

/

1.2_FactorAnalysis

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R Markdown

This is an R Markdown document displaying the code and output for the cfa and glmm's ran for valence and arousal for two image sets.

This results in the following (clickable) structure

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General code

Used to load and prepare dataframes

```
##### Set environment #####
rm(list = ls()) # Clear environment
cat("\014") # Clear console
dev.off() # Clear plot window
options(contrasts=c("contr.sum", "contr.poly")) # Set contrast settings to effect coding

# Libraries
library(arrow)
library(lavaan)
library(lavaanPlot)
library(psych)
library(ltm)
library(car)
library(ggplot2)
library(ggstatsplot)
library(Polychrome)

#GLM specific
library(lme4)
library(lmerTest)
library(emmeans)
library(effects)
```

```
##### Loading data #####
imageData <- as.data.frame(read_parquet("../loc_data/df_session_tot_cleaned.parquet"))

piscesData <- imageData[imageData$DB == 'PiSCES',]
radboudData <- imageData[imageData$DB == 'Radboud',]
marloesData <- imageData[imageData$DB == 'marloes',]
```

1.0. Pisces Dataset

1.1. Valence

```
##### Valence #####
piscesDataClean = piscesData[c("ID", "pic_name", "valence")]
piscesDataClean$pic_name = as.factor(piscesDataClean$pic_name)
piscesDataClean = reshape(piscesDataClean, idvar = "ID", timevar = "pic_name", direction = "wide")
piscesDataCronbachs = piscesDataClean[, 2:16]
```

1.1.1. Cronbach's Alpha

```
# Calculate Cronbach's alpha using alpha()
alphavar = psych::alpha(piscesDataCronbachs, check.keys = TRUE)
summary(alphavar)
```

```
##
## Reliability analysis
## raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
## 0.84 0.84 0.88 0.26 5.4 0.025 57 8 0.27
```

1.1.2. CFA

```
names(piscesDataClean)[2:16] = c("Picture_105", "Picture_82", "Picture_118", "Picture_65",  
"Picture_88", "Picture_87", "Picture_59", "Picture_93", "Picture_56", "Picture_81",  
"Picture_110", "Picture_96", "Picture_132", "Picture_80",  
"Picture_98" )  
  
HS.model <- 'pisces =~ Picture_105 + Picture_82 + Picture_118 + Picture_65 + Picture_88 + Pic  
ture_87 + Picture_59 + Picture_93 + Picture_56 + Picture_81 + Picture_110 + Picture_96 + Pict  
ure_132 + Picture_80 + Picture_98'
```

Fit and visualize

```
## lavaan 0.6-9 ended normally after 56 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters    30
##
##                               Used      Total
##      Number of observations        84        89
##
## Model Test User Model:
##
##      Test statistic                188.181
##      Degrees of freedom             90
##      P-value (Chi-square)           0.000
##
## Model Test Baseline Model:
##
##      Test statistic                466.939
##      Degrees of freedom            105
##      P-value                        0.000
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)    0.729
##      Tucker-Lewis Index (TLI)      0.684
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)  -4979.918
##      Loglikelihood unrestricted model (H1) -4885.827
##
##      Akaike (AIC)                  10019.835
##      Bayesian (BIC)                 10092.760
##      Sample-size adjusted Bayesian (BIC) 9998.124
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                          0.114
##      90 Percent confidence interval - lower 0.091
##      90 Percent confidence interval - upper 0.137
##      P-value RMSEA <= 0.05           0.000
##
## Standardized Root Mean Square Residual:
##
##      SRMR                           0.099
##
## Parameter Estimates:
##
##      Standard errors                Standard
##      Information                    Expected
##      Information saturated (h1) model Structured
##
## Latent Variables:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      pisces =~
##      Picture_105    5.297    1.451    3.651    0.000    5.297    0.407
##      Picture_82     4.740    1.734    2.733    0.006    4.740    0.311
##      Picture_118    8.769    1.328    6.603    0.000    8.769    0.673
```

