IIVL Analyses Tour

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Import Data

```
cdi.dat <- read.csv("cdi.csv",header=T,na.strings = c("-99","99"))
attach(cdi.dat)
names(cdi.dat)
## [1] "country" "collact" "arab" "id" "eff" "sdo" "anger"</pre>
```

Descriptives

id

```
summary(cdi.dat)
##
                        collact
       country
                                            arab
                                                              id
##
   Min.
          : 1.000
                     Min.
                            : 0.000
                                      Min.
                                             : 0.000
                                                        Min.
                                                               : 0.000
   1st Qu.: 3.000
                     1st Qu.: 1.000
                                      1st Qu.: 4.000
                                                        1st Qu.: 2.000
##
  Median : 7.000
                     Median : 2.000
                                      Median : 7.000
                                                        Median : 5.000
## Mean : 6.607
                     Mean
                           : 3.522
                                      Mean
                                             : 6.171
                                                        Mean
                                                               : 4.749
   3rd Qu.:10.000
                     3rd Qu.: 6.000
                                      3rd Qu.: 9.000
                                                        3rd Qu.: 7.000
##
##
  Max.
          :12.000
                     Max.
                            :10.000
                                      {\tt Max.}
                                              :10.000
                                                        Max.
                                                               :10.000
##
                     NA's
                            :167
                                      NA's
                                              :65
                                                        NA's
                                                               :75
##
         eff
                          sdo
                                          anger
##
   Min.
          : 0.000
                            :1.000
                                            : 0.000
                     Min.
                                     Min.
##
   1st Qu.: 3.000
                     1st Qu.:1.750
                                     1st Qu.: 1.500
                     Median :3.000
##
  Median : 5.000
                                     Median: 4.000
  Mean
          : 5.333
                     Mean
                            :3.166
                                     Mean
                                            : 3.902
   3rd Qu.: 7.000
                     3rd Qu.:4.500
                                     3rd Qu.: 6.000
##
           :19.000
                            :9.750
                                             :10.000
   Max.
                     Max.
                                     Max.
   NA's
           :6
                     NA's
                                     NA's
                                             :230
                            :17
sapply(cdi.dat, mean, na.rm=TRUE)
  country collact
                                    id
                                             eff
                         arab
                                                             anger
## 6.607432 3.522468 6.171025 4.748754 5.333107 3.165641 3.902000
sapply(cdi.dat, sd, na.rm=TRUE)
## country collact
                         arab
                                     id
                                             eff
## 3.501851 3.053585 3.084227 3.178040 2.719451 1.667729 2.820809
cor(cdi.dat,use="complete.obs")
##
                                                         id
                                                                     eff
               country
                           collact
                                           arab
## country 1.00000000 -0.01179741 -0.08342191 -0.01700412 -0.031781936
## collact -0.01179741 1.00000000
                                    0.39787729
                                                0.62820863 0.161226858
## arab
           -0.08342191 0.39787729
                                    1.00000000
                                                0.45077795 0.101235956
```

```
## eff
           -0.03178194 0.16122686 0.10123596 0.18461145 1.000000000
## sdo
           0.01527249 -0.15384594 -0.19150921 -0.25454925 -0.077079998
## anger
           0.06099956 0.34585402 0.22414986 0.35094051 0.008137974
##
                   sdo
                              anger
## country 0.01527249
                       0.060999563
## collact -0.15384594
                       0.345854020
## arab
          -0.19150921
                       0.224149858
## id
           -0.25454925
                       0.350940509
## eff
          -0.07708000 0.008137974
## sdo
           1.00000000 -0.113507365
## anger
          -0.11350736 1.000000000
```

