#### Continuous Moderator Variables in Multiple Regression Analysis®

A moderator variable is one which alters the relationship between other variables. Suppose that we are using regression analysis to test the model that continuous variable Y is a linear function of continuous variable X, but we think that the slope for the regression of Y on X varies across levels of a moderator variable, M. Put another way, we think that there is an **interaction** between X and M with respect to their effect on Y.

If the moderator variable is categorical, we can conduct a "Potthoff analysis" to determine if the regression of Y on X differs across levels of the categorical moderator -- see my handout "Comparing Regression Lines From Independent Samples." Here I shall treat the moderator variable as a continuous variable. The data that we shall use are from the research project which is described at Misanthropy, Idealism, and Attitudes About Animals. Please do read that document to familiarize yourself with the research. Do note that the analysis presented in that document treats ethical idealism as a dichotomous moderator variable. This dichotomy was produced by my classifying cases with idealism scores greater than the median as being "idealistic" and those with scores less than or equal to the median as being "nonidealistic."

The attitude towards animal rights and animal research scale we used consisted of 28 items. An unrotated factor analysis on these 28 items indicated that there was only one factor on which all items loaded well. Item-total correlations were good, with Cronbach's alpha being .91. In other words, these 28 items were internally consistent, all measuring the same basic construct. A rotated factor analysis did indicate that two subscale scores could be created, one containing items reflecting concern with the violation of animal rights by using animals for food, clothing, and fur, and the other containing items reflecting concern with animal research. Here we shall employ the first subscale, which I shall refer to as the Animal Rights (AR) scale. After dropping from this subscale three items with relatively low item-total correlations, we have a 12 item scale with Cronbach's alpha being .87.

Run the program Moderate.sas, which is available at my <u>SAS Programs Page</u>. You will need to download the data file first. It is Moderate.dat, available at my <u>StatData</u> page and in SPSS format on my <u>SPSS Data Page</u>. Much of the output is available here.

Also useful reading is the section titled "Moderating Relationships" on pages 551-556 in David Howell's *Statistical Methods for Psychology*, 8<sup>th</sup> ed.

Look at the first page of output. Notice that there are no cases with missing data. If there were cases with missing data on variables to be used in the regression analysis I would delete them at this point, prior to standardizing the variables.

In the program note the use of Proc Standard to standardize the variables (in the output data set "Zs") to mean 0, standard deviation 1. Also note the creation of the interaction term in the Data Interaction step. It is not necessary to standardize the variables (or, according to Hayes (2013), to center any variables), but I find it easier to deal with standardized results.

The simple correlation matrix shows a significant relationship between misanthropy and support of animal rights, but our moderator variable, idealism, is not significantly related to either misanthropy or support of animal rights. While ethical idealism may not be related to support of animal rights here, it may moderate the relationship between misanthropy and support of animal rights.

-

<sup>©</sup> Copyright 2021, Karl L. Wuensch - All rights reserved.

It is commonly recommended that one **center** all of the variables involved in the interaction (in this case, misanthropy and idealism) -- that is, subtract from each score on each variable the mean of all scores on that variable -- to reduce multicollinearity and other problems. We could center the criterion variable too, if we wanted to interpret scores on it in terms of deviations of the score from the mean. Centering the outcome variable would affect the intercept but not the other regression coefficients. We could just standardize our variables to *z* scores, which are, of course, centered. If we standardize all of the variables, then our regression coefficients will also be standardized, that is, Beta weights. That is what I have done here. Rarely is the psychological researcher interested in unstandardized slopes.

It turns out that the recommendation to center variables to avoid problems with multicollinearity is BS (Bad Statistics), as explained on pages 282 through 290 of Hayes (2013). The test of the interaction is absolutely unaffected by such transformations. Centering may, however, be useful with respect to making some of the regression coefficients have more meaningful interpretations than they otherwise would have. See <a href="Effect of Centering and Standardization in Moderation Analysis">Effect of Centering and Standardization in Moderation Analysis</a>.

Our model is statistically significant,  $R^2$  = .113, p < .001, AR = -.02 + .303\*Misanth + .067\*Ideal – .146\*Interact – where all of the variables are standardized. Of most interest, the interaction term is statistically significant, p = .049, indicating that ethical idealism does function as a moderator of the relationship between misanthropy and support of animal rights. Yes, we could look at this interaction from another perspective -- that misanthropy serves as a moderator of the relationship between idealism and support of animal rights.

