

PAIRS Wave 7

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1 Workspace

1.1 Packages

```
library(ggplot2)
library(qgraph)
library(mlVAR)
library(graphicalVAR)
library(bootnet)
library(NetworkComparisonTest)
library(knitr)
library(plyr)
library(dplyr)
library(gridExtra)
library(tidyr)
library(Rmisc)
library(psych)
library(lme4)
```

1.2 Load Data

```
wave7_all <- read.csv("~/Box Sync/network/PAIRS/Wave 7/esm_w7_RENAMED_all.csv")
wave7_all <- tbl_df(wave7_all)
```

Going to add item table here.

1.3 Prepare data

```
#Getting necessary columns, subject ID in col2
#Keeping subject ID and all esm.BFI items
w7 <- dplyr::select(wave7_all, esm.IDnum.w7, esm.PR001.w7, esm.PR003.w7, esm.PR004.w7, esm.PR005.w7, m
w7 <- w7[,-which(colnames(w7) %in% c("esm.BFI20.w7", "esm.BFI12.w7"))]

#Renaming variables for better ease of use
varnames <- c("SID", "freq", "hourBlock", "day", "beepvar",
              "E_outgoing", "E_quiet",
              "C_lazy", "C_reliable",
              "N_worried", "N_relaxed",
              "N_depressed", "A_rude",
              "A_kind")#, "authentic",
              #"SE", "Happy", "Lonely",
              #"pos_emotion", "neg_emotion")
colnames(w7) <- varnames
#varnames <- varnames[-c(17:18)]
#w7 <- dplyr::select(w7, -pos_emotion, -neg_emotion)
w7$SID <- factor(w7$SID)

refcols <- c("SID", "freq", "hourBlock", "day", "beepvar",
             "A_rude", "E_quiet", "C_lazy",
             "N_relaxed", "N_depressed", "E_outgoing",
```

```

      "A_kind", "C_reliable", "N_worried")#,
      #"pos_emotion", "neg_emotion", "authentic",
      #"SE", "Happy", "Lonely")
w7 <- w7[,c(refcols, setdiff(names(w7), refcols))]

```

1.4 Screen Participants

To be able to construct individual networks for participants, we ideally need approximately 50 measurement points. However, for current purposes, we will keep all participants who have at least 20 responses, lest we eliminate a large portion of our subjects.

```

problem <- plyr::count(w7$SID)
problem <- problem$x[which(problem$freq < 10)]

excluded <- w7[which(w7$SID %in% problem),]
w7 <- w7[-which(w7$SID %in% problem),]

```

1.5 Replace Agreeableness items

```

# replace NA values for agreeableness items with person average #
for (i in unique(w7$SID)){
  mean_A_rude <- mean(w7$A_rude[w7$SID == i])
  w7$A_rude[is.na(w7$A_rude)] <- mean_A_rude
  mean_A_kind <- mean(w7$A_kind[w7$SID == i])
  w7$A_kind[is.na(w7$A_kind)] <- mean_A_kind
}

```

1.6 Within-Person Centering

```

w7 <- w7[order(w7$SID, w7$day, w7$hourBlock),]
w7_com <- w7[complete.cases(w7),]
w7_com <- w7_com[order(w7_com$SID, w7_com$day, w7_com$hourBlock),]
w7_com$beepvar2 <- NA
w7_com$beepvar3 <- NA
for (i in unique(w7_com$SID)){
  for (k in unique(w7_com$day[which(w7_com$SID == i)])){
    n <- dim(w7_com[which(w7_com$SID == i & w7_com$day == k),])[1]
    w7_com$beepvar2[which(i == w7_com$SID & w7_com$day == k)] <- seq(1,n,1)
  }
  n2 <- dim(w7_com[which(w7_com$SID == i),])[1]
  w7_com$beepvar3[which(w7_com$SID == i)] <- seq(1,n2,1)
}

```

```

w7_centered <- data.frame(ddply(w7_com[, -c(2:5, 15:16)], .(SID), colwise(function(x) x - mean(x, na.rm = T)

```

```

# Make numeric subject IDs for each df because mlVAR #
# won't run for factors #
w7_com$SID2 <- as.numeric(w7_com$SID)
w7_centered$SID2 <- as.numeric(w7_centered$SID)

```

```
write.csv(w7, "~/Box Sync/network/PAIRS/Wave 7/esm_w7_networks.csv", row.names = F)

write.csv(w7, "~/Box Sync/network/PAIRS/Wave 7/esm_w7_centered.csv", row.names = F)
```

Variable	New Name	Description
esm.IDnum.w7	SID	numeric variable; identification number
esm.BFI37.w7	A_rude	agreeableness, negative; "During the last hour, how rude were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI21.w7	E_quiet	extraversion, negative; "During the last hour, how quiet were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI23.w7	C_lazy	conscientiousness, negative; "During the last hour, how lazy were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI09.w7	N_relaxed	neuroticism, positive; "During the last hour, how relaxed were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI04.w7	N_depressed	neuroticism, positive; "During the last hour, did you feel 'depressed, blue'?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI36.w7	E_outgoing	extraversion, positive; "During the last hour, how 'outgoing, sociable' were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI32.w7	A_kind	agreeableness, positive; "During the last hour, how 'considerate, kind' were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI13.w7	C_reliable	conscientiousness, positive; "During the last hour, how reliable were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI19.w7	N_worried	neuroticism, positive; "During the last hour, how worried were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI20.w7	O_imagine	openness, positive; "During the last hour, how actively imaginative were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very
esm.BFI12.w7	A_quarrel	neuroticism, positive; "During the last hour, how quarrelsome were you?" Likert scale from 1 to 5; 1 = Not a lot, 3 = Somewhat, 5 = Very

```
# raw data #
fit1_w7 <-
  mlVAR_test(w7_com,
    vars = colnames(w7)[6:14], #4:18
    idvar = "SID2",
    lags = 1,
    #dayvar = "day",
    beepvar = "beepvar3",
    temporal = "orthogonal",
    verbose = FALSE,
    scale = FALSE)
```

```

# within-person centered data #
fit1_w7_centered <-
  mlVAR_test(w7_centered,
    vars = colnames(w7_centered)[2:10], #2:16
    idvar = "SID2",
    lags = 1,
    #dayvar = "day",
    beepvar = "beepvar3",
    temporal = "orthogonal",
    verbose = FALSE,
    scale = FALSE)

