0.0004943, B = -2.229, W = W, C = 4 summary(\operatorname{reslm_1})$$ $\operatorname{summary}(\operatorname{reslm_1})$$
```

```
##
## Formula: y \sim cons + A * t + B * sin(W * t) + C * cos(W * t)
##
## Parameters:
        Estimate Std. Error t value Pr(>|t|)
## cons 6.587e+01 6.015e-02 1095.05
                                      <2e-16 ***
## A
       5.654e-04 1.783e-05
                               31.72
                                      <2e-16 ***
## B
        4.088e+00 7.089e-02
                               57.67
                                      <2e-16 ***
        1.717e-02 3.986e-06 4307.08
                                      <2e-16 ***
## C
        4.823e+00 6.419e-02
                              75.14
                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.296 on 5839 degrees of freedom
## Number of iterations to convergence: 5
## Achieved convergence tolerance: 3.867e-06
```

```
plot(y~date,ylim=c(min(y)-0.1*rg,max(y)+0.1*rg),cex=0.2)
lines(fitted(reslm2)~date,col='red',lwd=4)
lines(fitted(reslm_1)~date,col='blue',lwd=4)
```



```
fitmean2 <- fitted(reslm_1)</pre>
```

#### Sigma

```
sigma1=sd(diff(y)) #using difference of y, the one on the paper
sigma2=sd(y-fitted(reslm2)) #approximate residual standard error
```

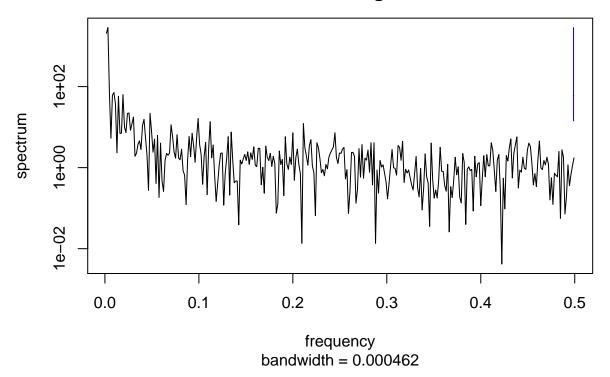
#### Setting the trigger

```
#El Nino
for (i in 366:(length(y)-90)){
diff=mean(y[i:(i+30)])-mean(y[(i-365):(i-365+30)])
diff2=mean(y[(i+30):(i+60)])-mean(y[(i+30-365):(i-365+60)])
diff3=mean(y[(i+60):(i+90)])-mean(y[(i+60-365):(i-365+90)])
if (diff>0.1){
```

#### Linear Modeling for El Nino[2]

```
y2<-y[4906:5515]
t2<-4906:5515
date2<-data$date[4906:5515]
ssp <- spectrum(y2)
```

### Series: x Raw Periodogram



```
per <- 1/ssp$freq[ssp$spec==max(ssp$spec)]

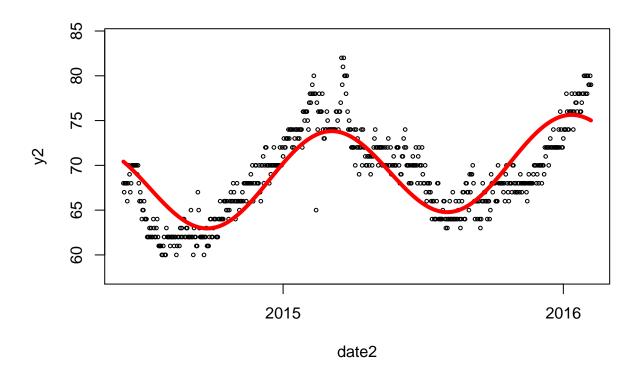
rg <- diff(range(y2))
plot(y2~date2,ylim=c(min(y2)-0.1*rg,max(y2)+0.1*rg),cex=0.5)

