Moderation analysis

The moderation model considered OHS (DMT) as *independent variable* and OHIP as *dependent variable*, with HUDBI as the *moderator*.

<pre>Model = 1 Y = ZOHIP X = ZDMT M = zD_X_ZHU Sample size 191 *********************************</pre>							
<pre>Model = 1 Y = ZOHTP X = ZDMT M = zD_x_zHU Sample size 191 *******************************</pre>							
<pre>Y = ZOHTP X = ZDMT M = zD_X_ZHU Sample size 191 **********************************</pre>							
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$							
$M = zD_x_zHU$ Sample size 191 Model Summary R R-sq MSE F dfl df2 p .3330 .1109 .9034 5.9307 3.0000 187.0000 .0007 Model coeff se t p LLCI ULCI constant0130 .07141816 .85611539 .1279 zD_x_zHU .2758 .1072 2.5714 .0109 .0642 .4873 ZDMT .2905 .0986 2.9459 .0036 .0959 .4850 int_10989 .0816 -1.2123 .22692598 .0620 Product terms key: int_1 ZDMT X zD_x_zHU R-square increase due to interaction(s): R2-chng F dfl df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 							
M - 2D_x_2NO Sample size 191 ************************************							
Sample size 191 *********************************							
Sample size 191 *********************************							
191 							
Amountain of the moderator(s): 2011							
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Outcome: ZOHIP Model Summary R R-sq MSE F df1 df2 p .3330 .1109 .9034 5.9307 3.0000 187.0000 .0007 Model Coeff se t p LLCI ULCI constant0130 .07141816 .85611539 .1279 zD_x_zHU .2758 .1072 2.5714 .0109 .0642 .4873 ZDMT .2905 .0986 2.9459 .0036 .0959 .4850 int_10989 .0816 -1.2123 .22692598 .0620 Product terms key: int_1 ZDMT X zD_x_zHU R-square increase due to interaction(s): R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 							
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$\begin{array}{c cccccc} \mbox{Model Summary} & R & R-sq & MSE & F & dfl & df2 & p \\ \hline .3330 & .1109 & .9034 & 5.9307 & 3.0000 & 187.0000 & .0007 \\ \label{eq:model} \mbox{Model} & & & & & & & & \\ \mbox{constant} &0130 & .0714 &1816 & .8561 &1539 & .1279 \\ \mbox{zD_x_zHU} & .2758 & .1072 & 2.5714 & .0109 & .0642 & .4873 \\ \mbox{zDMT} & .2905 & .0986 & 2.9459 & .0036 & .0959 & .4850 \\ \mbox{int_1} &0989 & .0816 & -1.2123 & .2269 &2598 & .0620 \\ \mbox{Product terms key:} & & & & \\ \mbox{int_1} & ZDMT & X & zD_x_zHU \\ \mbox{R-square increase due to interaction(s):} & & & \\ \mbox{R2-chng} & F & dfl & df2 & p \\ \mbox{int_1} & .0212 & 1.4698 & 1.0000 & 187.0000 & .2269 \\ \end{tabular} \end{array}$							
Model Summary R R-sq MSE F dfl df2 p .3330 .1109 .9034 5.9307 3.0000 187.0000 .0007 Model coeff se t p LLCI ULCI 0.0007 Model coeff se t p LLCI ULCI 0.0007 Zonstant 0130 .0714 1816 .8561 1539 .1279 zD_x_zHU .2758 .1072 2.5714 .0109 .0642 .4873 ZDMT .2905 .0986 2.9459 .0036 .0959 .4850 int_1 0989 .0816 -1.2123 .2269 2598 .0620 Product terms key: int_1 ZDMT X zD_x_zHU R-square increase due to interaction(s): R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 .2269							
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Model coeff se t p LLCI ULCI constant 0130 .0714 1816 .8561 1539 .1279 zD_x_zHU .2758 .1072 2.5714 .0109 .0642 .4873 ZDMT .2905 .0986 2.9459 .0036 .0959 .4850 int_1 0989 .0816 -1.2123 .2269 2598 .0620 Product terms key: int_1 ZDMT X zD_x_zHU ZD_x_zHU R-square increase due to interaction(s): R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 ************************************							
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ZDMT .2905 .0986 2.9459 .0036 .0959 .4850 int_1 0989 .0816 -1.2123 .2269 2598 .0620 Product terms key: int_1 ZDMT X zD_x_zHU ZD_x_zHU R-square increase due to interaction(s): R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 ************************************							
<pre>int_10989 .0816 -1.2123 .22692598 .0620 Product terms key: int_1 ZDMT X zD_x_ZHU R-square increase due to interaction(s):</pre>							
Product terms key: int_1 ZDMT X zD_x_zHU R-square increase due to interaction(s): R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 ************************************							
Product terms key: int_1 ZDMT X zD_x_ZHU R-square increase due to interaction(s): R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 ************************************							
<pre>int_1 ZDMT X zD_x_zHU R-square increase due to interaction(s):</pre>							
<pre>int_1 ZDMT X zD_x_zHU R-square increase due to interaction(s): R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 ***********************************</pre>							
<pre>int_1 ZDMT X zD_x_zHU R-square increase due to interaction(s):</pre>							
R-square increase due to interaction(s): R2-chng F dfl df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 ************************************							
<pre>R-square increase due to interaction(s):</pre>							
R2-chng F df1 df2 p int_1 .0212 1.4698 1.0000 187.0000 .2269 ************************************							
<pre>int_1 .0212 1.4698 1.0000 187.0000 .2269 ***********************************</pre>							
Inte_1 .0212 Integer Integer .2203 ************************************							
<pre>************************************</pre>							
Conditional effect of X on Y at values of the moderator(s): zD_x_zHU Effect se t p LLCI ULCI							
Conditional effect of X on Y at values of the moderator(s): zD_x_zHU Effect se t p LLCI ULCI							
Conditional effect of X on Y at values of the moderator(s): zD_x_zHU Effect se t p LLCI ULCI							
zD_x_zHU Effect se t p LLCI ULCI							
-1 0889 3981 1038 3.8371 0002 1934 6028							
.0000 .2905 .0900 2.9459 .0050 .0959 .4650							
1.0889 .1828 .1564 1.1689 .24391257 .4913							
Values for quantitative moderators are the mean and plus/minus one SD from							
variable for quantificative moderators are the mean and prab, minds one ob from							
mean.							