```

1.7 Random Networks for Comparison

```

npeople <- length(names(fit1_w7$results$Beta$subject))
nv <- 9
subs <- names(fit1_w7$results$Beta$subject)
nTime <- mean(plyr::count(w7$SID2[which(w7$SID2 %in% subs)])$freq)
Model <- mlVARsim(nPerson = npeople, nNode = nv, nTime = nTime, lag=1)
colnames(Model$Data)[1:9] <- varnames[6:14]
Model$vars <- varnames[6:14]
fit_random <- mlVAR_test(Model$Data, vars = Model$vars, idvar = Model$idvar, lags = 1,
  temporal = "orthogonal", scale = FALSE, verbose = FALSE)

```

1.8 Temporal Relationships

The graphs below show (1) the raw, directed network using gLASSO estimation and (2) the estimated, directed network using a univariate multilevel vector autoregressive models.

```

# Compare temporal relationships
cors <- cor_auto(w7[,6:14], detectOrdinal = F)
par(mfrow = c(1,2))
# graph <- qgraph(
#   cors,
#   title = "True temporal relationships",
#   layout = "spring",
#   graph = "glasso",
#   sampleSize = length(unique(w7$SID2)),
#   tuning = .5,
#   directed = T,
#   details = T)#,
#   #minimum = .2)
#qgraph(getWmat(graph), directed = T)

plot_w7 <-
  #makeBW(
    plot(fit1_w7,
      "temporal",
      #title = "Raw Estimated temporal relationships",
      layout = "spring",

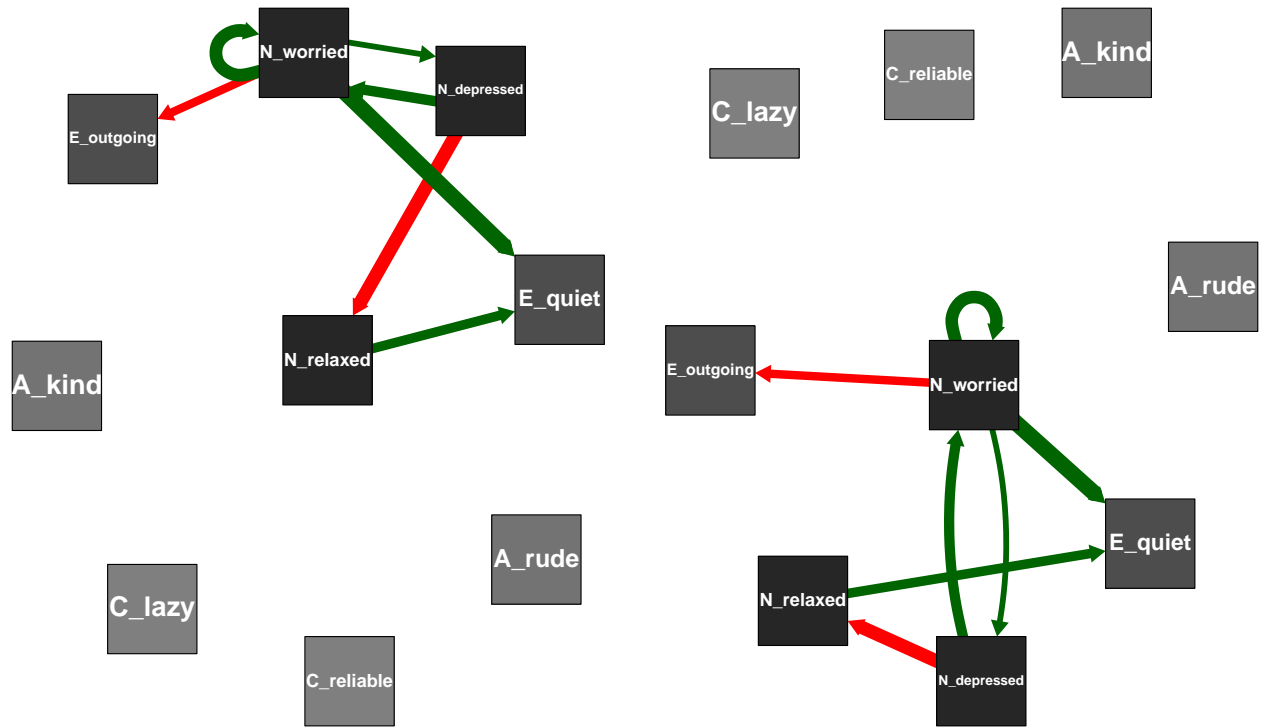
```

```

groups = list(Neuroticism = c(4,5,9),
              Extraversion = c(2,6),
              Agreeableness = c(1,7),
              Conscientiousness = c(3,8)),
shape = "square",
cut = .015,
label.color = "white",
label.font = 2,
nonsig = 'hide',
curve = -1,
#legend.cex = .3,
legend = FALSE,
repulsion = .5,
node.width = 1.4,
edge.width = 2,
label.cex = 6,
label.fill.vertical = 1,
label.fill.horizontal = 1,
#gray = T,#),
color = c("gray15", "gray30", "gray45", "gray50"))
# text(.7,1, labels = "negative", adj = 0, cex = 1.2)
# text(.7,1.1, labels = "positive", adj = 0, cex = 1.2)
# text(.55, 1.2, labels = "Edge Weights", adj = 0, cex = 1.2)
# segments(.55, 1, x1 = .65, lty = 2, lwd = 1.5)
# segments(.55, 1.1, x1 = .65, lty = 1, lwd = 1.5)

plot_w7_centered <-
#makeBW(
plot(fit1_w7_centered,
     "temporal",
     #title = "Raw Estimated temporal relationships",
     layout = "spring",
     groups = list(Neuroticism = c(4,5,9),
                   Extraversion = c(2,6),
                   Agreeableness = c(1,7),
                   Conscientiousness = c(3,8)),
     shape = "square",
     cut = .015,
     label.color = "white",
     label.font = 2,
     nonsig = 'hide',
     curve = -1,
     #legend.cex = .3,
     legend = FALSE,
     repulsion = .5,
     node.width = 1.4,
     edge.width = 2,
     label.cex = 6,
     label.fill.vertical = 1,
     label.fill.horizontal = 1,
     gray = F,#),
     color = c("gray15", "gray30", "gray45", "gray50"))

