feedback_response

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Your questions are in quotes. Our answers follow.

"1. Biojtscore and BioJftotal are measuring the same (number of biocentric justifications the participant endorsed), the only difference is that the former was obtained from an open-ended question and the other from a close-ended question. Therefore, I think it would be better that they both have the same transformation into levels (e.g. level1=0,1,2; level2=3; level3=4), unless there was a mathematical reason to choose the levels."

We understand that Biojtscore and BioJftotal are similar, and we also tried the same level transformation (level1=0,1,2; level2=3; level3=4) to them. However, the result of BioJftotal shows that level 2 has 0 values in it, which means there is no value being predicted as 2.

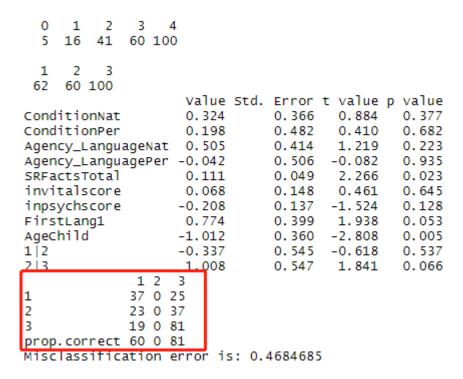


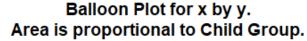
Figure 1: (BioJftotal Model Fit Result)

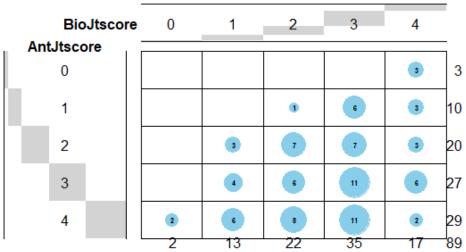
Level 2 is less meaningful in this case, and the model does not fit the data properly if we set the level transformation in this way. Thus, we switched the way of transforming BioJftotal.

And we have a question for you. Since Biojtscore and BioJftotal are so similar, do you think it is reasonable to combine them together into one dependent variable (e.g. Bioscore = Biojtscore + BioJftotal)?

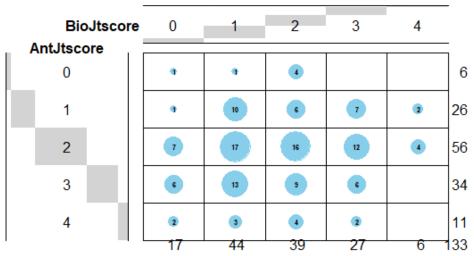
"2. Also, while I was looking at the results, I saw that children were more biocentric than adults but they were also more anthropocentric (both groups were able to provide more than one kind of justification for each situation). It seems that the two separate scores (Biojtscore and Antjtscore) don't let us see differences between those participants who only provided biocentric justifications and those who provided both antrhopocentric and biocentric justifications (for instance, children could be providing more justifications of different kind than adults but we cannot assure that with this analysis). Do you think it would be a good idea to integrate both scores, what alternatives do you know to deal with this kind of issue? I was thinking that we could calculate the proportion of Biocentric justifications over the total amount of justifications (Biocentric+antrhopocentric). Are there any limitations to this approach?"

It is hard to see the differences between individuals who provided both anthropocentric and biocentric justifications and who only provided one kind of justification by the barplot in the last report. Instead of applying the proportion method you mentioned and redoing the barplot, we created balloonplots for the Child group and Adult group separately.





Balloon Plot for x by y. Area is proportional to Adult Group.



On the balloonplots, the vertical axis represents AntJtscore and the horizontal axis is BioJtscore. You may notice the gray region behind each level. The size of the shaded region corresponds to the total frequency of each level for that variable. For instance, in the Child group, most children get score 3 for BioJtscore, so the shaded region for BioJtscore level 3 is greater than that of the other levels. The size of each blue dot matches the number of individuals who get the corresponding scores for BiojtScore and AntJtscore, for example, in the Adult group, there are only two individuals get BioJtscore 0 and AntJtscore 4.

In such a way, we can find that children have higher frequency of getting BioJtscore in range [2,3] and AntJtscore in range [2,4], and adults have higher frequency of getting BioJtscore in range [1,3] and AntJtscore in range [1,3]. To be specific, the Child group gets the highest frequency for BioJtscore at level 3 and AntJtscore at level 4 (AntJtscore 4 and 3 are close, level 4 only has two more individuals than level 3), which indicates children tend to get high score for both kinds of justifications. Meanwhile, the adult group gets the highest frequency for BioJtscore at level 1 and AntJtscore at level 2, which implies most of them do not have strong attitude for both kinds of justifications.

Please let us know if this is still not clear.

"3. You mentioned the possibility to do a Principal component analysis, how this could work? Can you explain more about it? Would it group agency language and condition as a factor?"

We did look into Principal Component Analysis (PCA) for your predictor variables. PCA is a way to construct new variables, "principal components", to represent several variables. PCA is mostly used in situations where there are a multitude of variables that can be combined into a smaller and more understandable set of variables (your data set does not suffer from a problem of a multitude of predictor variables). PCA is sometimes used for predictor variables that are correlated. But PCA is only used for continuous variables. And as such, we could not combined Condition and Agency_Language, and we've continued to run the models using just Condition (leaving Agency_Language out of the model).