According to the results from PROCESS, the overall moderation model had significant effects F(3, 187) = 5.93, p < .001. However, the overall interaction effect was not significant: *C.I.* = $-.26 \div .06$. We further used the Johnson-Neyman ¹ technique to probe for interaction and to identify ranges of values of the moderator for which the interaction effect is significant. Hayes' PROCESS macro for SPSS ² incorporates the Johnson-Neyman technique ³.

Moderator value(s) defining Johnson-Neyman significance region(s)							
Value	% below	% above					
.5158	75.3927	24.6073					
Conditional e	ffect of X	on Y at valu	les of the r	noderator	(M)		
zD_x_zHU	Effect	se	t	р	LLCI	ULCI	
-4.4882	.7342	.3399	2.1604	.0320	.0638	1.4047	
-3.9709	.6831	.2994	2.2814	.0237	.0924	1.2737	
-3.4536	.6319	.2595	2.4349	.0158	.1199	1.1439	
-2.9364	.5808	.2205	2.6339	.0091	.1458	1.0158	
-2.4191	.5296	.1829	2.8961	.0042	.1689	.8904	
-1.9019	.4785	.1477	3.2387	.0014	.1870	.7700	
-1.3846	.4274	.1174	3.6416	.0004	.1958	.6589	
8674	.3762	.0963	3.9065	.0001	.1862	.5662	
3501	.3251	.0913	3.5603	.0005	.1450	.5052	
.1672	.2739	.1047	2.6168	.0096	.0674	.4804	
.5158	.2394	.1214	1.9727	.0500	.0000	.4789	
.6844	.2228	.1309	1.7017	.0905	0355	.4810	
1.2017	.1716	.1639	1.0469	.2965	1518	.4951	
1.7189	.1205	.2004	.6011	.5485	2749	.5159	
2.2362	.0693	.2388	.2904	.7719	4018	.5405	
2.7535	.0182	.2783	.0654	.9479	5308	.5672	
3.2707	0329	.3185	1034	.9177	6612	.5953	
3.7880	0841	.3591	2341	.8151	7925	.6244	
4.3052	1352	.4001	3380	.7357	9245	.6540	
4.8225	1864	.4413	4224	.6732	-1.0569	.6841	
5.3398	2375	.4826	4921	. 6232	-1.1896	.7146	
5.8570	2887	.5241	5508	.5825	-1.3226	.7453	
	. 2007				1.0220	• • • • • • •	

* the values corresponding to the identified regions of significance are show above, in blue.

One such region of significant moderation, from -4.49 to .52 values of HUDI (in z-scores) was identified. The moderation effect was graphically depicted in Figures 4 and 5, in the main text, and showed that the moderator HUDBI strengthened the positive relation between DMT and OHIP, for the regions of significance indicated by the Johnson-Neyman technique.

^{1.} Johnson PO, Neyman J. Tests of certain linear hypotheses and their application to some educational problems. *Statistical research memoirs*. 1936;1:57-93.

^{2.} Hayes AF. PROCESS: A versatile computational tool for observed variable mediation, moderation, and conditional process modeling. 2012. http://www.afhayes.com/public/process2012.pdf.

^{3.} Hayes AF, Matthes J. Computational procedures for probing interactions in OLS and logistic regression: SPSS and SAS implementations. *Behavior research methods*. 2009;41(3):924-936.