# 20july2020

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# 1. progress summary

This week, as we discussed from last meeting, I have: - 1. selected the 1st complete sin curve of every node and align them to register at the origin. - 2. Scale the the time frame of each node to [-1,1] while time=0 remain unchanged. I also make sure that each node have the same number of rows so that fPCA will be able to apply. - 3. Apply fPCA to the new dataset of curves (1st cycle of each node), and the result is that the first two Principle Component functions covers 91.5% of total variance. - 4. repeat the process above for 2nd cycle of each node, and first two Principle component functions covered in total 93.7% of total variance.

# 2. read data and attach packages

# 3. Defined fourier smoothing functions

To study a single brain node response, specify the node number in the node\_subset list.

```
f_fourier_smooth <- function(time_subset, data_mat, node_subset, k) {
  basis <- create.fourier.basis(c(time_subset[1],time_subset[length(time_subset)]), k)
  fd_obj <- smooth.basis(time_subset, data_mat[time_subset, node_subset], basis)
  smoothfd <- fd_obj$fd
  #plot(smoothfd)
  #title(main=paste("Fourier Basis Smoothing of node:", node_subset, ", Basis_number:",k
))
  return(fd_obj)
}</pre>
```

# 4. define the function to extract periodic cycle of a single node response

```
transform.Cycle = function(data, register){
  # obtain index at which curve crosses 0
 #crossed 0---> -1: pos to neg,
                                     1: neg to pos
 #returns: location index where curve crosses X-axis
 x=diff(ifelse(data>0,1,0))
 z_idx=(1:599)[x!=0]
 # skip first crossing if it is from positive to negative
 if (x[z_idx[1]]==-1){
    z_idx=z_idx[-1]
 }
 #put every complete cycle in a Dataframe
 i=1
 c1=1
 result=data.frame(cycle=integer(), time=integer(), y_value=integer())
 while (i+2<=length(z_idx)){</pre>
      if(register==0){
        tmp=data.frame(cycle=cl, time=seq(z_idx[i],z_idx[i+2]), y_value=smoothed_curve[z
_idx[i]:z_idx[i+2]])
      else{
        tmp=data.frame(cycle=cl, time=seq(1,length(seq(z idx[i],z idx[i+2]))), y value=s
moothed_curve[z_idx[i]:z_idx[i+2]])
        tmp[,2]=tmp[,2]-length(seq(z idx[i],z idx[i+1]))-1
      result=rbind(result,tmp)
      i=i+2
      c1=c1+1
 return(result)
}
```

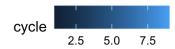
### 5. Oversmooth with basis functions

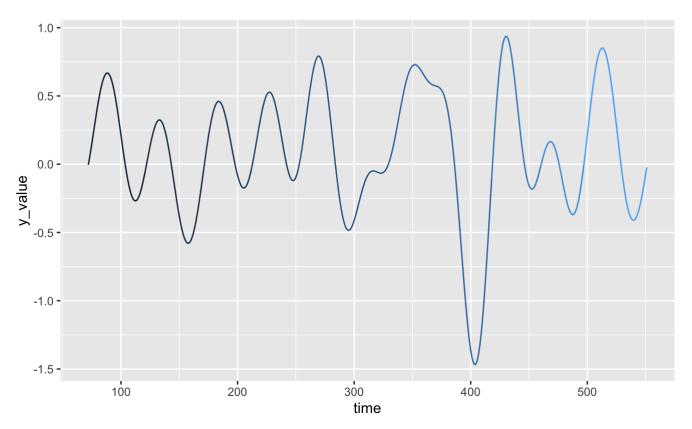
```
# OverSmoothed, k=32

## register: 0 --> plot data on the original timeline
## 1 --> register every complete sinusoidal curve starting at 0;

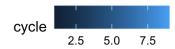
result_obj <- f_fourier_smooth(time_subset=c(1:600), data_mat, node_subset=c(1), k=32)
smoothed_curve = eval.fd(c(1:600), result_obj$fd)

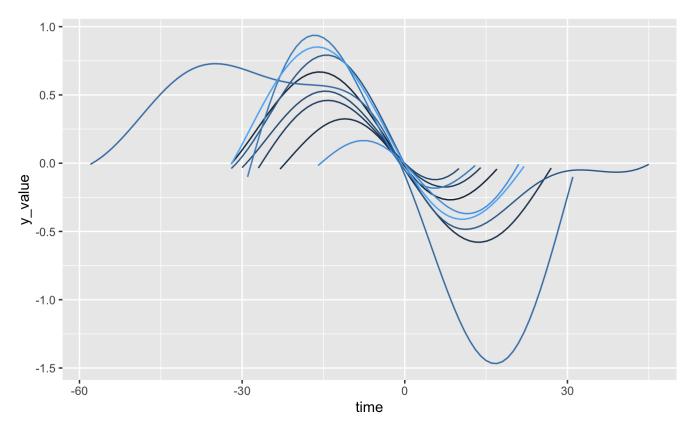
transformed_nodel = transform.Cycle(smoothed_curve, register=0)
ggplot(transformed_nodel, aes(time, y_value,group=cycle, colour=cycle)) + geom_line() +
theme(legend.position="top")</pre>
```