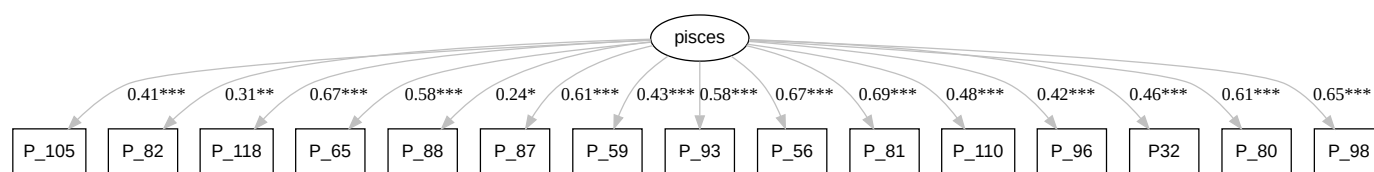
```

##      Picture_65      8.353      1.519      5.498      0.000      8.353      0.582
##      Picture_88      4.194      1.977      2.122      0.034      4.194      0.244
##      Picture_87     11.781      2.013      5.853      0.000     11.781      0.612
##      Picture_59      5.198      1.336      3.891      0.000      5.198      0.431
##      Picture_93      7.133      1.309      5.451      0.000      7.133      0.578
##      Picture_56      8.063      1.239      6.509      0.000      8.063      0.665
##      Picture_81      9.692      1.413      6.861      0.000      9.692      0.692
##      Picture_110      6.620      1.515      4.369      0.000      6.620      0.478
##      Picture_96      5.934      1.575      3.766      0.000      5.934      0.419
##      Picture_132      6.329      1.508      4.196      0.000      6.329      0.462
##      Picture_80      9.759      1.681      5.807      0.000      9.759      0.608
##      Picture_98      8.113      1.287      6.302      0.000      8.113      0.649
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .Picture_105    141.212    22.442    6.292    0.000    141.212    0.834
##      .Picture_82     209.994    32.918    6.379    0.000    209.994    0.903
##      .Picture_118      93.033    16.340    5.693    0.000     93.033    0.548
##      .Picture_65     136.514    22.772    5.995    0.000    136.514    0.662
##      .Picture_88     278.204    43.328    6.421    0.000    278.204    0.941
##      .Picture_87     231.777    39.210    5.911    0.000    231.777    0.625
##      .Picture_59     118.188    18.868    6.264    0.000    118.188    0.814
##      .Picture_93     101.667    16.930    6.005    0.000    101.667    0.666
##      .Picture_56      81.859    14.300    5.724    0.000     81.859    0.557
##      .Picture_81     101.959    18.198    5.603    0.000    101.959    0.520
##      .Picture_110     147.831    23.846    6.199    0.000    147.831    0.771
##      .Picture_96     165.490    26.356    6.279    0.000    165.490    0.825
##      .Picture_132     148.002    23.780    6.224    0.000    148.002    0.787
##      .Picture_80     162.325    27.408    5.923    0.000    162.325    0.630
##      .Picture_98      90.502    15.635    5.788    0.000     90.502    0.579
##      pisces           1.000

```

1.1.3. CFA Visualization

Pisces dataset - Valence



1.1.4. Distributions

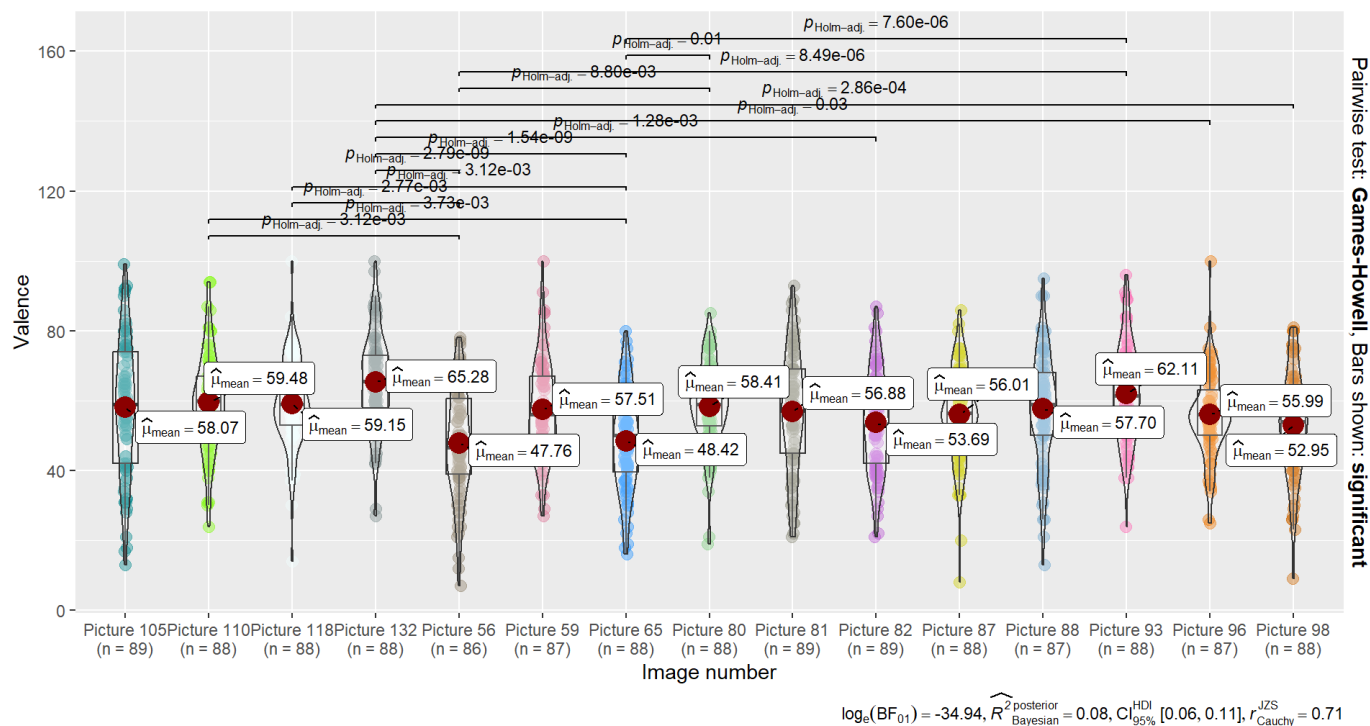
```
# Re-prep data
piscesDataClean = piscesData[c("ID", "pic_name", "valence")]
piscesDataClean$pic_name = as.factor(piscesDataClean$pic_name)
piscesDataClean$ID = as.factor(piscesDataClean$ID)
```