```



```
# text(.7,1, labels = "negative", adj = 0, cex = 1.2)
# text(.7,1.1, labels = "positive", adj = 0, cex = 1.2)
# text(.55, 1.2, labels = "Edge Weights", adj = 0, cex = 1.2)
# segments(.55, 1, x1 = .65, lty = 2, lwd = 1.5)
# segments(.55, 1.1, x1 = .65, lty = 1, lwd = 1.5)
```

```
#model summary
sum_fit1_w7      <- summary(fit1_w7)
sum_fit1_w7_centered <- summary(fit1_w7_centered)
```

```
kable(sum_fit1_w7$temporal, caption = "Estimated Temporal Relationships")
```

Table 2: Estimated Temporal Relationships

from	to	lag	fixed	SE	P	ran_SD
A_rude	A_rude	1	0.022	0.024	0.370	0.059
A_rude	E_quiet	1	-0.050	0.047	0.284	0.000
A_rude	C_lazy	1	-0.025	0.050	0.625	0.140
A_rude	N_relaxed	1	0.019	0.039	0.629	0.019
A_rude	N_depressed	1	0.010	0.030	0.736	0.042
A_rude	E_outgoing	1	0.057	0.047	0.220	0.000
A_rude	A_kind	1	-0.017	0.036	0.635	0.084
A_rude	C_reliable	1	0.015	0.037	0.689	0.061
A_rude	N_worried	1	0.037	0.036	0.305	0.000
E_quiet	A_rude	1	-0.009	0.016	0.597	0.000
E_quiet	E_quiet	1	0.001	0.035	0.984	0.000
E_quiet	C_lazy	1	0.002	0.033	0.951	0.021
E_quiet	N_relaxed	1	-0.025	0.029	0.380	0.000
E_quiet	N_depressed	1	0.011	0.021	0.582	0.000
E_quiet	E_outgoing	1	0.011	0.035	0.754	0.000

from	to	lag	fixed	SE	P	ran_SD
E_quiet	A_kind	1	0.044	0.024	0.068	0.000
E_quiet	C_reliable	1	0.037	0.026	0.156	0.000
E_quiet	N_worried	1	0.003	0.027	0.903	0.000
C_lazy	A_rude	1	0.012	0.017	0.496	0.064
C_lazy	E_quiet	1	-0.004	0.029	0.896	0.000
C_lazy	C_lazy	1	0.024	0.035	0.501	0.138
C_lazy	N_relaxed	1	0.019	0.024	0.416	0.000
C_lazy	N_depressed	1	-0.006	0.019	0.753	0.037
C_lazy	E_outgoing	1	0.010	0.030	0.734	0.048
C_lazy	A_kind	1	0.020	0.020	0.326	0.013
C_lazy	C_reliable	1	-0.010	0.027	0.713	0.091
C_lazy	N_worried	1	-0.029	0.022	0.189	0.000
N_relaxed	A_rude	1	-0.027	0.017	0.106	0.000
N_relaxed	E_quiet	1	0.096	0.037	0.009	0.069
N_relaxed	C_lazy	1	0.007	0.033	0.823	0.000
N_relaxed	N_relaxed	1	-0.021	0.032	0.514	0.081
N_relaxed	N_depressed	1	-0.007	0.021	0.738	0.000
N_relaxed	E_outgoing	1	-0.056	0.035	0.112	0.000
N_relaxed	A_kind	1	-0.041	0.025	0.099	0.000
N_relaxed	C_reliable	1	-0.045	0.027	0.093	0.000
N_relaxed	N_worried	1	0.019	0.033	0.558	0.120
N_depressed	A_rude	1	-0.013	0.021	0.529	0.000
N_depressed	E_quiet	1	0.025	0.045	0.574	0.000
N_depressed	C_lazy	1	0.002	0.042	0.960	0.000
N_depressed	N_relaxed	1	-0.134	0.041	0.001	0.083
N_depressed	N_depressed	1	0.029	0.040	0.466	0.175
N_depressed	E_outgoing	1	-0.036	0.044	0.412	0.000
N_depressed	A_kind	1	-0.059	0.040	0.139	0.134
N_depressed	C_reliable	1	-0.035	0.038	0.355	0.082
N_depressed	N_worried	1	0.104	0.035	0.003	0.000
E_outgoing	A_rude	1	0.011	0.017	0.538	0.000
E_outgoing	E_quiet	1	0.014	0.037	0.708	0.000
E_outgoing	C_lazy	1	0.017	0.035	0.622	0.000
E_outgoing	N_relaxed	1	-0.043	0.030	0.161	0.000
E_outgoing	N_depressed	1	0.005	0.022	0.820	0.023
E_outgoing	E_outgoing	1	-0.019	0.037	0.613	0.000
E_outgoing	A_kind	1	0.007	0.026	0.785	0.000
E_outgoing	C_reliable	1	0.005	0.028	0.863	0.019
E_outgoing	N_worried	1	-0.012	0.029	0.662	0.023
A_kind	A_rude	1	-0.007	0.018	0.691	0.021
A_kind	E_quiet	1	-0.021	0.038	0.578	0.000
A_kind	C_lazy	1	0.014	0.036	0.687	0.000
A_kind	N_relaxed	1	0.043	0.031	0.166	0.000
A_kind	N_depressed	1	-0.022	0.025	0.384	0.067
A_kind	E_outgoing	1	0.042	0.038	0.260	0.000
A_kind	A_kind	1	0.030	0.030	0.307	0.081
A_kind	C_reliable	1	0.045	0.029	0.123	0.000
A_kind	N_worried	1	-0.012	0.029	0.689	0.000
C_reliable	A_rude	1	-0.014	0.017	0.408	0.015
C_reliable	E_quiet	1	-0.021	0.036	0.565	0.000
C_reliable	C_lazy	1	0.012	0.034	0.722	0.000
C_reliable	N_relaxed	1	-0.002	0.030	0.943	0.000