ANOVA

```
cdi.aov <- aov(arab~as.factor(country))</pre>
summary(cdi.aov)
                        Df Sum Sq Mean Sq F value
                                                    Pr(>F)
                                    68.22
## as.factor(country)
                        11
                              750
                                            7.537 1.03e-12 ***
                            12700
## Residuals
                      1403
                                     9.05
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## 65 observations deleted due to missingness
TukeyHSD(cdi.aov)
##
     Tukey multiple comparisons of means
##
       95% family-wise confidence level
##
## Fit: aov(formula = arab ~ as.factor(country))
##
## $`as.factor(country)`
##
                diff
                                        upr
                                                p adj
                                 0.81875473 0.9875442
## 2-1
         -0.48545084 -1.7896564
         -0.54611777 -1.8104936
## 3-1
                                 0.71825802 0.9609150
## 4-1
         -1.23291673 -2.8300101
                                 0.36417659 0.3236948
## 5-1
         0.37002853 -1.2438531
                                 1.98391020 0.9998475
## 6-1
         -0.37550668 -1.6449537
                                 0.89394032 0.9982873
## 7-1
         -1.04182841 -2.2198729
                                 0.13621609 0.1436457
## 8-1
         -2.41515082 -3.8602887 -0.97001295 0.0000035
         0.64686469 -1.0798088
## 9-1
                                 2.37353817 0.9868765
## 10-1
        -0.55826352 -1.9817968 0.86526975 0.9811221
## 11-1
        -1.74177507 -2.9230685 -0.56048168 0.0000977
        -1.06577899 -2.5064241 0.37486609 0.3918061
## 12-1
## 3-2
         -0.06066693 -1.2348455
                                 1.11351168 1.0000000
                                0.77921559 0.9082537
## 4-2
        -0.74746590 -2.2741474
## 5-2
         0.85547936 -0.6887562
                                 2.39971488 0.8106716
## 6-2
         0.10994416 -1.0696935
                                 1.28958180 1.0000000
        -0.55637758 -1.6370432 0.52428802 0.8749437
## 7-2
## 8-2
        -1.92969998 -3.2966209 -0.56277908 0.0002602
## 9-2
         1.13231552 -0.5294458 2.79407682 0.5274495
## 10-2
        -0.07281268 -1.4168723 1.27124694 1.0000000
## 11-2 -1.25632424 -2.3405306 -0.17211790 0.0085117
```

```
-0.58032816 -1.9424983
                                 0.78184201 0.9646365
         -0.68679896 -2.1795985
## 4-3
                                 0.80600053 0.9390920
## 5-3
          0.91614630 -0.5946010
                                  2.42689357 0.7036730
                                  1.30605768 0.9999980
## 6-3
          0.17061109 -0.9648355
## 7-3
         -0.49571064 -1.5279568
                                 0.53653551 0.9189776
## 8-3
         -1.86903305 -3.1980051 -0.54006099 0.0002823
## 9-3
          1.19298246 -0.4377059
                                 2.82367083 0.4101076
## 10-3
         -0.01214575 -1.3175921
                                 1.29330063 1.0000000
## 11-3
         -1.19565730 -2.2316097
                                -0.15970491 0.0090018
## 12-3
         -0.51966122 -1.8437464
                                 0.80442395 0.9810087
## 5-4
          1.60294526 -0.1955036
                                 3.40139411 0.1356677
## 6-4
          0.85741006 -0.6396871
                                  2.35450721 0.7748736
## 7-4
          0.19108832 -1.2293317
                                  1.61150835 0.9999994
## 8-4
         -1.18223408 -2.8309389
                                  0.46647076 0.4432092
## 9-4
          1.87978142 -0.0205358
                                  3.78009864 0.0556710
## 10-4
          0.67465322 -0.9551478
                                  2.30445418 0.9714315
## 11-4
         -0.50885834 -1.9319740
                                 0.91425737 0.9910416
## 12-4
          0.16713774 -1.4776305
                                  1.81190596 1.0000000
## 6-5
         -0.74553521 -2.2605292
                                 0.76945881 0.9052416
## 7-5
         -1.41185694 -2.8511276
                                 0.02741372 0.0603524
## 8-5
         -2.78517935 -4.4501522 -1.12020648 0.0000034
## 9-5
          0.27683616 -1.6376122
                                 2.19128451 0.9999986
         -0.92829205 -2.5745478
## 10-5
                                 0.71796375 0.7921305
         -2.11180360 -3.5537347 -0.66987251 0.0001144
## 11-5
## 12-5
         -1.43580752 -3.0968823
                                 0.22526727 0.1686513
## 7-6
         -0.66632173 -1.7047733
                                 0.37212984 0.6225483
## 8-6
         -2.03964414 -3.3734418 -0.70584645 0.0000402
## 9-6
          1.02237136 -0.6122522
                                 2.65699491 0.6604823
## 10-6
                                 1.12760181 0.9999991
         -0.18275684 -1.4931155
## 11-6
         -1.36626840 -2.4084041 -0.32413265 0.0011423
## 12-6
         -0.69027231 -2.0192009
                                 0.63865624 0.8682147
## 8-7
         -1.37332241 -2.6204429 -0.12620194 0.0168378
## 9-7
          1.68869310
                     0.1239927
                                 3.25339349 0.0215854
## 10-7
          0.48356489 -0.7384551
                                  1.70558493 0.9797514
         -0.69994666 -1.6285693
                                 0.22867602 0.3615173
         -0.02395058 -1.2658621
                                  1.21796095 1.0000000
## 12-7
## 9-8
          3.06201550
                      1.2874953
                                  4.83653570 0.0000013
## 10-8
          1.85688730 0.3756825
                                 3.33809207 0.0025180
          0.67337574 -0.5768141
                                  1.92356561 0.8376786
## 11-8
          1.34937183 -0.1482860
## 12-8
                                 2.84702962 0.1250665
         -1.20512821 -2.9620987
                                 0.55184232 0.5166355
         -2.38863976 -3.9557877
                                -0.82149186 0.0000436
## 11-9
## 12-9
         -1.71264368 -3.4835069
                                 0.05821959 0.0689500
## 11-10 -1.18351156 -2.4086639
                                 0.04164076 0.0696833
## 12-10 -0.50751547 -1.9843372
                                 0.96930624 0.9935816
## 12-11 0.67599608 -0.5689977
                                 1.92098986 0.8302532
```