Visualizations

Pisces - Valence

Valence values

$F_{\text{Welch}}(14, 496.24) = 8.99, p = 1.18\text{e-}17, \hat{\omega}_p^2 = 0.18, \text{CI}_{95\%} [0.11, 1.00], n_{\text{obs}} = 1,318$



1.2. Arousal

Arousal

```
piscesDataClean = piscesData[c("ID", "pic_name", "arousal")]
piscesDataClean$pic_name = as.factor(piscesDataClean$pic_name)
piscesDataClean = reshape(piscesDataClean, idvar = "ID", timevar = "pic_name", direction = "wide")
piscesDataCronbachs = piscesDataClean[, 2:16]
```

1.2.1. Cronbach's Alpha

```
# Calculate Cronbach's alpha using alpha()
alphavar = psych::alpha(piscesDataCronbachs, check.keys = TRUE)
summary(alphavar)
```

```
##
## Reliability analysis
## raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
## 0.94 0.94 0.95 0.49 14 0.01 48 14 0.51
```

1.2.2. CFA

```
names(piscesDataClean)[2:16] = c("Picture_105", "Picture_82", "Picture_118", "Picture_65",
"Picture_88", "Picture_87", "Picture_59", "Picture_93", "Picture_56", "Picture_81",
"Picture_110", "Picture_96", "Picture_132", "Picture_80",
"Picture_98" )

HS.model <- 'pisces =~ Picture_105 + Picture_82 + Picture_118 + Picture_65 + Picture_88 + Pic
ture_87 + Picture_59 + Picture_93 + Picture_56 + Picture_81 + Picture_110 + Picture_96 + Pict
ure_132 + Picture_80 + Picture_98'
```

Fit and visualize

```
## lavaan 0.6-9 ended normally after 19 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 30
##
## Used Total
## Number of observations 84 89
##
## Model Test User Model:
##
## Test statistic 193.015
## Degrees of freedom 90
## P-value (Chi-square) 0.000
##
## Model Test Baseline Model:
##
## Test statistic 858.041
## Degrees of freedom 105
## P-value 0.000
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.863
## Tucker-Lewis Index (TLI) 0.840
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -5201.631
## Loglikelihood unrestricted model (H1) -5105.123
##
## Akaike (AIC) 10463.261
## Bayesian (BIC) 10536.186
## Sample-size adjusted Bayesian (BIC) 10441.550
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.117
## 90 Percent confidence interval - lower 0.094
## 90 Percent confidence interval - upper 0.139
## P-value RMSEA <= 0.05 0.000
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.070
##
## Parameter Estimates:
##
## Standard errors Standard
## Information Expected
## Information saturated (h1) model Structured
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##
## pisces =~
## Picture_105 11.707 2.013 5.816 0.000 11.707 0.591
## Picture_82 16.310 1.828 8.923 0.000 16.310 0.812
## Picture_118 15.903 2.012 7.904 0.000 15.903 0.747
```



