Ancestral Graphs example with Stop-signal data

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Before running throught the example bellow get an intution for the method and stats reading Lourens Waldorp's 2011 NI paper here.

1. Getting started

Ancestral Graphs relies on a number of packages such as graphs, RBGL and ggm. These packages are updated in a fast rate, and for now you will have to run an older version of R (version 3.0.2) in order to use the validated versions of these packages for AG. You can find the checked package versions in the folder AG_codes/Rpackage_checked.

For me it works really well to use RSwich to swich between the newest version of R and this older version for AG. After installing the older version of R, and using RSwich make sure that you are running R version 3.0.2 (2013-09-25).

```
# are you running the correct version of R?
sessionInfo()
```

Now set the Base directory path to the AG_stopexample folder on your computer

```
# set datadir Base
dir_base='/Users/sarajahfari/Github/AG_stopexample'

#dir AG_codes, example_codes,
dir_agcodes=paste(dir_base,'/AG_codes',sep='')
dir_excodes=paste(dir_base,'/Example_codes',sep='')
dir_data=paste(dir_base,'/Example_data',sep='')
```

and get/source the other required codes and packages

2. Get data, and define trials/conditions/regions

Read the single subjects data list with single trial estimates for a number of ROI's that will be used to infer connectivity. Next, I always create an index file that is easy to use with the AG codes.

```
# this reads the single subject data termed yDat, and creates
# the condition indexes in a seperate list
source(paste(dir_excodes, "/index_yDatstop.R", sep = ""), chdir = F)
head(yDat[[1]])
```

```
CaudateR40exc maxGPeR30exc maxGPiR30exc maxSTNR25exc
               -0.595375
                                         0.066915 -0.031055 -0.329973
## 01GoL 1
                             0.442298
## 01GoL 10
                            -0.026887
                0.105857
                                          0.057072
                                                     -0.083707 -0.144638
                                                     -0.017357 0.291495
## 01GoL_11
                                         0.500733
                0.161698
                            0.380750
## 01GoL 12
               -0.032499
                             0.463627
                                        -0.033962
                                                     -0.052651 0.278092
                                                      0.159087 0.207860
## 01GoL 13
                0.106706
                             0.105174
                                         0.095832
## 01GoL 14
               -0.500952
                            -0.032343
                                        -0.253548
                                                      0.308119 0.607734
##
           MotorCBA4aL PreSMARsmall ThalamusR40exc
                                                      type
## 01GoL_1
             -0.398792
                         -0.369048
                                        -0.183989 01GoL 1
## 01GoL_10
            -0.194387
                          0.450988
                                         0.012323 01GoL_10
## 01GoL_11
              0.816016
                          0.066621
                                          0.534921 01GoL_11
                                          0.272632 01GoL_12
## 01GoL_12
              0.268594
                          -0.064747
## 01GoL_13
              0.298172
                         0.209098
                                          0.401299 01GoL_13
## 01GoL_14 -0.012316
                        -0.080152
                                         -0.199831 01GoL_14
```

Then, define the trial types of this task which you would like to model, and the nodes which should be in your model space

```
# these are the conditions that are evaluated seperatly in
# FitmodelsStop.R (see below)
cond = list(ST, SR, Go, GoL, GoR)
names(cond) = c("ST", "SR", "Go", "GoL", "GoR")
# this is just a list with the row_indexes per subject for
# the succesfull stop trials
head(cond$ST)
## $\sub-001\
## [1] 50 51 52 53 54 55 56 57 58 59 60 61 124 125 126 127 128
## [18] 129 130 131 132 133 134 135
## $\sub-002\
## [1] 50 51 52 53 54 55 56 57 58 59 60 125 126 127 128 129 130
## [18] 131 132 133 134 135 136
##
## $\sub-003\
  [1] 51 52 53 54 55 56 57 58 59 60 61 126 127 128 129 130 131
## [18] 132 133 134 135 136 137 138 139 140
##
## $\sub-004\
## [1] 51 52 53 54 55 56 57 58 59 60 61 126 127 128 129 130 131
## [18] 132 133 134 135 136 137 138 139 140 141
##
## $\sub-005\
  [1] 49 50 51 52 53 54 55 56 57 58 122 123 124 125 126 127 128
## [18] 129 130 131 132 133 134
##
## $\sub-006\
## [1] 50 51 52 53 54 55 56 57 58 59 60 61 62 63 124 125 126
## [18] 127 128 129 130 131 132 133
# if responses are made with the left and right it can be
# difficult to fit the model with collapsed trials but also
# note that this wil reduce your number of trials by half
# cond = list(STL, STR, SRL, SRR, GoL, GoR)
# names(cond)=c('STL', 'STR', 'SRL', 'SRR', 'GoL', 'GoR')
```