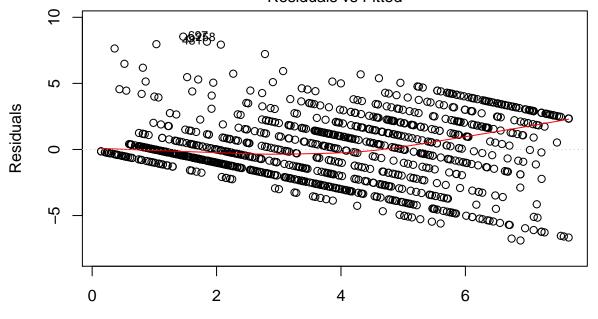
Regression

```
cdi.reg <- lm(collact~anger+eff+id)
summary(cdi.reg)</pre>
```

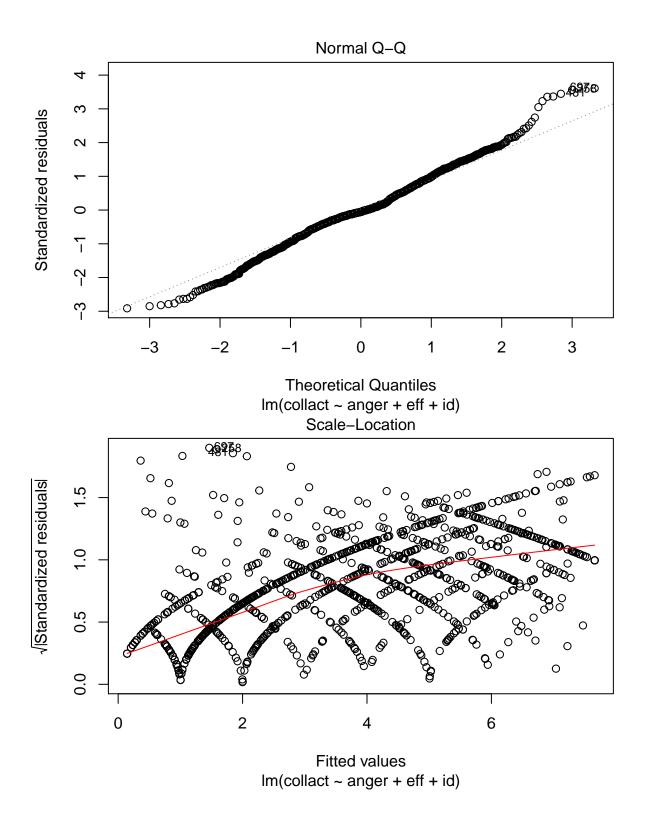
```
##
## Call:
  lm(formula = collact ~ anger + eff + id)
##
##
  Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
   -6.8856 -1.3022 -0.1279 1.4736
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
               0.01847
   (Intercept)
                           0.19083
                                     0.097
                                             0.9229
                0.15482
                           0.02634
                                     5.878 5.48e-09 ***
  anger
                0.06252
                           0.02678
                                     2.334
## eff
                                             0.0198 *
                0.54678
                           0.02413 22.657 < 2e-16 ***
## id
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.37 on 1107 degrees of freedom
     (369 observations deleted due to missingness)
## Multiple R-squared: 0.4153, Adjusted R-squared: 0.4137
## F-statistic: 262.1 on 3 and 1107 DF, p-value: < 2.2e-16
```

plot(cdi.reg)