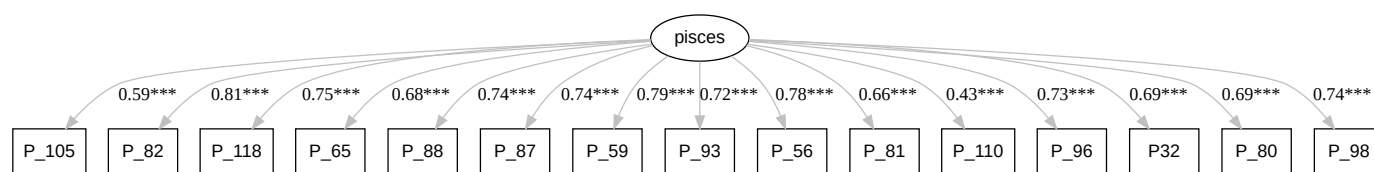
```

##      Picture_65      13.560      1.953      6.944      0.000      13.560      0.680
##      Picture_88      14.346      1.846      7.771      0.000      14.346      0.738
##      Picture_87      13.571      1.747      7.770      0.000      13.571      0.738
##      Picture_59      16.185      1.898      8.528      0.000      16.185      0.788
##      Picture_93      14.186      1.891      7.502      0.000      14.186      0.720
##      Picture_56      15.444      1.835      8.415      0.000      15.444      0.781
##      Picture_81      12.237      1.831      6.682      0.000      12.237      0.660
##      Picture_110       7.739      1.935      4.000      0.000       7.739      0.427
##      Picture_96      13.904      1.818      7.648      0.000      13.904      0.730
##      Picture_132      13.627      1.914      7.121      0.000      13.627      0.693
##      Picture_80      13.176      1.872      7.039      0.000      13.176      0.687
##      Picture_98      14.812      1.906      7.772      0.000      14.812      0.738
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .Picture_105    255.417   40.745   6.269   0.000   255.417   0.651
##      .Picture_82     137.429   24.055   5.713   0.000   137.429   0.341
##      .Picture_118    199.787   33.412   5.979   0.000   199.787   0.441
##      .Picture_65     214.218   34.882   6.141   0.000   214.218   0.538
##      .Picture_88     171.575   28.568   6.006   0.000   171.575   0.455
##      .Picture_87     153.637   25.580   6.006   0.000   153.637   0.455
##      .Picture_59     160.072   27.447   5.832   0.000   160.072   0.379
##      .Picture_93     187.074   30.896   6.055   0.000   187.074   0.482
##      .Picture_56     152.821   26.069   5.862   0.000   152.821   0.391
##      .Picture_81     194.121   31.433   6.176   0.000   194.121   0.565
##      .Picture_110    267.851   41.901   6.392   0.000   267.851   0.817
##      .Picture_96     169.447   28.105   6.029   0.000   169.447   0.467
##      .Picture_132    201.360   32.925   6.116   0.000   201.360   0.520
##      .Picture_80     194.607   31.758   6.128   0.000   194.607   0.529
##      .Picture_98     182.901   30.454   6.006   0.000   182.901   0.455
##      pisces          1.000

```

1.2.3. CFA Visualization

Pisces dataset - Arousal



1.2.4. Distributions

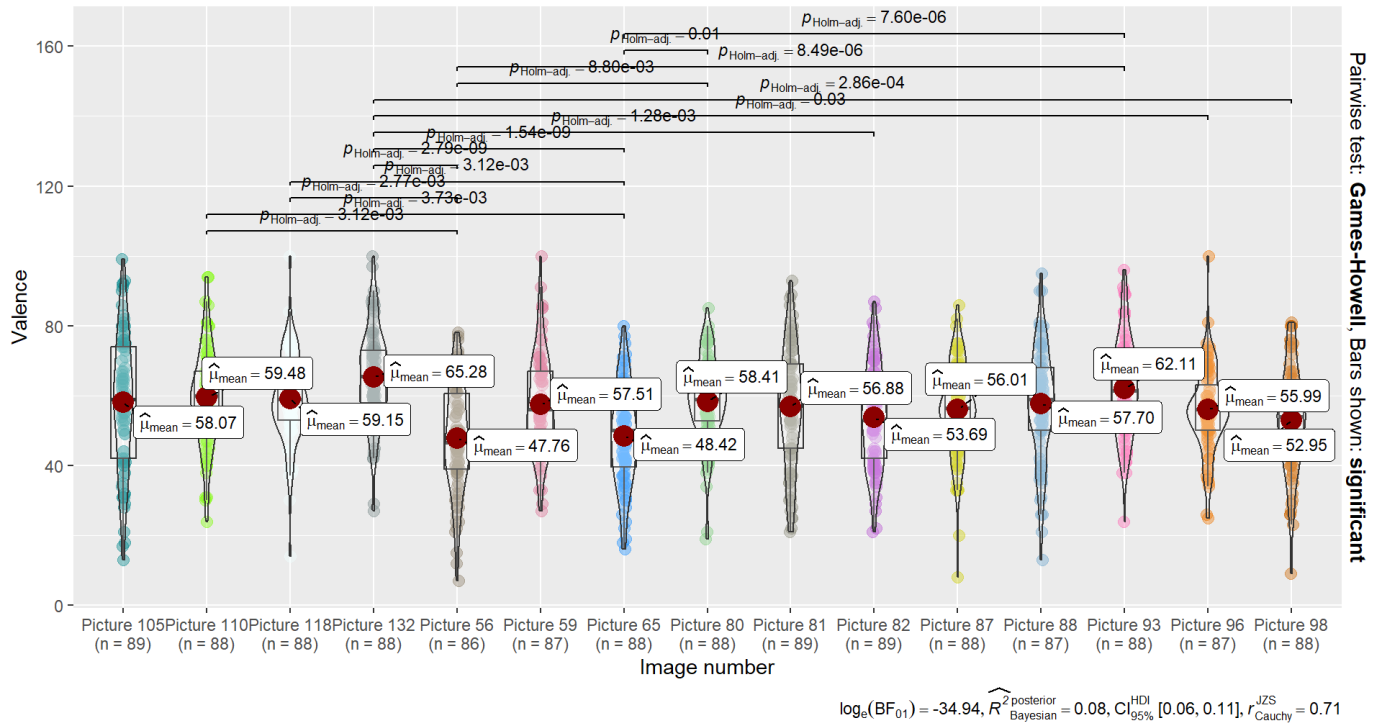
```
# Re-prep data
piscesDataClean = piscesData[c("ID", "pic_name", "valence")]
piscesDataClean$pic_name = as.factor(piscesDataClean$pic_name)
piscesDataClean$ID = as.factor(piscesDataClean$ID)
```