from	to	lag	fixed	SE	P	ran_SD
C_reliable	N_depressed	1	0.005	0.025	0.853	0.077
C_reliable	E_outgoing	1	0.037	0.036	0.293	0.000
C_reliable	A_kind	1	0.008	0.025	0.757	0.000
C_reliable	C_reliable	1	0.052	0.027	0.056	0.000
C_reliable	N_worried	1	-0.016	0.030	0.595	0.069
N_worried	A_rude	1	0.023	0.018	0.195	0.000
N_worried	E_quiet	1	0.158	0.038	0.000	0.000
N_worried	C_lazy	1	-0.025	0.035	0.482	0.000
N_worried	N_relaxed	1	-0.043	0.038	0.254	0.136
N_worried	N_depressed	1	0.060	0.023	0.008	0.000
N_worried	E_outgoing	1	-0.084	0.038	0.028	0.046
N_worried	A_kind	1	-0.022	0.028	0.438	0.051
N_worried	C_reliable	1	-0.051	0.029	0.075	0.000
N_worried	N_worried	1	0.127	0.036	0.000	0.138

```
kable(sum_fit1_w7_centered$temporal, caption = "Estimated Temporal Relationships")
```

Table 3: Estimated Temporal Relationships

from	to	lag	fixed	SE	P	ran_SD
A_rude	A_rude	1	0.021	0.024	0.376	0.061
A_rude	E_quiet	1	-0.050	0.046	0.277	0.000
A_rude	C_lazy	1	-0.024	0.050	0.634	0.145
A_rude	N_relaxed	1	0.018	0.039	0.640	0.053
A_rude	N_depressed	1	0.011	0.029	0.697	0.044
A_rude	E_outgoing	1	0.057	0.046	0.210	0.000
A_rude	A_kind	1	-0.016	0.037	0.662	0.095
A_rude	C_reliable	1	0.015	0.037	0.688	0.066
A_rude	N_worried	1	0.039	0.036	0.279	0.000
E_quiet	A_rude	1	-0.009	0.016	0.591	0.000
E_quiet	E_quiet	1	0.001	0.034	0.980	0.000
E_quiet	C_lazy	1	0.002	0.033	0.946	0.047
E_quiet	N_relaxed	1	-0.025	0.028	0.373	0.000
E_quiet	N_depressed	1	0.012	0.020	0.571	0.000
E_quiet	E_outgoing	1	0.011	0.034	0.756	0.000
E_quiet	A_kind	1	0.044	0.024	0.066	0.000
E_quiet	C_reliable	1	0.037	0.026	0.152	0.000
E_quiet	N_worried	1	0.002	0.026	0.925	0.021
C_lazy	A_rude	1	0.012	0.017	0.496	0.065
C_lazy	E_quiet	1	-0.004	0.029	0.892	0.015
C_lazy	C_lazy	1	0.023	0.035	0.516	0.139
C_lazy	N_relaxed	1	0.019	0.023	0.405	0.000
C_lazy	N_depressed	1	-0.006	0.019	0.758	0.040
C_lazy	E_outgoing	1	0.010	0.030	0.729	0.056
C_lazy	A_kind	1	0.021	0.020	0.307	0.023
C_lazy	C_reliable	1	-0.009	0.027	0.742	0.094
C_lazy	N_worried	1	-0.029	0.022	0.186	0.000
N_relaxed	A_rude	1	-0.027	0.016	0.103	0.000
N_relaxed	E_quiet	1	0.096	0.037	0.009	0.080
N_relaxed	C_lazy	1	0.007	0.033	0.826	0.000
N_relaxed	N_relaxed	1	-0.022	0.032	0.494	0.086

