JD-R Theory: Using the Content of the O*Net

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- 7 must be indented, like this line.

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The Job Demands-Resources Theory [JD-R; Demerouti et al. (2001)] has received wide support across contexts and varied research questions. We contribute to this literature by exploring the interaction between *perceptions* of job demands and resources on stress using the broad set of job characteristics provided on O*Net.

A series of evaluations were made that used: direct O*Net terminology (both descriptor and response option), and JD-R influenced ratings of demand and hindrance stressors. Prior to a description of results, a brief overview of both the JD-R theory, the stress appraisal process, and O*Net is provided.

23 The Job demands-Resources Theory

The job demands-resources theory is an expansion of the well-studied job 24 demands-resources model (Demerouti et al., 2001). One of the major advantages of the job 25 demands-resources theory is that it allows us to model both work environment and job 26 characteristics via job resources and demands. Resources include physical, psychological, social, or organizational aspects of the job that may help an employee achieve work goals, reduce job demands, or promote personal growth and development (Demerouti et al., 2001). In contrast, demands include components of a job that require sustained effort, and as such, produce psychological or physiological strain (e.g., high work pressure is frequently 31 cited as a common demand; Demerouti et al. (2001)). Cognitively, the perception of an element of ones job as a resource or demand activates one of two distinct processes: either health impairment (resulting from demands) or motivation [resulting from resources; A. B. Bakker and Demerouti (2014). Pertinent to the current study, demanding job characteristics are frequently associated with negative outcomes (e.g., A. Bakker et al., 2003), whereas job characteristics deemed resources have been associated with positive 37 organizational outcomes like engagement and motivation (A. B. Bakker et al., 2007).

52

Objective vs. Subjective Nature of Demands and Resources: The Role of Appraisal

Searle and Auton (2015) note that the majority of the research on workplace 41 demands is based on apriori classifications of demands. However, the stress experience, or 42 process, described early on by Lazarus and Folkman (1984) is grounded in the assumption that individual appraisals of stressors/demands vary. Their transactional theory of stress and coping states that people continuously appraise stimuli in their environments. An appraisal is the cognitive process whereby meaning is assigned to a stimulus. If a stimulus is appraised as a stressor (threat, challenge, potentially harmful), emotional distress leads to coping of some kind. This action to cope is also associated with another appraisal about the outcome itself and the process continues if the outcomes is not appraised as favorable (Lazarus & Folkman, 1984). The stress appraisal process suggests that classifying a job 50 characteristic or environmental condition as an objective demand or resource might be in 51 error.

We next consider the (limited) empirical evidence on this topic. First, some relatively 53 recent research suggests that job demands and resources may not be universally appraised or assigned as such. Starting with job demands, Webster et al. (2011), for example, studied workload, role ambiguity, and role conflict demands, and found that while each could be appraised primarily as a challenge or hindrance demand, they could also simultaneously be perceived as being both a challenge and hindrance to different degrees. While their study did include resources, it nonetheless points to individual differences on how people perceive stressors at work. Although part of a much larger study on retirement, Sonnega et al. (2018) compared self-reported (subjective) ratings of degree of physical demand, stress, and need for intense concentration from the Health and Retirement Study with objective ratings from O*Net. Correlations physical demand (r = .52), stress (r = .10), and need for intense concentration (r = .14), again suggesting perhaps that our objective ratings of job demands (and resources) may be subject to a greater level of individual difference than assumed.

Next considering resources, Schmitz et al. (2019) also captured subjective and objective resources in their study of retirement. Correlations of composite variables for the resources of autonomy (r = .12. p > .01), recognition of work (r = .07, p > .01), and decision freedom (r = .08, p > .01), while significant, certainly do not reflect high levels of overlap.

We do acknowledge as well, that demands and resources are not necessarily consistent across days, or seasons, for many employees. Downes et al. (2021) meta-analysis addresses this reality in depth, although it is beyond the scope of this project.

O*Net Resource

Originally, the Advisory Panel for the Dictionary of Occupational Titles 74 recommended a system that would "... promote the effective education, training, 75 counseling, and employment of the American workforce. It should accomplish its purpose 76 by providing a database system that identifies, defines, classifies, and describes occupations 77 in the economy in an accessible and flexible manner" (Dictionary of Occupational Titles 78 (US) and Service (1993), p. 6). The result was the now commonly used O*NET. The Occupational Information Network (O*NET; onetonline.org) contains a comprehensive description of occupations (Peterson et al., 2001). This widely accessed database houses hundreds of standardized and occupation-specific descriptors most occupations in the US and these descriptions are continually updated. In fact, there was a call to work with experienced I/O psychologists over the summer to update the content for the Industrial and Organizational Psychologist listing on O*Net. These data, and the tools provided for free on the website (e.g., Career Exploration Tools, "My Next Move for Veterans", "My Next Move", Toolkit for Business) are frequently used by counselors, students, human resources departments, and researchers to assist potential applicants discover the skills and training they need for the job of their choice. It is also useful to employers by providing 89 them with information with which to craft job descriptions and help employees determine

91 what skills are needed for promotion.