Visualizations

Pisces - Arousal

Valence values

$F_{\text{Welch}}(14, 496.24) = 8.99, p = 1.18\text{e-}17, \hat{\omega}_p^2 = 0.18, \text{CI}_{95\%} [0.11, 1.00], n_{\text{obs}} = 1,318$



2.0. Radboud faces

2.1. Valence

```
##### Valence #####
radboudDataClean = radboudData[c("ID", "pic_name", "valence")]
radboudDataClean$pic_name = as.factor(radboudDataClean$pic_name)
radboudDataClean = reshape(radboudDataClean, idvar = "ID", timevar = "pic_name", direction = "wide")
radboudDataCronbachs = radboudDataClean[, 2:16]
```

2.1.1. Cronbach's Alpha

```
# Calculate Cronbach's alpha using alpha()
alphavar = psych::alpha(radboudDataCronbachs, check.keys = TRUE)
summary(alphavar)
```

```
##
## Reliability analysis
## raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
## 0.89 0.89 0.91 0.36 8.3 0.017 51 8.6 0.35
```

2.1.2. CFA

```
names(radboudDataClean)[2:16] = c('Face_01', 'Face_36', 'Face_32', 'Face_61', 'Face_04', 'Face_24', 'Face_02', 'Face_49', 'Face_58', 'Face_46', 'Face_05', 'Face_33', 'Face_57', 'Face_47', 'Face_27')
```

```
HS.model <- 'radboud =~ Face_01 + Face_36 + Face_32 + Face_61 + Face_04 + Face_24 + Face_02 + Face_49 + Face_58 + Face_46 + Face_05 + Face_33 + Face_57 + Face_47 + Face_27'
```

Fit and visualize

```
## lavaan 0.6-9 ended normally after 20 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 30
##
## Used Total
## Number of observations 85 89
##
## Model Test User Model:
##
## Test statistic 174.182
## Degrees of freedom 90
## P-value (Chi-square) 0.000
##
## Model Test Baseline Model:
##
## Test statistic 571.377
## Degrees of freedom 105
## P-value 0.000
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.819
## Tucker-Lewis Index (TLI) 0.789
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -4927.772
## Loglikelihood unrestricted model (H1) -4840.681
##
## Akaike (AIC) 9915.544
## Bayesian (BIC) 9988.824
## Sample-size adjusted Bayesian (BIC) 9894.180
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.105
## 90 Percent confidence interval - lower 0.081
## 90 Percent confidence interval - upper 0.128
## P-value RMSEA <= 0.05 0.000
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.078
##
## Parameter Estimates:
##
## Standard errors Standard
## Information Expected
## Information saturated (h1) model Structured
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## radboud =~
## Face_01 7.066 1.485 4.757 0.000 7.066 0.505
## Face_36 7.284 1.263 5.767 0.000 7.284 0.594
## Face_32 8.577 1.308 6.556 0.000 8.577 0.658
```

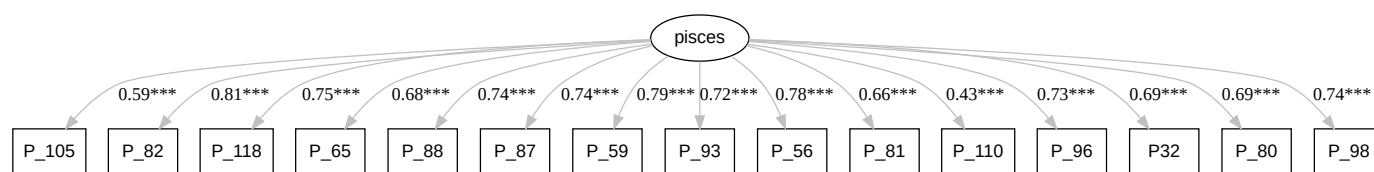
```

##      Face_61      7.407      1.319      5.617      0.000      7.407      0.581
##      Face_04      8.736      1.527      5.723      0.000      8.736      0.590
##      Face_24      7.528      1.344      5.600      0.000      7.528      0.580
##      Face_02     10.139      1.364      7.433      0.000     10.139      0.723
##      Face_49      9.735      1.498      6.499      0.000      9.735      0.653
##      Face_58      8.523      1.404      6.070      0.000      8.523      0.619
##      Face_46      7.598      1.506      5.045      0.000      7.598      0.531
##      Face_05      7.625      1.377      5.537      0.000      7.625      0.575
##      Face_33      9.031      1.364      6.620      0.000      9.031      0.663
##      Face_57      6.207      1.432      4.334      0.000      6.207      0.466
##      Face_47      9.368      1.350      6.941      0.000      9.368      0.687
##      Face_27      7.324      1.228      5.962      0.000      7.324      0.610
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .Face_01    145.570   23.206    6.273    0.000   145.570    0.745
##      .Face_36     97.307   15.881    6.127    0.000    97.307    0.647
##      .Face_32     96.528   16.166    5.971    0.000    96.528    0.568
##      .Face_61    107.425   17.461    6.152    0.000   107.425    0.662
##      .Face_04    142.697   23.260    6.135    0.000   142.697    0.652
##      .Face_24    111.813   18.166    6.155    0.000   111.813    0.664
##      .Face_02     94.130   16.420    5.733    0.000    94.130    0.478
##      .Face_49    127.330   21.279    5.984    0.000   127.330    0.573
##      .Face_58    116.898   19.251    6.072    0.000   116.898    0.617
##      .Face_46    146.669   23.518    6.237    0.000   146.669    0.718
##      .Face_05    117.988   19.138    6.165    0.000   117.988    0.670
##      .Face_33    104.206   17.496    5.956    0.000   104.206    0.561
##      .Face_57    139.054   22.001    6.320    0.000   139.054    0.783
##      .Face_47     98.264   16.722    5.876    0.000    98.264    0.528
##      .Face_27     90.419   14.840    6.093    0.000    90.419    0.628
##      radboud       1.000