transformed\_node1 = transform.Cycle(smoothed\_curve, register=1)
ggplot(transformed\_node1, aes(time, y\_value,group=cycle, colour=cycle)) + geom\_line() +
theme(legend.position="top")

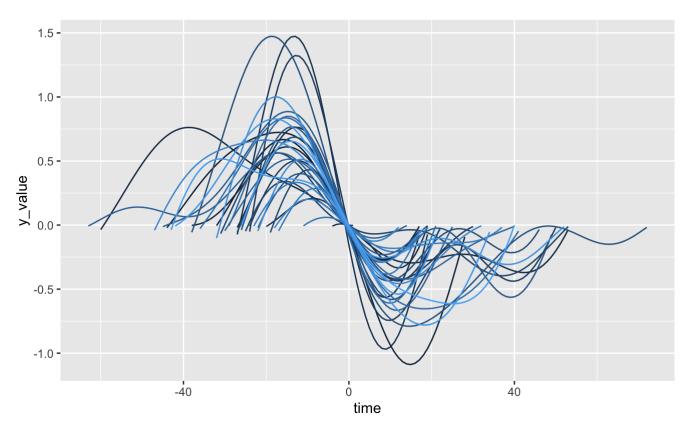




#### 5.1 1st cycle across different node

```
df = data.frame(node=integer(),cycle=integer(), time=integer(), y_value=integer())
for(i in 1:32){
    result_obj <- f_fourier_smooth(time_subset=c(1:600), data_mat, node_subset=c(i), k=32)
    smoothed_curve = eval.fd(c(1:600),result_obj$fd)
    transformed_node = transform.Cycle(smoothed_curve, register=1)
    tmp = subset(transformed_node, cycle==1)
    tmp$node=i
    df=rbind(df,tmp)
}
ggplot(df, aes(time, y_value,group=node, colour=node)) + geom_line() + theme(legend.posi
    tion="top")</pre>
```

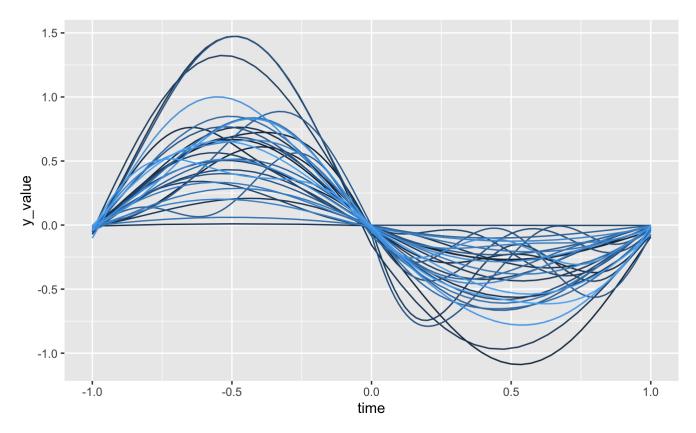




#### 5.2 Scale the timeframe of each node

```
df_tmp = data.frame(node=integer(),cycle=integer(), time=integer(), y_value=integer())
for(i in 1:32){
   tmp=subset(df, node==i)
   tmp_1=subset(tmp, time<=0)
   tmp_2=subset(tmp, time>0)
   tmp_1$time=(tmp_1$time) / (max(abs(tmp_1$time)))
   tmp_2$time=(tmp_2$time) / (max(abs(tmp_2$time)))
   tmp_rbind(tmp_1,tmp_2)
   df_tmp=rbind(df_tmp,tmp)
}
ggplot(df_tmp, aes(time, y_value,group=node, colour=node)) + geom_line() + theme(legend.position="top")
```