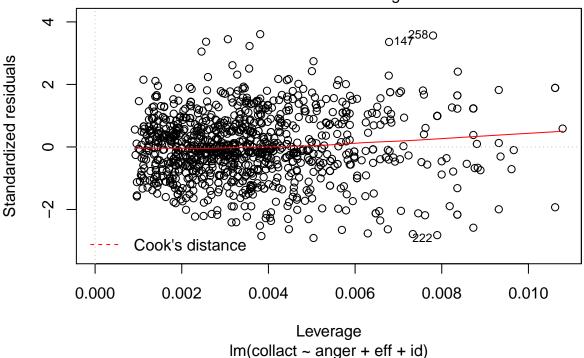
Residuals vs Fitted



Fitted values Im(collact ~ anger + eff + id)



Residuals vs Leverage



Structural Equation Modeling

```
library(lavaan)
## This is lavaan 0.5-20
## lavaan is BETA software! Please report any bugs.
library(semPlot)
path.model <- '
  # regressions
    collact ~ anger + eff + arab + sdo + id
    anger ~ sdo + id
    eff ~ sdo + id
    arab ~ sdo + id
path.fitted<-sem(path.model, data=cdi.dat, fixed.x=FALSE)</pre>
## Found more than one class "Model" in cache; using the first, from namespace 'lavaan'
summary(path.fitted, fit.measures=TRUE, standardized=TRUE)
## lavaan (0.5-20) converged normally after 22 iterations
##
##
                                                      Used
                                                                  Total
##
     Number of observations
                                                       1099
                                                                   1480
##
##
     Estimator
                                                         ML
```

## ##	Degrees of freed		11.374								
##	P-value (Chi-square) 0.010										
	Model test baseline model:										
##											
##	Minimum Function		1071.344								
##	0	lom			14						
##	P-value			0.000							
##	Hann madal manna basalina madal.										
##	User model versus baseline model:										
##	Comparative Fit		0.992								
##	=		0.963								
##											
	Loglikelihood and Information Criteria:										
##											
## ##	0		5379.347								
##	5										
##	Number of free p	parameters			18						
##					0794.694						
##	Bayesian (BIC)				0884.733						
##	Sample-size adjusted Bayesian (BIC) 30827.560										
##											
##	Root Mean Square Error of Approximation:										
##	RMSEA				0.050						
##	90 Percent Confidence Interval			0.02							
##	P-value RMSEA <=		0.431								
##											
	Standardized Root	Mean Squar	e Residua	1:							
##	SRMR				0.019						
##	Dittill				0.013						
##	Parameter Estimate	es:									
##											
##	Information				Expected						
##	Standard Errors				Standard						
## ##	Regressions:										
##	nogrobbions.	Estimate	Std.Err	Z-value	P(> z)	Std.lv	Std.all				
##	collact ~										
##	anger	0.146	0.026	5.630	0.000	0.146	0.137				
##	eff	0.061	0.026	2.290	0.022	0.061	0.053				
##	arab	0.135	0.026	5.214	0.000	0.135	0.134				
## ##	sdo id	0.042 0.497	0.044 0.027	0.959 18.674	0.338 0.000	0.042 0.497	0.023 0.516				
##	anger ~	0.491	0.021	10.074	0.000	0.491	0.510				
##	sdo	-0.045	0.051	-0.885	0.376	-0.045	-0.026				
##	id	0.310	0.026	11.794	0.000	0.310	0.344				
##	eff ~										
##	sdo	-0.052	0.050	-1.050	0.294	-0.052	-0.032				
##	id	0.149	0.026	5.758	0.000	0.149	0.176				

##	arab ~						
##	sdo	-0.151	0.051	-2.960	0.003	-0.151	-0.082
##	id	0.411	0.027	15.501	0.000	0.411	0.430
##							
##	Covariances:						
##		Estimate	Std.Err	Z-value	P(> z)	Std.lv	Std.all
##	sdo ~~						
##	id	-1.367	0.167	-8.178	0.000	-1.367	-0.255
##							
##	Variances:						
##		Estimate	Std.Err	Z-value	P(> z)	Std.lv	Std.all
##	collact	5.470	0.233	23.441	0.000	5.470	0.571
##	anger	7.353	0.314	23.441	0.000	7.353	0.876
##	eff	7.112	0.303	23.441	0.000	7.112	0.965
##	arab	7.465	0.318	23.441	0.000	7.465	0.790
##	sdo	2.792	0.119	23.441	0.000	2.792	1.000
##	id	10.324	0.440	23.441	0.000	10.324	1.000

semPaths(path.fitted)