from	to	lag	fixed	SE	P	ran_SD
N_relaxed	N_depressed	1	-0.007	0.021	0.747	0.000
N_relaxed	E_outgoing	1	-0.056	0.034	0.101	0.000
N_relaxed	A_kind	1	-0.041	0.024	0.091	0.000
N_relaxed	C_reliable	1	-0.045	0.026	0.086	0.000
N_relaxed	N_worried	1	0.019	0.033	0.563	0.127
N_depressed	A_rude	1	-0.013	0.021	0.534	0.000
N_depressed	E_quiet	1	0.025	0.044	0.562	0.000
N_depressed	C_lazy	1	0.001	0.041	0.974	0.000
N_depressed	N_relaxed	1	-0.133	0.041	0.001	0.090
N_depressed	N_depressed	1	0.029	0.040	0.470	0.175
N_depressed	E_outgoing	1	-0.036	0.044	0.409	0.000
N_depressed	A_kind	1	-0.060	0.040	0.136	0.137
N_depressed	C_reliable	1	-0.034	0.038	0.369	0.089
N_depressed	N_worried	1	0.104	0.034	0.002	0.000
E_outgoing	A_rude	1	0.011	0.017	0.534	0.000
E_outgoing	E_quiet	1	0.014	0.036	0.699	0.000
E_outgoing	C_lazy	1	0.017	0.034	0.613	0.000
E_outgoing	N_relaxed	1	-0.042	0.030	0.155	0.000
E_outgoing	N_depressed	1	0.006	0.022	0.801	0.031
E_outgoing	E_outgoing	1	-0.019	0.036	0.599	0.000
E_outgoing	A_kind	1	0.007	0.025	0.776	0.000
E_outgoing	C_reliable	1	0.005	0.028	0.868	0.025
E_outgoing	N_worried	1	-0.012	0.028	0.660	0.032
A_kind	A_rude	1	-0.007	0.018	0.694	0.024
A_kind	E_quiet	1	-0.021	0.037	0.573	0.000
A_kind	C_lazy	1	0.015	0.035	0.670	0.000
A_kind	N_relaxed	1	0.042	0.031	0.171	0.000
A_kind	N_depressed	1	-0.022	0.025	0.384	0.071
A_kind	E_outgoing	1	0.042	0.037	0.257	0.000
A_kind	A_kind	1	0.029	0.030	0.327	0.085
A_kind	C_reliable	1	0.044	0.028	0.119	0.000
A_kind	N_worried	1	-0.011	0.029	0.705	0.000
C_reliable	A_rude	1	-0.014	0.017	0.400	0.018
C_reliable	E_quiet	1	-0.021	0.035	0.549	0.000
C_reliable	C_lazy	1	0.012	0.033	0.717	0.000
C_reliable	N_relaxed	1	-0.001	0.029	0.967	0.022
C_reliable	N_depressed	1	0.005	0.025	0.840	0.078
C_reliable	E_outgoing	1	0.038	0.035	0.282	0.000
C_reliable	A_kind	1	0.008	0.025	0.742	0.000
C_reliable	C_reliable	1	0.052	0.027	0.055	0.000
C_reliable	N_worried	1	-0.016	0.030	0.585	0.073
N_worried	A_rude	1	0.023	0.018	0.186	0.000
N_worried	E_quiet	1	0.159	0.037	0.000	0.000
N_worried	C_lazy	1	-0.026	0.035	0.464	0.000
N_worried	N_relaxed	1	-0.041	0.038	0.274	0.143
N_worried	N_depressed	1	0.060	0.022	0.007	0.000
N_worried	E_outgoing	1	-0.086	0.038	0.024	0.055
N_worried	A_kind	1	-0.022	0.028	0.434	0.054
N_worried	C_reliable	1	-0.052	0.028	0.067	0.018
N_worried	N_worried	1	0.126	0.036	0.000	0.143

1.9 Population Level Questions

1.9.1 Global Network Structure

1.9.1.1 Network Size

```
network_size <- data.frame(  
  #temporal - significant only  
  temporal = c(  
    sum(sum_fit1_w7$temporal$P < .05),  
    sum(sum_fit1_w7_centered$temporal$P < .05))  
  rownames(network_size) <- c("Raw", "Within-Person Centered")  
  kable(t(network_size), caption = "Network Size")
```

Table 4: Network Size

	Raw	Within-Person Centered
temporal	7	7

1.9.1.2 Network Density

Calculating Density:

1. From Bringmann et al. (2016): averaging over the absolute values of the slopes or edges in the network of the emotions of interest
2. Per Yoed's suggestion:
 - Potential Connections: $PC = \frac{n*(n-1)}{2}$, where n is the number of nodes.
 - Network Density: $\frac{ActualConnections}{PotentialConnections}$, where actual connections is the sum of the number of edges in the network.

```
n <- 9  
PC <- (n * (n-1))/2  
density <- data.frame(  
  #temporal  
  temporal = c(  
    sum(sum_fit1_w7$temporal$P < .05)/PC,  
    sum(sum_fit1_w7_centered$temporal$P < .05)/PC)  
  rownames(density) <- c("Raw", "Within-Person Centered")  
  kable(t(density), caption = "Network Density")
```