The variables invitalscore and inpsychscore are also significantly correlated $(p = 5.0 * 10^{-4})$, with a correlation of cor.test(AC_df\$invitalscore, AC_df\$inpsychscore)\$estimate). Though a correlation below .5 is generally not a worry in terms of multicollinearity in a model, we did experiment with using

PCA to replace the two correlated variables with a "combined" variable. Something we could do because invitalscore and inpsychscore could both be considered to be continuous variables.

We then tried the linear model fit three ways. (1) Using both the invitalscore and inpsychscore in the model. (NOTE: We used the "normalized", not the "compressed" method for combining the adult and children's invitalscore and inpsychscore variables.) (2) By combining the two scores into a new variable postqs = invitalscore + inpsychscore. (3) By doing principal component analysis to combine invitalscore and inpsychscore into a new variable, postqca.

As you can see in the summary of the model fits, there is no difference in the $Adjusted.R^2$. And we've concluded that the choice of what to do with the phase 3 questionnaire comes down to what makes the most sense in the context of your experiment. Since invitalscore and inpsychscore are correlated we would suggest looking at the seventeen questions that were asked in the post questionnaire to see if it makes sense to combine them into one variable. Or if it makes to keep these as two separate variables, we can continue to do that.

Call:

```
lm(formula = MeanSever ~ Condition + SRFactsTotal + invitalscore +
inpsychscore + FirstLang + Age, data = AC_df)
```

Residuals:

```
Min 1Q Median 3Q Max
-1.68513 -0.44849 0.01683 0.41194 1.42652
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept)
              1.623603
                        0.182912
                                    8.876 2.81e-16 ***
ConditionNat 0.060888
                        0.109228
                                           0.5778
                                    0.557
ConditionPer 0.075715
                        0.111191
                                    0.681
                                            0.4966
SRFactsTotal 0.019409
                        0.015928
                                    1.219
                                            0.2243
invitalscore 0.001812
                        0.182530
                                    0.010
                                            0.9921
inpsychscore 0.036534
                        0.156307
                                    0.234
                                            0.8154
FirstLang1
            -0.286629
                        0.136141 -2.105
                                            0.0364 *
AgeChild
                                    5.338 2.39e-07 ***
             0.717942
                        0.134497
```

Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.6612 on 214 degrees of freedom Multiple R-squared: 0.1969, Adjusted R-squared: 0.1706 F-statistic: 7.494 on 7 and 214 DF, p-value: 4.551e-08

Call:

```
lm(formula = MeanSever ~ Condition + SRFactsTotal + postqs +
FirstLang + Age, data = AC_df)
```

Residuals:

```
Min 1Q Median 3Q Max -1.69202 -0.44600 0.01459 0.41722 1.42268
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept)
             1.59740
                        0.18291
                                  8.733 7.01e-16 ***
ConditionNat 0.05914
                        0.10894
                                  0.543
                                          0.5878
ConditionPer 0.07321
                        0.11011
                                  0.665
                                          0.5068
                                 1.232
SRFactsTotal 0.01949
                        0.01583
                                          0.2195
```

```
postas
             0.08651
                        0.17163
                                 0.504
                                          0.6147
FirstLang1
            -0.28637
                        0.13559 -2.112
                                          0.0358 *
AgeChild
             0.72821
                        0.10643
                                 6.842 7.99e-11 ***
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
Residual standard error: 0.6594 on 215 degrees of freedom
Multiple R-squared: 0.1975,
                               Adjusted R-squared: 0.1751
F-statistic: 8.819 on 6 and 215 DF, p-value: 1.333e-08
Call:
lm(formula = MeanSever ~ Condition + SRFactsTotal + postpca +
   FirstLang + Age, data = AC_df)
Residuals:
    Min
              1Q
                   Median
                                3Q
                                        Max
-1.68055 -0.44268 0.01771 0.41462 1.43161
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept)
             1.666554
                        0.189025
                                   8.817 4.07e-16 ***
ConditionNat 0.061168
                        0.108959
                                   0.561
                                           0.5751
ConditionPer 0.077673
                        0.110040
                                   0.706
                                           0.4810
                                           0.2261
SRFactsTotal 0.019215
                        0.015830
                                   1.214
postpca
            -0.009992
                        0.039490 -0.253
                                           0.8005
FirstLang1
            -0.287664
                        0.135626 -2.121
                                           0.0351 *
AgeChild
             0.729131
                        0.107378
                                   6.790 1.08e-10 ***
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.6597 on 215 degrees of freedom
Multiple R-squared: 0.1968,
                               Adjusted R-squared: 0.1744
F-statistic: 8.78 on 6 and 215 DF, p-value: 1.455e-08
```

"4. Regarding the Anova test, I saw that the model with age was significantly different than the model without it which is interesting. However, I'm not sure I'm interpreting the results in a right way, is this Anova comparing the between group variances (children vs. adults) with the within group variances for each dependent variable? Can you please describe more those results?"