Here I find the simple (conditional) slopes for predicting AR from Misanth at each of three levels of Ideal:

```
Low Idealism (one standard deviation below the mean):
```

```
AR = -.02 + .303*Misanth + .067*Ideal - .146*Interact
```

AR = -.02 + .303\*Misanth + .067\*(-1) + (-.146)\*(-1)\*Misanth

AR = .449\*Misanth -.069

Medium Idealism (at the mean)

AR = -.02 + .303\*Misanth + .067\*Ideal - .146\*Interact

AR = -.02 + .303\*Misanth + .067\*(0) + (-.146)\*(0)\*Misanth

AR = .303\*Misanth - .02

High Idealism (one standard deviation above the mean)

AR = -.02 + .303\*Misanth + .067\*Ideal - .146\*Interact

AR = -.02 + .303\*Misanth + .067\*(1) + (-.146)\*(1)\*Misanth

AR = .157\*Misanth + .065

Now we shall prepare a plot to illustrate the moderator effect. First we need to obtain some points. I obtain predicted AR for low (-1), mean (0), and high (1) values of idealism paired with low (-1) and high (1) values of idealism.

```
Low Idealism: AR = .449*Misanth -.069
```

& Low Misanthropy: AR = .449(-1) - .069 = -.518

& High Misanthropy: AR = .449(1) - .069 = .380

Mean Idealism: AR = .303\*Misanth - .02

& Low Misanthropy: AR = .303(-1) - .02 = -.305

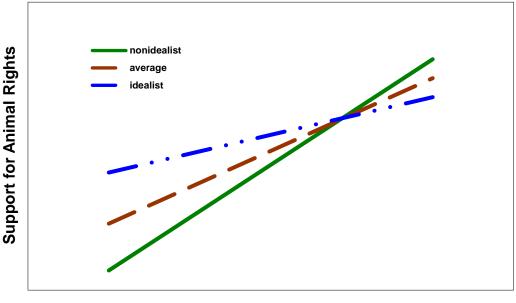
& High Misanthropy: AR = .303(1) - .02 = .301

High Idealism: AR = .157 \* Misanth + .065

& Low Misanthropy: AR = .157(-1) + .065 = -.092

& High Misanthropy: AR = .157(1) + .065 = .222

Here are the conditional regression lines, plotted with Microsoft Graph:

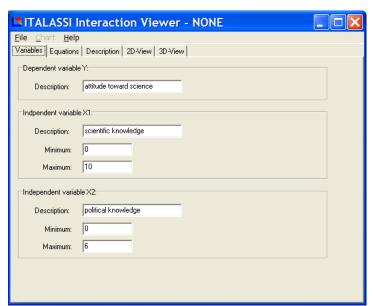


Misanthropy

#### Italassi

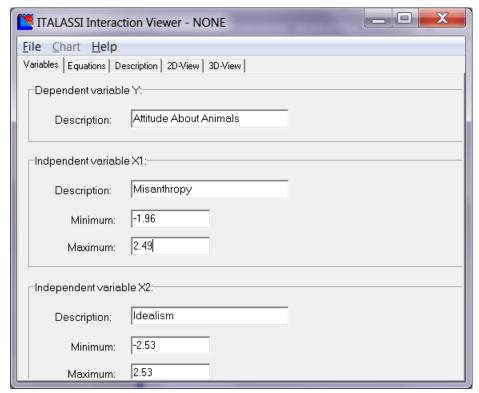
Italassi is a free program that will allow you to display the interaction on your computer screen with a plot that shows the relationship between the criterion variable and the predictor variable at various levels of the moderator variable. In the 2-D display, at any given moment the plot displays a single conditional regression line, but there is a slider which you can move to change the condition, the level of the moderator, and as you do so, the regression line changes. I would like you to try this software. First you need to download it to your personal computer. It is available at <a href="Provalis Research">Provalis</a> Research Simply download the file (ITAL32.EXE) and install it. By default it will install itself in the Program Files folder in a subfolder named "Provalis Research." You may want to create a desktop shortcut icon to the Italassi.exe file in that folder.