Table 5: Network Density

	Raw	Within-Person Centered
temporal	0.1944444	0.1944444

1.9.1.3 Network Activation

```
#####  
# temporal - significant only #  
#####
```

```

#raw
temporal <- data.frame(positive_edges = c(
sum(sum_fit1_w7$temporal$P < .05 & sum_fit1_w7$temporal$fixed > 0), # positive edges
sum(sum_fit1_w7_centered$temporal$P < .05 & sum_fit1_w7_centered$temporal$fixed > 0))) # positive edges

temporal$negative_edges <- c(
sum(sum_fit1_w7$temporal$P < .05 & sum_fit1_w7$temporal$fixed < 0), # negative edges
sum(sum_fit1_w7_centered$temporal$P < .05 & sum_fit1_w7_centered$temporal$fixed < 0)) # negative edges

temporal$t_value <- c(
#t test comparing positive and negative edges; not run because of too few observations
t.test(abs(sum_fit1_w7$temporal$fixed[which(sum_fit1_w7$temporal$P < .05 & sum_fit1_w7$temporal$fixed >
#centered
#t test comparing positive and negative edges; not run because of too few observations
t.test(abs(sum_fit1_w7_centered$temporal$fixed[which(sum_fit1_w7_centered$temporal$P < .05 & sum_fit1_w
rownames(temporal) <- c("Raw", "Within-Person Centered")
kable(t(temporal), caption = "Network Activation (Positive v. Negative Edges)")

```

Table 6: Network Activation (Positive v. Negative Edges)

	Raw	Within-Person Centered
positive_edges	5	5.0000000
negative_edges	2	2.0000000
t_value	0	-0.0174453

1.9.1.4 Bootstrapped Confidence Intervals

1.9.2 Local Network Structure

1.9.2.1 Centrality

Not sure why centrality plots for contemporaneous and between-subjects networks are identical.

```

#temporal
#raw
temporal_centrality_w7_raw <- centrality_auto(sum_fit1_w7$temporal[,c(1,2,4)])

## Node 1, Reach 9, Total 9
## Node 2, Reach 9, Total 18
## Node 3, Reach 9, Total 27
## Node 4, Reach 9, Total 36
## Node 5, Reach 9, Total 45
## Node 6, Reach 9, Total 54
## Node 7, Reach 9, Total 63
## Node 8, Reach 9, Total 72
## Node 9, Reach 9, Total 81

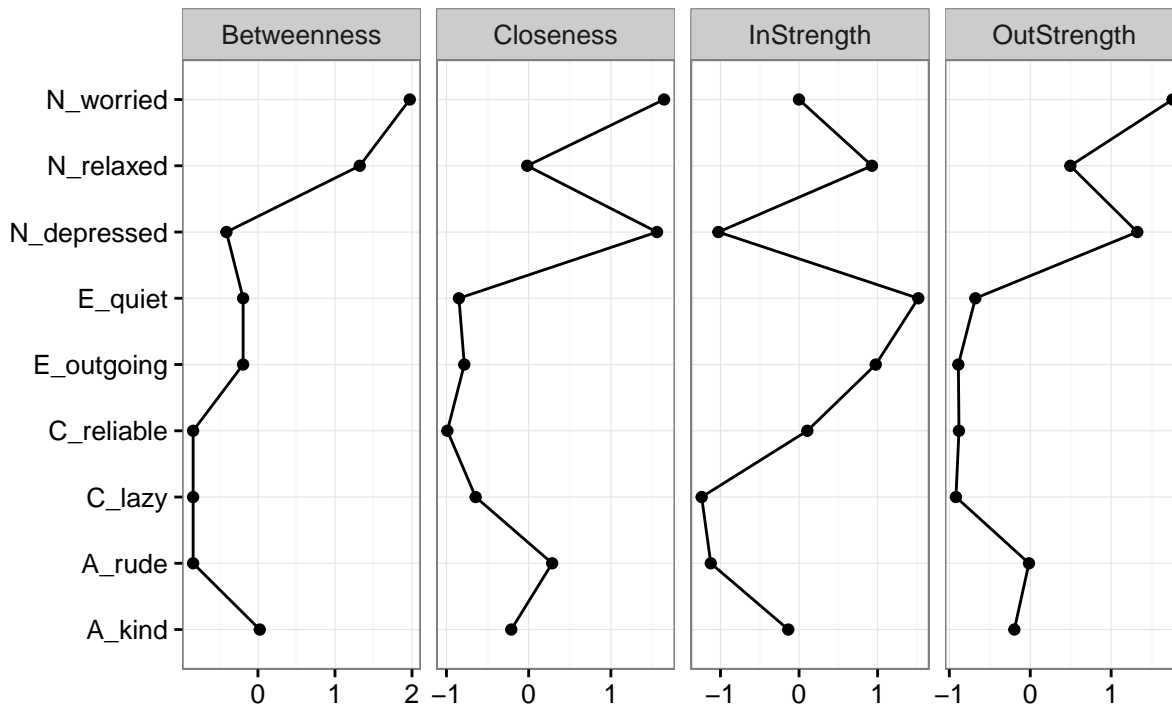
temporal_centrality_w7_raw_plot <- centralityPlot(sum_fit1_w7$temporal[,c(1,2,4)])

## Note: z-scores are shown on x-axis rather than raw centrality indices.

## Node 1, Reach 9, Total 9
## Node 2, Reach 9, Total 18

```

```
## Node 3, Reach 9, Total 27
## Node 4, Reach 9, Total 36
## Node 5, Reach 9, Total 45
## Node 6, Reach 9, Total 54
## Node 7, Reach 9, Total 63
## Node 8, Reach 9, Total 72
## Node 9, Reach 9, Total 81
```



```
#centered
temporal_centrality_w7_centered <- centrality_auto(sum_fit1_w7_centered$temporal[,c(1,2,4)])
```