3. Define the connectivity between nodes in different models

```
# source the models to use, these are the defined PFC-BG
# models to evaluate for connectivity/fits
source(paste(dir_excodes, "/StopModel.R", sep = ""), chdir = F)
# try [this is an interactive plot, you can change the layout
# of nodes and get out with esc]
drawGraph(Ag0) # see StopModel.R for definitions, but this should be the hyperdirect-Indirect model
# Now do the connectivity evaluation for each subject, each
# model/condition, to compute aic/bic and n-fits
source(paste(dir_excodes, "/FitmodelsStop.R", sep = ""), chdir = F)
Now evaluate fit and random effects AIC/BIC at group level
# this is the grouplevel AIC (see FitmodelsStop for code) for
# each model(cols)/condition(rows)
round(AMT, digits = 0)
##
       hindi
               hyp indir
                          dir nfit-hindi nfit-hyp nfit-indir nfit-dir
## ST
       7161 8298 7509 7748
                                       43
                                                 43
                                                                     43
## SR
       7456 8772 7955 8238
                                       43
                                                 43
                                                            43
                                                                     43
## Go 21763 25600 22884 23953
                                        0
                                                 0
                                                                      0
                                                             1
## GoL 11762 13772 12387 12902
                                       11
                                                 5
                                                            16
                                                                     13
## GoR 11397 13353 11998 12516
                                       13
                                                 13
                                                                     18
# this is the grouplevel BIC (see FitmodelsStop for code) for
# each model(cols)/condition(rows)
round(BMT, digits = 0)
##
                           dir nfit-hindi nfit-hyp nfit-indir nfit-dir
       hindi
               hyp indir
## ST
        7241 8357 7574 7807
                                       43
                                                 43
                                                            43
                                                                     43
## SR
        7536 8832 8020 8298
                                       43
                                                 43
                                                            43
                                                                     43
## Go 21864 25675 22965 24028
                                        0
                                                 0
                                                            1
                                                                      0
## GoL 11852 13840 12460 12969
                                                 5
                                       11
                                                            16
                                                                     13
## GoR 11486 13419 12070 12582
                                       13
                                                 13
                                                            19
                                                                     18
```

4. Select the winning model (hindi) to derive single subject connectivity estimates

```
# ag0 was the hindi model in StopModel.R
hindi = ag0
# conditions to get beta for [these are the conditions where
# the model fitted all ppn]
```

```
cond2 = list(ST, SR)
names(cond2) = c("ST", "SR")
# Make ST and SR lists using yDat
ST = list()
SR = list()
for (i in 1:length(yDat)) {
    ST[[i]] = yDat[[i]][cond2[[1]][[i]], C.Lab]
    SR[[i]] = yDat[[i]][cond2[[2]][[i]], C.Lab]
}
# compute covarience matrix
covList.ST = lapply(ST, cov)
covList.SR = lapply(SR, cov)
fitST = list()
fitSR = list()
for (i in 1:length(yDat)) {
    fitST[[i]] <- fitAncestralGraph(hindi, covList.ST[[i]], dim(ST[[i]])[1])</pre>
    fitSR[[i]] <- fitAncestralGraph(hindi, covList.SR[[i]], dim(SR[[i]])[1])</pre>
}
# compute the variance of beta per subject
var.g.ST <- ag.var.group(fitST, covList.ST, ST)</pre>
var.g.SR <- ag.var.group(fitSR, covList.SR, SR)</pre>
# save matrix for var beta and beta itself (zijn nog leeg)
varbeta.ST = matrix(, dim(var.g.ST)[3], dim(var.g.ST)[1])
varbeta.SR = matrix(, dim(var.g.SR)[3], dim(var.g.SR)[1])
beta.ST = matrix(, length(yDat), length(ag.theta(fitST[[1]])))
beta.SR = matrix(, length(yDat), length(ag.theta(fitSR[[1]])))
for (i in 1:length(yDat)) {
    varbeta.ST[i, ] = diag(var.g.ST[, , i])
    varbeta.SR[i, ] = diag(var.g.SR[, , i])
    beta.ST[i, ] = ag.theta(fitST[[i]])
    beta.SR[i, ] = ag.theta(fitSR[[i]])
}
# these are the standarized connections (to use for group
# comparisons)
sbeta.ST = round(beta.ST/varbeta.ST, dig = 8)
sbeta.SR = round(beta.SR/varbeta.SR, dig = 8)
```