```

2.1.3. CFA Visualization

Radboud dataset - Valence



2.1.4. Distributions

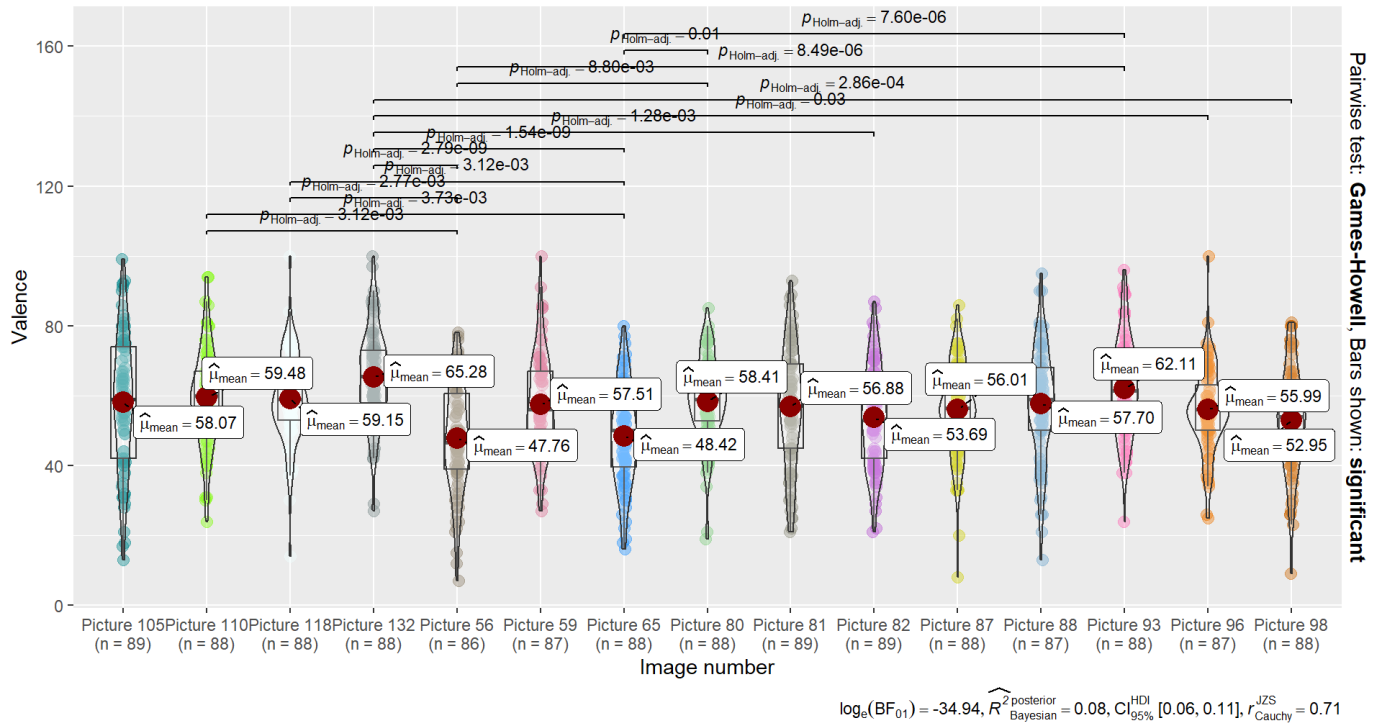
```
# Re-prep data
piscesDataClean = piscesData[c("ID", "pic_name", "valence")]
piscesDataClean$pic_name = as.factor(piscesDataClean$pic_name)
piscesDataClean$ID = as.factor(piscesDataClean$ID)
```

Visualizations

Pisces - Valence

Valence values

$F_{\text{Welch}}(14, 496.24) = 8.99, p = 1.18\text{e-}17, \hat{\omega}_p^2 = 0.18, \text{CI}_{95\%} [0.11, 1.00], n_{\text{obs}} = 1,318$