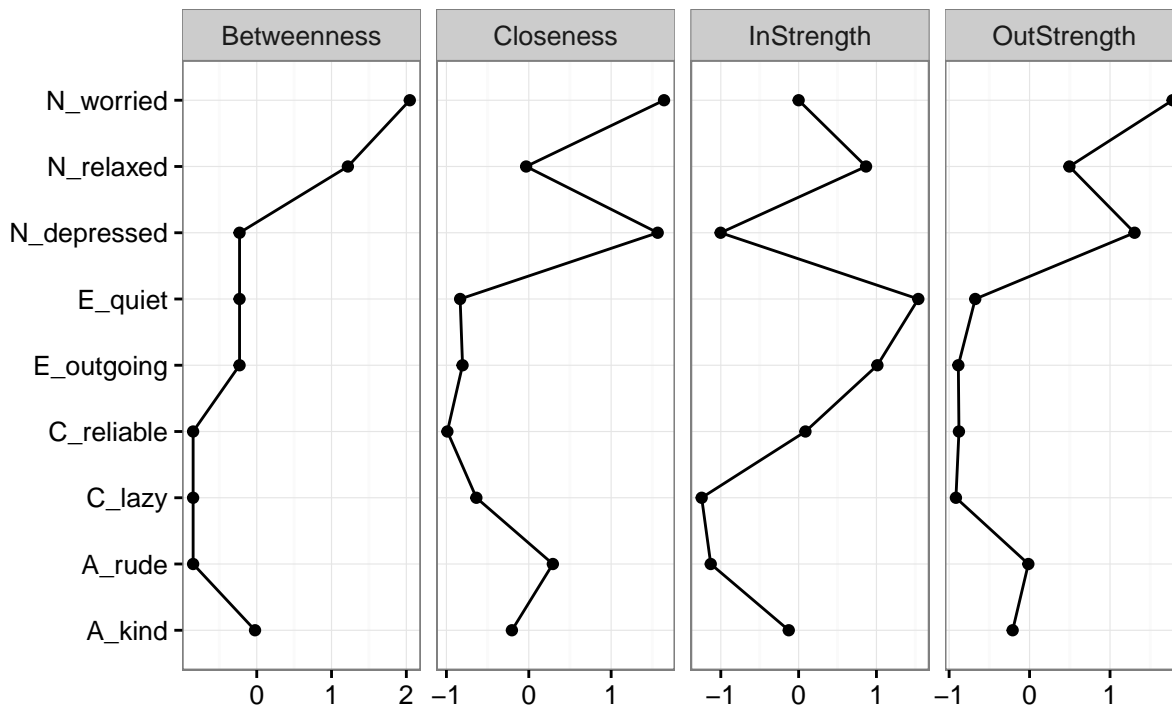
```
## Node 1, Reach 9, Total 9
## Node 2, Reach 9, Total 18
## Node 3, Reach 9, Total 27
## Node 4, Reach 9, Total 36
## Node 5, Reach 9, Total 45
## Node 6, Reach 9, Total 54
## Node 7, Reach 9, Total 63
## Node 8, Reach 9, Total 72
## Node 9, Reach 9, Total 81
```

```
temporal_centrality_w7_centered_plot <- centralityPlot(sum_fit1_w7_centered$temporal[,c(1,2,4)])
```

Note: z-scores are shown on x-axis rather than raw centrality indices.

```
## Node 1, Reach 9, Total 9
## Node 2, Reach 9, Total 18
## Node 3, Reach 9, Total 27
## Node 4, Reach 9, Total 36
## Node 5, Reach 9, Total 45
## Node 6, Reach 9, Total 54
## Node 7, Reach 9, Total 63
## Node 8, Reach 9, Total 72
```

Node 9, Reach 9, Total 81



1.9.2.1.1 Bootstrapped Confidence Intervals

1.9.2.2 Clustering Coefficients

```
#temporal
temporal_clustcoef_w7_raw      <- clustZhang(sum_fit1_w7$temporal[,c(1,2,4)])
temporal_clustcoef_w7_centered <- clustZhang(sum_fit1_w7_centered$temporal[,c(1,2,4)])
#clusteringTable(sum_fit1_w7$temporal[,c(1,2,4)])

save(temporal_centrality_w7_raw, temporal_centrality_w7_centered,
      temporal_clustcoef_w7_raw, temporal_clustcoef_w7_centered,
      file = "~/Box Sync/network/PAIRS/Wave 7/centrality_w7_pers.RData")
```

1.10 Inter-Individual Questions

Going to duplicate the analyses below (temporal subject networks) with contemporaneous subject networks

1.10.1 Bootstrapped Confidence Intervals of Parameter Estimates

```
raw_models <- fit1_w7$output
raw_boot_CI <- llply(raw_models, function(X) confint(X, method = "boot"))

centered_models <- fit1_w7_centered$output
centered_boot_CI <- llply(centered_models, function(X) confint(X, method = "boot"))
```

1.10.1.1 Raw Values

```
#Error in data[[idvar]] : subscript out of bounds
#Error in .subset2(x, i, exact = exact) :
# attempt to select less than one element in get1index
#data[[beepvar]] <- ave(data[[idvar]], data[[idvar]], data[[dayvar]],
# FUN = seq_along)
#problem appears to be in dayvar
#Error in validObject(.Object) :
#invalid class "dgCMatrix" object: invalid object for slot "i" in class "dgCMatrix":
#got class "numeric", should be or extend class "integer"
w7$beepvar <- ave(w7[["SID"]], w7[["SID"]], w7[["day"]],
                  FUN = seq_along)
bootnet(w7,
  default = "none",
  nBoots = 25,
  type = "nonparametric",
  prepFun = null,
  prepArgs = list(),
  estFun = mlVAR::mlVAR,
  estArgs = list(data = w7,
                 vars = colnames(w7)[4:16],
                 idvar = "SID2",
                 dayvar = "day",
                 beepvar = "beepvar",
                 lags = 1, temporal = "orthogonal",
                 verbose = FALSE),
  graphFun = function(x)x,
  graphArgs = list(type = "temporal",
                   title = "Estimated temporal relationships",
                   layout = "spring"),
  intFun = null,
  intArgs = list())
```