# including 2nd harmonic really improves the fit</pre>
```

```
 reslm4 \leftarrow lm(y2 \sim t2 + sin(2*pi/per*t2) + cos(2*pi/per*t2)) \# + sin(4*pi/per*t) + cos(4*pi/per*t)) summary(reslm4)
```

```
##
## Call:
## lm(formula = y2 \sim t2 + sin(2 * pi/per * t2) + cos(2 * pi/per *
##
## Residuals:
##
              1Q Median
                               3Q
      Min
                                     Max
## -8.4101 -1.9604 0.1918 1.9867 8.4258
##
## Coefficients:
                         Estimate Std. Error t value Pr(>|t|)
##
                       38.7837446 3.6248827 10.699 < 2e-16 ***
## (Intercept)
                       0.0058093 0.0006954 8.354 4.49e-16 ***
## sin(2 * pi/per * t2) -1.7693087 0.1630322 -10.853 < 2e-16 ***
## cos(2 * pi/per * t2) -4.6415275 0.1689548 -27.472 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.772 on 606 degrees of freedom
## Multiple R-squared: 0.6668, Adjusted R-squared: 0.6651
## F-statistic: 404.2 on 3 and 606 DF, p-value: < 2.2e-16
```

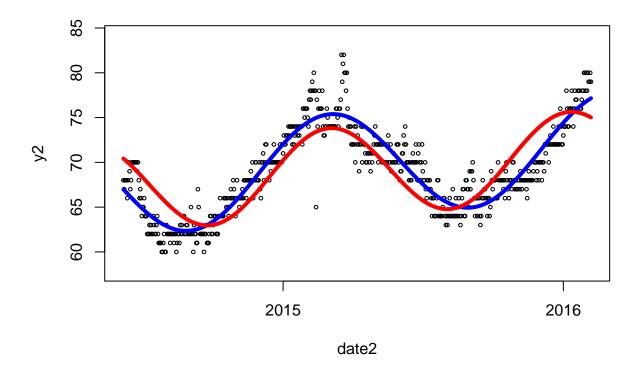
lines(fitted(reslm4)~date2,col="red",lwd=4)



# Nonlinear fit for El Nino

```
reslm_2 \leftarrow nls(y2 \sim cons+A*t2+B*sin(W*t2)+C*cos(W*t2), start=list(cons=38.7837446, A=0.0058093, B=-1.769, summary(reslm_2)
```

```
##
## Formula: y2 \sim cons + A * t2 + B * sin(W * t2) + C * cos(W * t2)
## Parameters:
         Estimate Std. Error t value Pr(>|t|)
## cons 3.306e+01 2.470e+00 13.386
                                       <2e-16 ***
## A
        7.045e-03
                  4.748e-04
                              14.838
                                       <2e-16 ***
        3.438e+00
                   2.670e+00
                               1.288
                                       0.1982
## B
## W
        1.712e-02
                   1.080e-04 158.466
                                        <2e-16 ***
                                       0.0145 *
## C
        4.741e+00
                   1.934e+00
                               2.452
##
## Signif. codes:
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.89 on 605 degrees of freedom
##
## Number of iterations to convergence: 18
## Achieved convergence tolerance: 2.764e-06
plot(y2~date2,ylim=c(min(y2)-0.1*rg,max(y2)+0.1*rg),cex=0.5)
lines(fitted(reslm_2)~date2,col='blue',lwd=4)
lines(fitted(reslm4)~date2,col="red",lwd=4)
```



```
fitmean2 <- fitted(reslm_2)</pre>
```

### Sigma

```
sigma3=sd(diff(y2)) #using difference of y, the one on the paper
sigma4=sd(y2-fitted(reslm_2)) #approximate residual standard error
```

reference: [1][2] R code, http://stats.stackexchange.com/questions/60994/fit-a-sinusoidal-term-to-data, April 7, 2017