2.2. Arousal

```
##### Valence #####
radboudDataClean = radboudData[c("ID", "pic_name", "arousal")]
radboudDataClean$pic_name = as.factor(radboudDataClean$pic_name)
radboudDataClean = reshape(radboudDataClean, idvar = "ID", timevar = "pic_name", direction =
"wide")
radboudDataCronbachs = radboudDataClean[, 2:16]
```

2.2.1. Cronbach's Alpha

```
# Calculate Cronbach's alpha using alpha()
alphavar = psych::alpha(radboudDataCronbachs, check.keys = TRUE)
summary(alphavar)
```

```
##
## Reliability analysis
## raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
## 0.95 0.95 0.96 0.57 20 0.0075 36 14 0.56
```

2.2.2. CFA

```
names(radboudDataClean)[2:16] = c('Face_01', 'Face_36', 'Face_32', 'Face_61', 'Face_04', 'Face_24',
'Face_02', 'Face_49', 'Face_58', 'Face_46', 'Face_05', 'Face_33', 'Face_57', 'Face_47', 'Face_27')

HS.model <- 'radboud =~ Face_01 + Face_36 + Face_32 + Face_61 + Face_04 + Face_24 + Face_02 +
Face_49 + Face_58 + Face_46 + Face_05 + Face_33 + Face_57 + Face_47 + Face_27'
```


Fit and visualize

```
## lavaan 0.6-9 ended normally after 17 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters    30
##
##                               Used      Total
##      Number of observations        85        89
##
## Model Test User Model:
##
##      Test statistic                222.273
##      Degrees of freedom             90
##      P-value (Chi-square)           0.000
##
## Model Test Baseline Model:
##
##      Test statistic                1087.748
##      Degrees of freedom             105
##      P-value                        0.000
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)    0.865
##      Tucker-Lewis Index (TLI)      0.843
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)   -5070.572
##      Loglikelihood unrestricted model (H1) -4959.436
##
##      Akaike (AIC)                   10201.145
##      Bayesian (BIC)                  10274.424
##      Sample-size adjusted Bayesian (BIC) 10179.780
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                          0.131
##      90 Percent confidence interval - lower 0.110
##      90 Percent confidence interval - upper 0.153
##      P-value RMSEA <= 0.05           0.000
##
## Standardized Root Mean Square Residual:
##
##      SRMR                          0.062
##
## Parameter Estimates:
##
##      Standard errors                Standard
##      Information                    Expected
##      Information saturated (h1) model Structured
##
## Latent Variables:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      radboud =~
##      Face_01    13.568    1.776    7.640    0.000    13.568    0.723
##      Face_36    13.139    1.701    7.724    0.000    13.139    0.729
##      Face_32    14.518    1.659    8.753    0.000    14.518    0.796
```

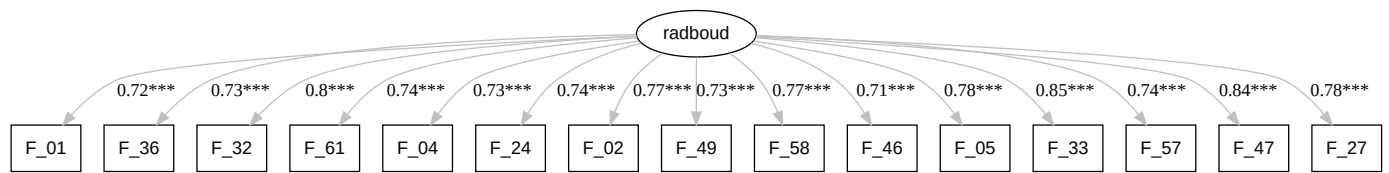
```

##      Face_61      14.030      1.776      7.901      0.000      14.030      0.741
##      Face_04      13.858      1.790      7.743      0.000      13.858      0.730
##      Face_24      13.351      1.706      7.827      0.000      13.351      0.736
##      Face_02      13.987      1.668      8.387      0.000      13.987      0.773
##      Face_49      12.272      1.577      7.780      0.000      12.272      0.733
##      Face_58      13.383      1.589      8.420      0.000      13.383      0.775
##      Face_46      13.872      1.852      7.490      0.000      13.872      0.713
##      Face_05      13.171      1.561      8.435      0.000      13.171      0.776
##      Face_33      15.258      1.575      9.687      0.000      15.258      0.850
##      Face_57      13.971      1.773      7.882      0.000      13.971      0.740
##      Face_47      14.586      1.535      9.504      0.000      14.586      0.840
##      Face_27      14.357      1.677      8.559      0.000      14.357      0.784
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .Face_01      167.812      27.096      6.193      0.000      167.812      0.477
##      .Face_36      152.107      24.605      6.182      0.000      152.107      0.468
##      .Face_32      122.170      20.341      6.006      0.000      122.170      0.367
##      .Face_61      161.597      26.245      6.157      0.000      161.597      0.451
##      .Face_04      167.974      27.182      6.179      0.000      167.974      0.467
##      .Face_24      150.716      24.436      6.168      0.000      150.716      0.458
##      .Face_02      131.867      21.695      6.078      0.000      131.867      0.403
##      .Face_49      129.747      21.014      6.174      0.000      129.747      0.463
##      .Face_58      119.136      19.620      6.072      0.000      119.136      0.399
##      .Face_46      186.223      29.977      6.212      0.000      186.223      0.492
##      .Face_05      114.656      18.891      6.069      0.000      114.656      0.398
##      .Face_33      89.525      15.584      5.745      0.000      89.525      0.278
##      .Face_57      161.447      26.209      6.160      0.000      161.447      0.453
##      .Face_47      88.986      15.323      5.807      0.000      88.986      0.295
##      .Face_27      129.495      21.419      6.046      0.000      129.495      0.386
##      radboud          1.000