1.10.1.2 Centered Values

```
#Error in data[[idvar]] : subscript out of bounds
#Error in .subset2(x, i, exact = exact) :
# attempt to select less than one element in get1index
#data[[beepvar]] <- ave(data[[idvar]], data[[idvar]], data[[dayvar]],
```

```

# FUN = seq_along)
#problem appears to be in dayvar
#Error in validObject(.Object) :
#invalid class "dgCMatrix" object: invalid object for slot "i" in class "dgCMatrix":
#got class "numeric", should be or extend class "integer"
w7_centered$beepvar <- ave(w7_centered[["SID"]], w7_centered[["SID"]], w7_centered[["day"]],
FUN = seq_along)
bootnet(w7_centered,
  default = "none",
  nBoots = 25,
  type = "nonparametric",
  prepFun = null,
  prepArgs = list(),
  estFun = mlVAR::mlVAR,
  estArgs = list(data = w7_centered,
    vars = colnames(w7_centered)[2:14],
    idvar = "SID2",
    dayvar = "day",
    beepvar = "beepvar",
    lags = 1, temporal = "orthogonal",
    verbose = FALSE),
  graphFun = function(x)x,
  graphArgs = list(type = "temporal",
    title = "Estimated temporal relationships",
    layout = "spring"),
  intFun = null,
  intArgs = list())

```


2 Graphical VAR Models

```
comp <- ddply(w7, .(SID), numcolwise(mean))

with(comp, {
  comp <- comp[order(E_quiet, C_lazy, N_relaxed, N_depressed, E_outgoing, C_reliable, N_worried, A_rude,

#10308 and 27 have similar composite profiles

fit10308 <- graphicalVAR(w7[which(w7$SID == 10308),4:16],
  gamma = 0, verbose = F)
fit10219 <- graphicalVAR(w7[which(w7$SID == 10219),4:16],
  gamma = 0, verbose = F)

#compare plots of partial contemporaneous correlations for both subjects
par(mfcol = c(1,2))
plot(fit10308, include = "PCC", layout = "spring")
plot(fit10219, include = "PCC", layout = "spring")

#compare plots of partial directed correlations for both subjects
plot(fit10308, include = "PDC", layout = "spring")
plot(fit10219, include = "PDC", layout = "spring")
```

2.1 Idiographic Questions

2.1.1 Global Network Structure

2.1.1.1 Network Size

```
#Partial Contemporaneous Correlations (PCC)
sum(fit10308$PCC != 0)
sum(fit10219$PCC != 0)

#Partial Directed Correlations (PDC)
sum(fit10308$PDC != 0)
sum(fit10219$PDC != 0)
```

2.1.1.2 Network Density

```
n <- 13
PC <- (n * (n-1))/2

#PCC
sum(fit10308$PCC != 0) / PC
sum(fit10219$PCC != 0) / PC

#PDC
sum(fit10308$PDC != 0) / PC
sum(fit10219$PDC != 0) / PC
```

2.1.1.3 Network Activation

```

#Partial Contemporaneous Correlations (PCC)
#Subject 10308
sum(fit10308$PCC > 0) # positive edges
sum(fit10308$PCC < 0) # negative edges
#t test comparing positive and negative edges; not run because of too few observations
#t.test(abs(sum(fit10308$PCC > 0)), abs(sum(fit10308$PCC < 0)), equal.var = T)

#Subject 10219
sum(fit10219$PCC > 0) # positive edges
sum(fit10219$PCC < 0) # negative edges
#t test comparing positive and negative edges; not run because of too few observations
#t.test(abs(sum(fit10219$PCC > 0)), abs(sum(fit10219$PCC < 0)), equal.var = T)

#Partial Directed Correlations (PDC)
#Subject 10308
sum(fit10308$PDC > 0) # positive edges
sum(fit10308$PDC < 0) # negative edges
#t test comparing positive and negative edges; not run because of too few observations
#t.test(abs(sum(fit10308$PDC > 0)), abs(sum(fit10308$PDC < 0)), equal.var = T)

#Subject 10219
sum(fit10219$PDC > 0) # positive edges
sum(fit10219$PDC < 0) # negative edges
#t test comparing positive and negative edges; not run because of too few observations
#t.test(abs(sum(fit10219$PDC > 0)), abs(sum(fit10219$PDC < 0)), equal.var = T)

```

2.1.2 Local Network Structure

2.1.2.1 Centrality

```

#PCC
#adjacency matrices
PCC_adj_10308 <- melt(fit10308$PCC)
PCC_adj_10219 <- melt(fit10219$PCC)

#Subject 10308
PCC_centrality_10308 <- centrality_auto(PCC_adj_10308)
centralityPlot(PCC_adj_10308)
#Subject 10219
PCC_centrality_10219 <- centrality(PCC_adj_10219)
centralityPlot(PCC_adj_10219)

#PDC
#adjacency matrices
PDC_adj_10308 <- melt(fit10308$PDC)
PDC_adj_10219 <- melt(fit10219$PDC)

#Subject 10308
PDC_centrality_10308 <- centrality_auto(PDC_adj_10308)
centralityPlot(PDC_adj_10308)
#Subject 10219
PDC_centrality_10219 <- centrality_auto(PDC_adj_10219)
centralityPlot(PDC_adj_10219)

```