Once you have downloaded and installed Italassi, run it. Click OK on the opening window and you get a window like this:

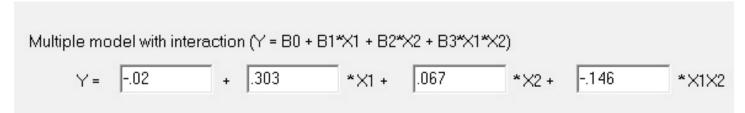


Notice that it already has variables labeled, for an example of a moderator effect. Under the Equations tab you can find regression coefficients for predicting the criterion variable from bivariate, trivariate, and interaction models. Under Description you can find a description of the example variables. Click the 2-D View tab and look at the regression line. Click on the slider and move it to see how the regression line changes as you change the level of the moderator variable.

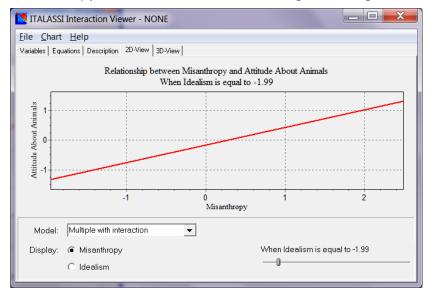
After playing with the example analysis a bit, click on the variables tab and enter the names of our centered variables and the lowest and highest values for the predictor (X<sub>1</sub>) and moderator (X<sub>2</sub>) variables, like this:

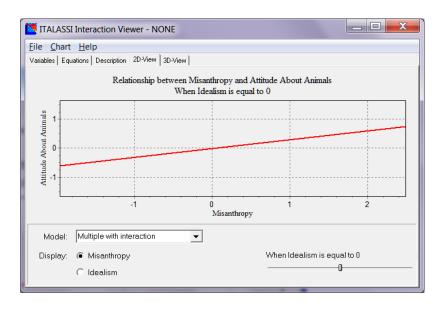


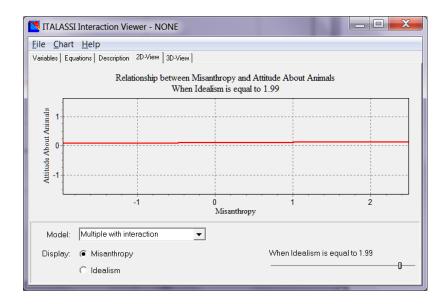
Next, click on the Equations tab and enter the regression coefficients for our interaction. Just clear out the coefficients for the reduced models, we don't have those coefficients (you can get them and use them if you wish).



Now click the 2D-View and move the slider so that you see how the relationship between misanthropy and attitude about animal rights changes as idealism moves from low to high.







#### **Process Hayes**

<u>Hayes' process analysis code</u> makes it easier to conduct a moderation analysis such that described above. While the data are still loaded in SAS, open and run the process macro (Version 2 or Version 3) and then open and run <u>Moderate Process-2.sas</u>. (code for Version 2 of Process) or <u>Moderate Process-3.sas</u> (code for Version 3 of Process).

Version 2: %process (data=Zs,vars=AR Misanth Ideal,y=ar,x=Misanth,m=Ideal,model=1,jn=1,plot=1);

Version 3: %process (data=Zs,y=ar,x=Misanth,w=Ideal,model=1,jn=1,plot=1);

Here are parts of the output not redundant with what has already been shown above.

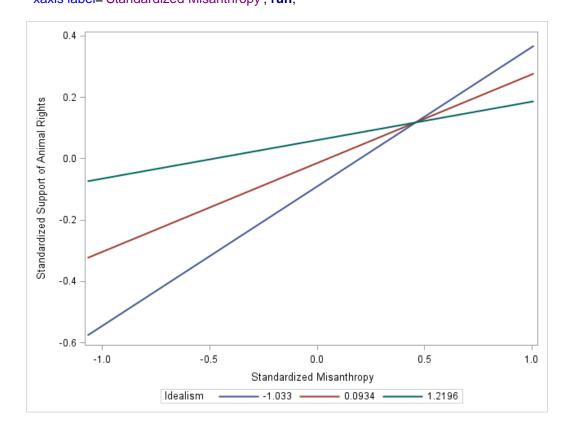
## Plotting the Simple Slopes.