Finally, select the defined directed and undirected connections and put this in a usable list which can be saved as tsy

```
# select the defined directed and undirected connections
beta.ST=beta.ST[,c(1:8,10)]
beta.SR=beta.SR[,c(1:8,10)]
sbeta.ST=sbeta.ST[,c(1:8,10)]
```

```
sbeta.SR=sbeta.SR[,c(1:8,10)]
colnames(beta.ST) =paste('ST-',c('caud->gpe','presma->caud','presma->stn','ifg->caud',
                               'ifg->stn','stn->gpi','gpe->gpi','gpi->thalamus',
                               'presma-ifg'), sep='')
colnames(beta.SR) =paste('SR-',c('caud->gpe','presma->caud','presma->stn','ifg->caud',
                               'ifg->stn','stn->gpi','gpe->gpi','gpi->thalamus',
                               'presma-ifg'), sep='')
colnames(sbeta.ST) =colnames(beta.ST)
colnames(sbeta.SR) =colnames(beta.SR)
Beta_output=list(Connectivity_ST=beta.ST, Connectivity_SR=beta.SR, standarized_con_ST=sbeta.ST, standariz
for (rn in 1:length(Beta_output)) {rownames(Beta_output[[rn]])=names(yDat)}
# arrow is direct in col.names, and - (presma-ifg) is undirected
head(Beta_output$Connectivity_ST)
          ST-caud->gpe ST-presma->caud ST-presma->stn ST-ifg->caud
## sub-001 -0.32534493
                         -0.55749704 -0.007316717 0.266109294
## sub-002 -0.48860562
                          -0.33756271 -0.276528991 0.002147515
## sub-003
          0.08619859
                         -0.40180361 -0.041358759 0.035535550
## sub-004 -0.24045212
                         ## sub-005 -0.12537864
                          -0.38403947
                                        0.126770183 -0.227053967
## sub-006 -0.44078481
                          -0.02563003
                                        0.128984510 -0.443016775
##
           ST-ifg->stn ST-stn->gpi ST-gpe->gpi ST-gpi->thalamus
-0.14515365
## sub-002 0.114201882 0.055687908 -0.7558637
                                                   -1.57800436
## sub-003 0.003940724 0.208368482 -0.8793710
                                                  -0.29854938
## sub-004 -0.816364543 -0.207181870 -0.2668323
                                                   -0.43127709
## sub-005 -0.396544915 0.061684750 -0.6557039
                                                   -0.05550985
## sub-006 -0.395269876 0.105049133 -0.5514878
                                                   -0.22893107
##
          ST-presma-ifg
## sub-001
             0.4615812
## sub-002
              0.3413610
## sub-003
              0.1925970
## sub-004
             0.2600340
## sub-005
              0.0893940
## sub-006
              0.3113932
# arrow is direct in col.names, and - (presma-ifq) is undirected
head(Beta_output$standarized_con_ST)
          ST-caud->gpe ST-presma->caud ST-presma->stn ST-ifg->caud
## sub-001 -0.00009765
                          -0.00005072
                                        -0.00000064 0.00002781
## sub-002 -0.00000096
                          -0.00074737
                                        -0.00002196 0.00001327
## sub-003
          0.00402757
                         -0.70181803 -0.02425740 0.02596681
## sub-004 -0.00017681
                         -0.00009260
                                        0.00002705 -0.00017788
## sub-005 -0.00122267
                          -0.00751744
                                         0.00037188 -0.00204955
## sub-006 -0.00001853
                          -0.00018219
                                         0.00050155 -0.00007106
##
          ST-ifg->stn ST-stn->gpi ST-gpe->gpi ST-gpi->thalamus ST-presma-ifg
```

```
## sub-001 -0.00001255 0.00000026 -0.00000768
                                                   -0.00000097
                                                                  0.00034208
## sub-002 0.00012644 0.00000100 -0.00000632
                                                                  0.00003346
                                                   -0.0000051
## sub-003 0.01039742 0.02229524 -0.00324629
                                                   -0.00356488
                                                                  0.01971200
## sub-004 -0.00007555 -0.00000733 -0.00001859
                                                   -0.00001560
                                                                  0.00133518
## sub-005 -0.00057094 0.00007966 -0.00022248
                                                   -0.00005725
                                                                  0.00468278
## sub-006 -0.00000868 0.00000170 -0.00000289
                                                   -0.00000834
                                                                  0.00011922
```

All of these steps are also in the *EvaluateStop_example.R*. If you have questions let me know. Have fun!