24july2020

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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.

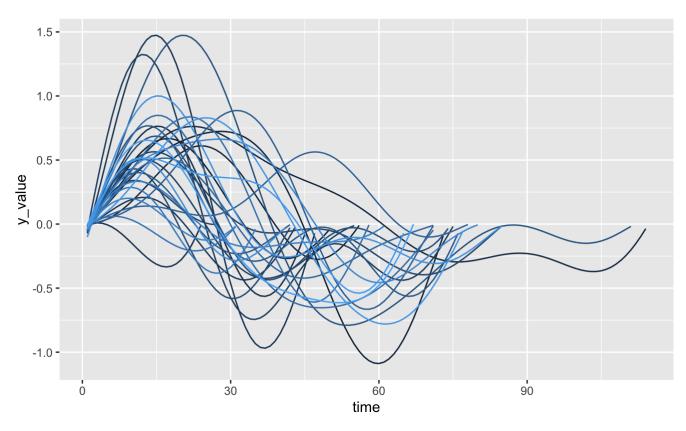
```
f_fourier_smooth <- function(time_subset, data_mat, node_subset, k) {
  basis <- create.fourier.basis(c(time_subset[1],time_subset[length(time_subset)]), k)
  fd_obj <- smooth.basis(time_subset, data_mat[time_subset, node_subset], basis)
  smoothfd <- fd_obj$fd
  #plot(smoothfd)
  #title(main=paste("Fourier 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

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.position="top")</pre>
```

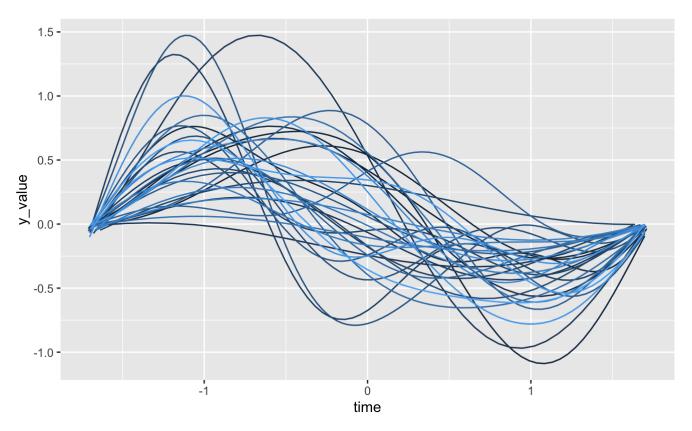




5.2 Scale the timeframe of each node

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

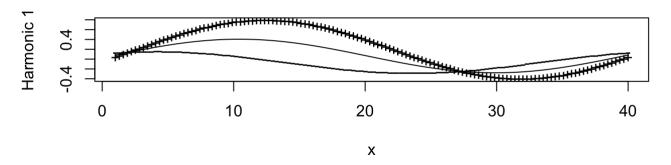




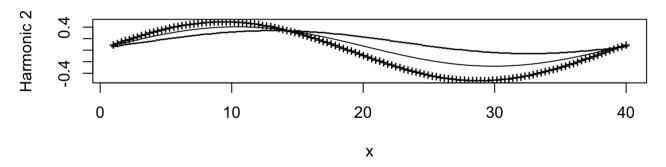
5.3 pivot the new dataframe

5.4 fPCA on new dataframe

PCA function 1 (Percentage of variability 63.3)



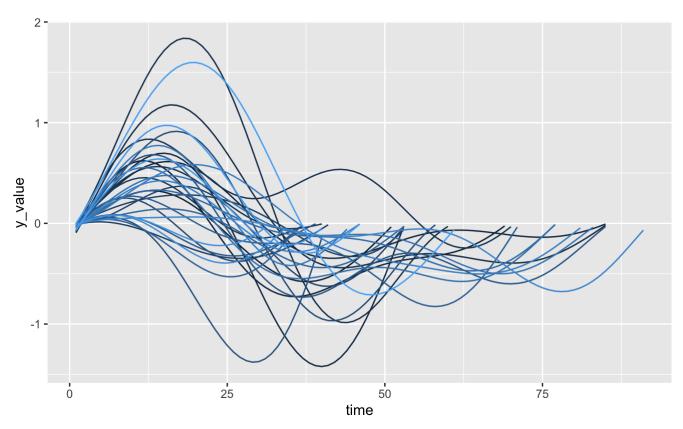
PCA function 2 (Percentage of variability 29.2)



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.position="top")</pre>
```

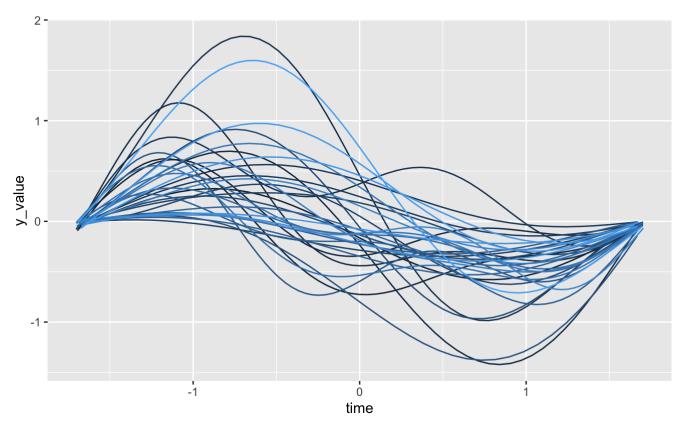




6.2 Scale the timeframe of each node

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

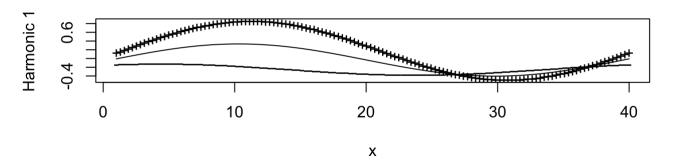




6.3 pivot the new dataframe

6.4 fPCA on new dataframe

PCA function 1 (Percentage of variability 63.1)



PCA function 2 (Percentage of variability 32.5)