Data for visualizing the conditional effect of the focal predictor:

MISANTH	IDEAL	AR
-1.0703	-1.0328	-0.5746
-0.1793	-1.0328	-0.1709
1.0085	-1.0328	0.3675
-1.0703	0.0934	-0.3235
-0.1793	0.0934	-0.0658
1.0085	0.0934	0.2777
-1.0703	1.2196	-0.0723
-0.1793	1.2196	0.0392
1.0085	1.2196	0.1879

#### These data can then be used with SGPLOT like this:

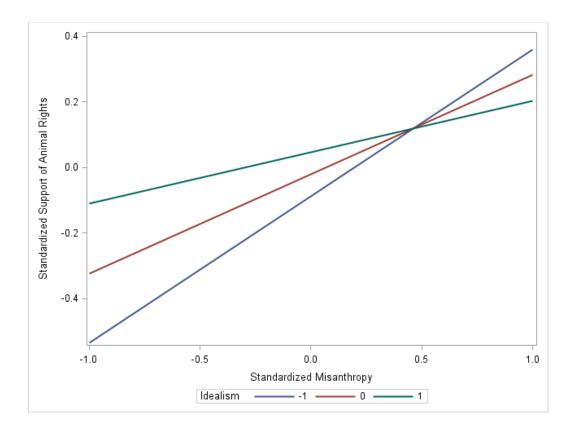
```
Data plot1; input Misanthropy Idealism Animal_Rights; cards;
-1.0703 -1.0328 -0.5746
-0.1793 -1.0328 -0.1709
1.0085 -1.0328 0.3675
-1.0703 0.0934 -0.3235
-0.1793 0.0934 -0.0658
1.0085 0.0934 0.2777
-1.0703 1.2196 -0.0723
-0.1793 1.2196 0.0392
1.0085 1.2196 0.1879
;
proc sgplot; reg x = misanthropy y = Animal_Rights / group = Idealism nomarkers; yaxis label='Standardized Support of Animal Rights'; xaxis label='Standardized Misanthropy'; run;
```



With Version 3 of Process the simple slopes are obtained at the 16<sup>th</sup>, 50<sup>th</sup>, and 84<sup>th</sup> percentiles of the moderator. If the moderator were perfectly normally distributed in the sample, one standard deviation below the mean would be the 15.87<sup>th</sup> percentile, the 50<sup>th</sup> percentile would be the mean, and one standard deviation above the mean would be the 84.13<sup>th</sup> percentile.

In Hayes' own words, "the 16<sup>th</sup> and 84<sup>th</sup> percentiles of the distribution of *W* will always be within the range of the observed data, regardless of the shape of the distribution. And the median (the 50<sup>th</sup> percentile) of the distribution of *W* is always a sensible description of the center. The mean may not be if *W* is skewed. Although these might seem like strange percentiles to use (why not nice round numbers like the 25<sup>th</sup> 50<sup>th</sup> and 75<sup>th</sup> percentiles, for instance?), the 16<sup>th</sup>, 50<sup>th</sup>, and 84<sup>th</sup> percentiles correspond to a standard deviation below the mean, the mean, and a standard deviation above the mean if *W* is exactly normally distributed."

This default can be changed in Process 3 to be the more commonly seen one standard deviation below the mean, the mean, and one standard deviation above the mean, which was the default in Process 2. The resulting plot would look like this:



### Making a Jonhson-Neyman Plot

Moderator values(s) defining Johnson-Neyman significance region(s)				
Value	% below	% above		
0.7788	77.2727	22.7273		

Misanthropy is significantly correlated with support for animal rights when the standardized value of idealism is .77 or lower.

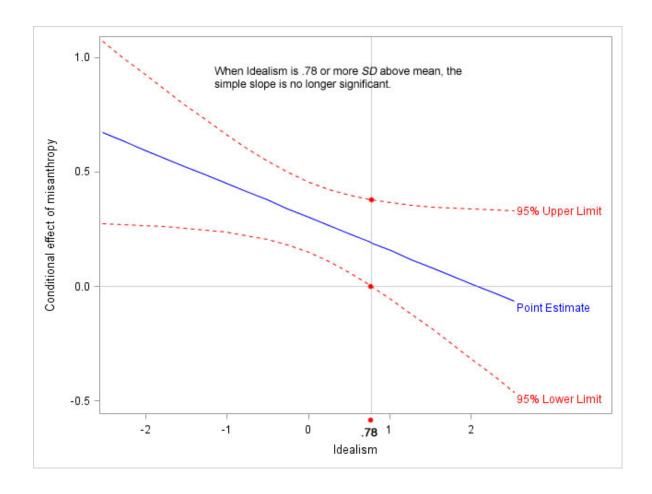
## Conditional effect of focal predictor at values of the moderator:

	IDEAL	Effect	se	t	р	LLCI	ULCI
	-2.5343	0.6719	0.2023	3.3219	0.0011	0.2722	1.0716
	-2.2809	0.6350	0.1852	3.4279	0.0008	0.2690	1.0010
	-2.0275	0.5981	0.1686	3.5483	0.0005	0.2650	0.9312
	-1.7742	0.5612	0.1523	3.6846	0.0003	0.2603	0.8622
	-1.5208	0.5243	0.1367	3.8367	0.0002	0.2543	0.7943

Conditional effect of focal predictor at values of the moderator:

IDEAL	Effect	se	t	р	LLCI	ULCI
-1.2674	0.4874	0.1218	4.0008	0.0001	0.2467	0.7281
-1.0140	0.4505	0.1082	4.1651	0.0001	0.2368	0.6642
-0.7606	0.4136	0.0962	4.3013	0.0000	0.2236	0.6036
-0.5072	0.3767	0.0865	4.3542	0.0000	0.2057	0.5476
-0.2538	0.3398	0.0801	4.2431	0.0000	0.1816	0.4980
-0.0004	0.3029	0.0777	3.8998	0.0001	0.1494	0.4563
0.2529	0.2660	0.0796	3.3400	0.0011	0.1086	0.4233
0.5063	0.2291	0.0857	2.6734	0.0083	0.0598	0.3984
0.7597	0.1922	0.0950	2.0220	0.0450	0.0044	0.3800
0.7788	0.1894	0.0959	1.9759	0.0500	0.0000	0.3788
1.0131	0.1553	0.1068	1.4533	0.1482	-0.0558	0.3664
1.2665	0.1184	0.1204	0.9834	0.3270	-0.1195	0.3562
1.5199	0.0815	0.1351	0.6030	0.5474	-0.1855	0.3484
1.7733	0.0446	0.1507	0.2957	0.7679	-0.2532	0.3423
2.0267	0.0076	0.1669	0.0458	0.9635	-0.3221	0.3374
2.2800	-0.0293	0.1835	-0.1594	0.8735	-0.3919	0.3333
2.5334	-0.0662	0.2005	-0.3300	0.7419	-0.4623	0.3300

The plot below was made using the statistics in the two leftmost and two rightmost columns of the table "Conditional effect of X on Y at values of the moderator (M)."



### Black and White Graphs

SAS makes good use of color in its graphics, but some publishers insist on having black and white graphs (those troglodytes). Maybe they are just cheap – it probably costs more to publish color images than black and white ones. In any case, here is the code you will need to make your interaction plot colorless:

# ods graphics / AttrPriority=NONE;

**data** plot; input Misanthropy Idealism Animal\_Rights; cards;

-1.0703 -1.0328 -0.5746

-0.1793 -1.0328 -0.1709

1.0085 -1.0328 0.3675

-1.0703 0.0934 -0.3235

-0.1793 0.0934 -0.0658

1.0085 0.0934 0.2777

-1.0703 1.2196 -0.0723

-0.1793 1.2196 0.0392

1.0085 1.2196 0.1879

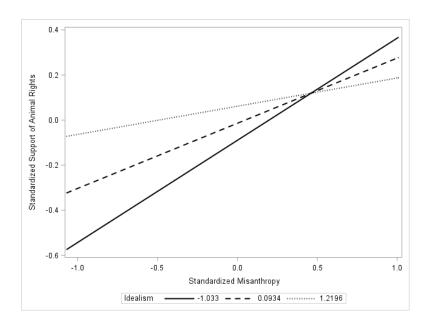
proc sgplot; reg  $x = misanthropy y = Animal\_Rights / group = Idealism nomarkers;$ 

styleattrs datacontrastcolors=(black black black)

datalinepatterns=(solid dash dot);

yaxis label='Standardized Support of Animal Rights';

# xaxis label='Standardized Misanthropy'; run;



- Fair Use of This Document
- Does it Matter Which of A and B is moderator and Which is Focal Predictor?
- Reporting the Results from a Simple Moderation Analysis
- Nearly Complete Statistical Output Version 3
- RMNET post on this topic
- Example of Sequential Analysis With Moderation, SPSS
- SAS Line Patterns
- SAS attrpriority
- Return to Wuensch's Statistics Lessons Page

The url for this document is http://core.ecu.edu/psyc/wuenschk/MV/MultReg/Moderator.pdf

Copyright 2021, Karl L. Wuensch - All rights reserved.