```

2.2.3. CFA Visualization

Radboud dataset - Arousal



2.2.4. Distributions

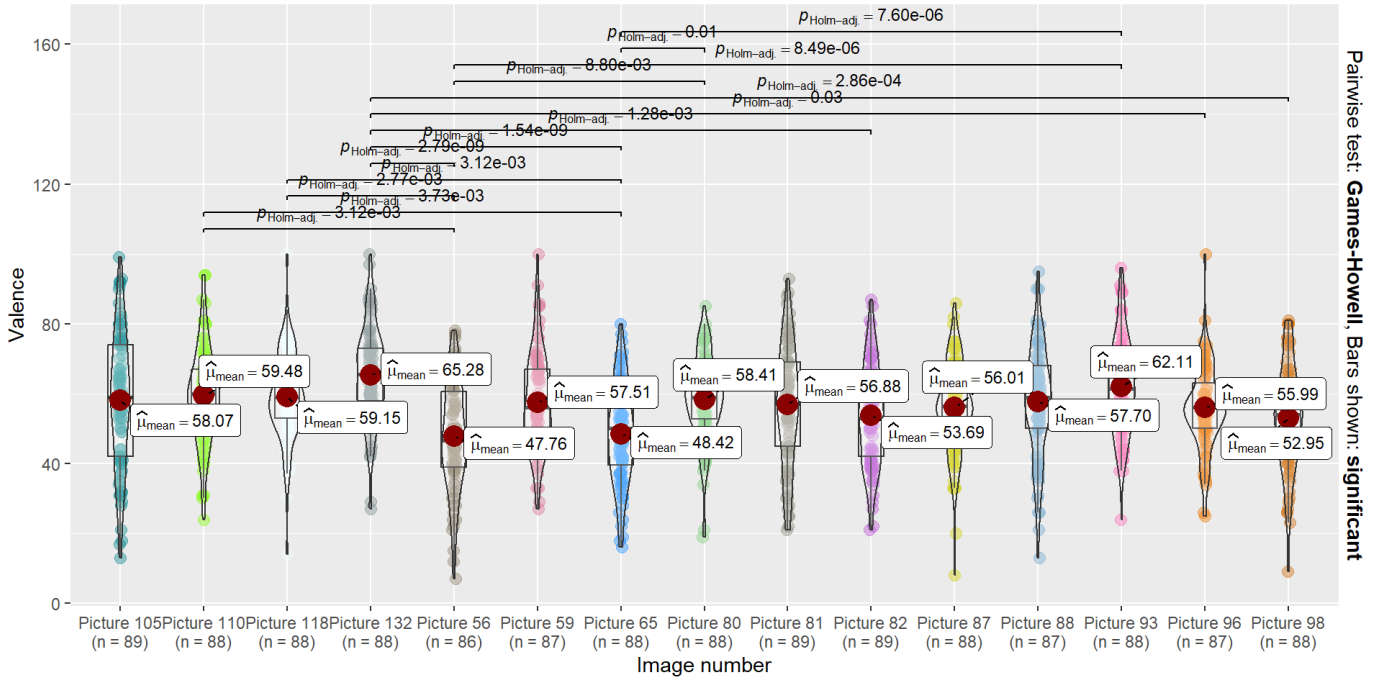
```
# Re-prep data
piscesDataClean = piscesData[c("ID", "pic_name", "valence")]
piscesDataClean$pic_name = as.factor(piscesDataClean$pic_name)
piscesDataClean$ID = as.factor(piscesDataClean$ID)
```

Visualizations

Pisces - Valence

Valence values

$F_{\text{Welch}}(14, 496.24) = 8.99, p = 1.18\text{e-}17, \hat{\omega}_p^2 = 0.18, \text{CI}_{95\%} [0.11, 1.00], n_{\text{obs}} = 1,318$



$\log_e(\text{BF}_{01}) = -34.94, \hat{R}^2_{\text{posterior Bayesian}} = 0.08, \text{CI}_{95\%}^{\text{HDI}} [0.06, 0.11], r^{\text{JZS}}_{\text{Cauchy}} = 0.71$