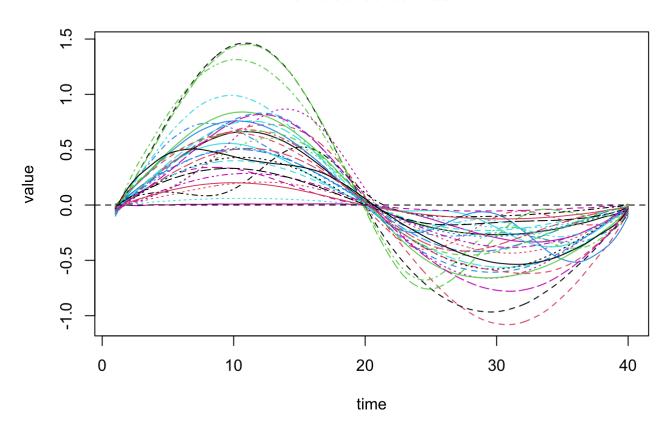


#### 5.3 pivot the new dataframe

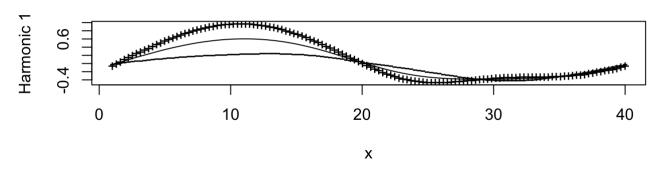
```
df_new = data.frame(matrix(nrow=40))
for(i in 1:32){
    xx=seq(-1,1,length.out=40)
    tmp=subset(df_tmp, node==i)
    s=smooth.spline(x=tmp$time, y=tmp$y_value, df = 10)
    df_new[,ncol(df_new)+1]=predict(s,xx)$y
}
df_new=df_new[,2:33]
oldnames = colnames(df_new)
newnames = colnames(data_mat)
for(i in 1:32) names(df_new)[names(df_new) == oldnames[i]] = newnames[i]
```

#### 5.4 fPCA on new dataframe

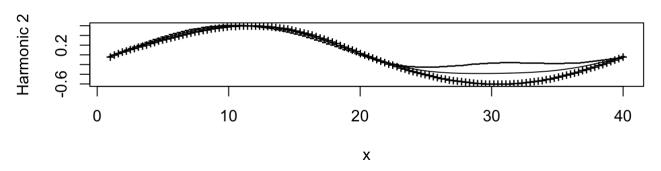
## smoothed curves



# PCA function 1 (Percentage of variability 62.5)



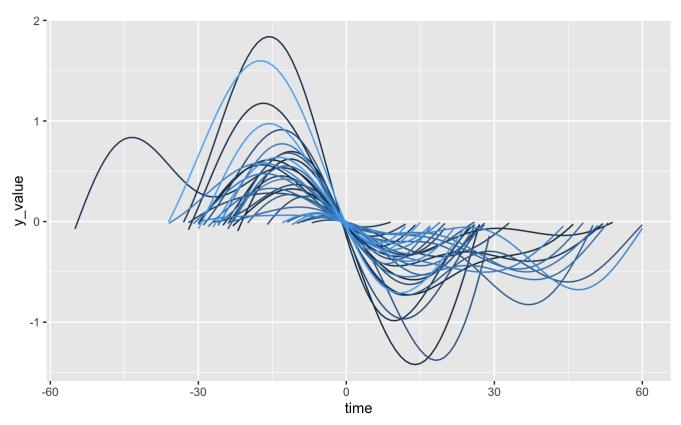
# PCA function 2 (Percentage of variability 29)



#### 6.1 2nd cycle across different node

```
df = data.frame(node=integer(),cycle=integer(), time=integer(), y_value=integer())
for(i in 1:32){
    result_obj <- f_fourier_smooth(time_subset=c(1:600), data_mat, node_subset=c(i), k=32)
    smoothed_curve = eval.fd(c(1:600),result_obj$fd)
    transformed_node = transform.Cycle(smoothed_curve, register=1)
    tmp = subset(transformed_node, cycle==2)
    tmp$node=i
    df=rbind(df,tmp)
}
ggplot(df, aes(time, y_value,group=node, colour=node)) + geom_line() + theme(legend.posi
    tion="top")</pre>
```

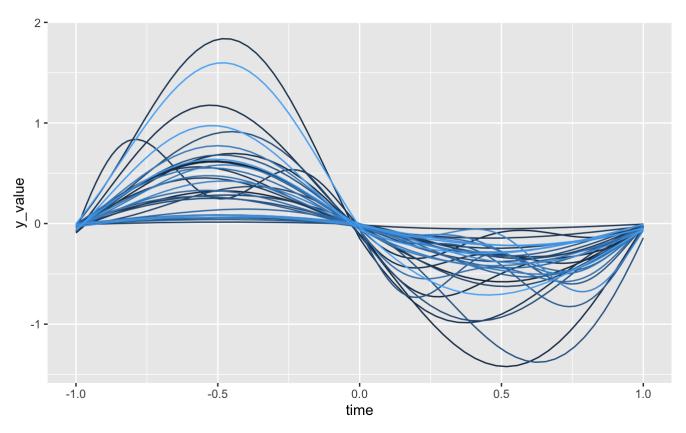




#### 6.2 Scale the timeframe of each node

```
df_tmp = data.frame(node=integer(),cycle=integer(), time=integer(), y_value=integer())
for(i in 1:32){
   tmp=subset(df, node==i)
   tmp_1=subset(tmp, time<=0)
   tmp_2=subset(tmp, time>0)
   tmp_1$time=(tmp_1$time) / (max(abs(tmp_1$time)))
   tmp_2$time=(tmp_2$time) / (max(abs(tmp_2$time)))
   tmp=rbind(tmp_1,tmp_2)
   df_tmp=rbind(df_tmp,tmp)
}
ggplot(df_tmp, aes(time, y_value,group=node, colour=node)) + geom_line() + theme(legend.position="top")
```



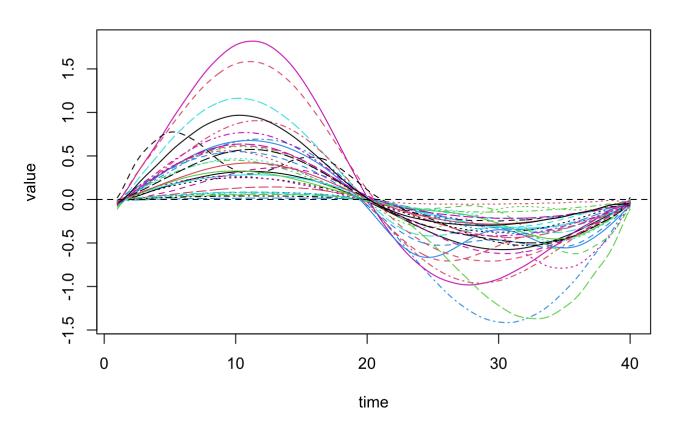


#### 6.3 pivot the new dataframe

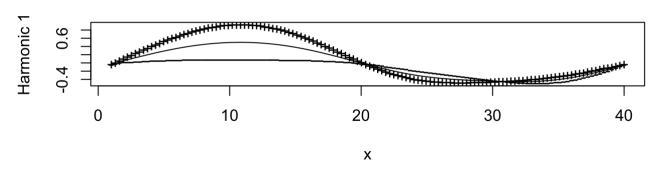
```
df_new = data.frame(matrix(nrow=40))
for(i in 1:32){
    xx=seq(-1,1,length.out=40)
    tmp=subset(df_tmp, node==i)
    s=smooth.spline(x=tmp$time, y=tmp$y_value, df = 10)
    df_new[,ncol(df_new)+1]=predict(s,xx)$y
}
df_new=df_new[,2:33]
oldnames = colnames(df_new)
newnames = colnames(data_mat)
for(i in 1:32) names(df_new)[names(df_new) == oldnames[i]] = newnames[i]
```

#### 6.4 fPCA on new dataframe

## smoothed curves



# PCA function 1 (Percentage of variability 61.9)



# PCA function 2 (Percentage of variability 31.8)