Of greatest interest here are statements taken from O*NET "activity" and "context" 92 classifications (e.g., items related to information input, interacting with others, physical work conditions, structural job characteristics). One of the first and basic questions is whether or not the categorical examples of "resources" and "demands" described in the Job Demands-Resources Theory (Demerouti et al., 2001), for example, are generally deemed resources or demands as we objectively define them. The next logical question surrounds how "universal" such ratings are. For instance, it is quite possible, given the theoretical 98 and empirical evidence presented above, that there is wide variability in individual 99 appraisal of work activities and context such that some people may rate a given activity as 100 a resource and others a hindrance. A second study extends the findings from Study 1 to a 101 potentially key moderator - job categories/classifications, examining whether ratings of 102 resources, challenge- and hindrance demands differ by job classification. 103

104 Methods

05 Participants

107

111

There were 568 respondents.

Participants.

- 568 respondents, 13.57% had been in their referent job less than 6 months, 19.20% between 6 months and a year, 49.12% between one and five years, 13.27% between 5 and 10 years, and 4.87% more than 10 years.
 - Ages ranged from 18 to 65 with an average of 28.18 years old (SD = 7.53).
- Gender: female (52.58%) or male (46.83%).
- Job classifications: International Standard Classification of Occupations (ISCO) via
 the package labour (R-labour R?), and further categorized into "knowledge" (n =

320) versus "skilled" (n = 214) occupations with knowledge workers being identified via ISCO classifications of: 1) professionals, and 2) managers.

The data for this study were collected through Prolific sample, 18 or older and holding
a full-time or part-time job. Participants were asked to think about their primary job while
answering the survey, and upon completion each participant was compensated in the
amount of six US dollars.

121 Materials

We used 98 statements taken directly from O*Net's "activity" and "context" 122 classifications. Each of the 98 descriptors has potentially unique response categories, but 123 scaling was consistently 1 (low) to 5 (high). Subsequent to these self-evaluations, 124 respondents were asked to rate elements in terms of 1) ... this aspect of your job is a 125 resource that can be functional in achieving work goals, reduce job demands, or stimulate 126 personal growth/development, 2) ... this aspect of your job is a challenge that can promote 127 mastery, personal growth, or future gains, and 3) ... this aspect of your job is a hindrance 128 that can inhibit personal growth, learning, and work goal attainment. 129

130 Procedure

We used PROCESS for R Version 4.1.1 (Hayes, 2022) to assess the extent to which the relationship between demands and stress are moderated by resources.

133 Results

```
##
       Documentation available in Hayes (2022). www.guilford.com/p/hayes3
  ##
139
  ##
141
  ## PROCESS is now ready for use.
142
  ## Copyright 2022 by Andrew F. Hayes ALL RIGHTS RESERVED
143
  ## Workshop schedule at http://haskayne.ucalgary.ca/CCRAM
144
  ##
145
  ##
146
     ************* PROCESS for R Version 4.1.1 ****************
  ##
147
  ##
              Written by Andrew F. Hayes, Ph.D. www.afhayes.com
  ##
149
       Documentation available in Hayes (2022). www.guilford.com/p/hayes3
  ##
150
  ##
151
    ******************************
152
  ##
153
  ## Model : 1
154
  ##
        Y: stress
155
        X : overall.hindrance
  ##
156
  ##
        W : overall.resource
157
  ##
158
  ## Sample size: 568
  ##
160
  ##
  ## Outcome Variable: stress
  ##
164
```

```
## Model Summary:
   ##
                 R
                                                  F
                                                           df1
                                                                      df2
                         R-sq
                                     MSE
                                                                                   p
166
            0.1311
                       0.0172
                                  0.7790
                                            3.2876
                                                       3.0000 564.0000
                                                                             0.0205
167
   ##
168
   ## Model:
169
   ##
                              coeff
                                                                           LLCI
                                                                                      ULCI
                                             se
                                                        t
                                                                   р
170
                                                                        -0.7063
   ## constant
                             1.2688
                                        1.0055
                                                   1.2618
                                                              0.2075
                                                                                    3.2439
171
   ## overall.hindrance
                             0.8336
                                        0.4031
                                                   2.0677
                                                              0.0391
                                                                        0.0417
                                                                                    1.6254
172
   ## overall.resource
                             0.3319
                                        0.2518
                                                   1.3181
                                                              0.1880
                                                                        -0.1627
                                                                                    0.8264
173
   ## Int 1
                            -0.1918
                                        0.1024
                                                  -1.8725
                                                              0.0616
                                                                        -0.3929
                                                                                    0.0094
174
   ##
175
   ## Product terms key:
176
   ## Int 1 : overall.hindrance x overall.resource
177
178
   ## Test(s) of highest order unconditional interaction(s):
179
   ##
             R2-chng
                              F
                                       df1
                                                  df2
180
                                                               р
              0.0061
                         3.5064
                                    1.0000 564.0000
   ## X*W
                                                          0.0616
   ## -----
182
   ## Focal predictor: overall.hindrance (X)
183
   ##
             Moderator: overall.resource (W)
184
   ##
185
   ## Conditional effects of the focal predictor at values of the moderator(s):
186
         overall.resource
                              effect
                                                                                       ULCI
   ##
                                              se
                                                          t
                                                                            LLCI
                                                                     р
187
   ##
                   3.2983
                              0.2010
                                         0.0802
                                                    2.5065
                                                               0.0125
                                                                          0.0435
                                                                                     0.3586
188
                                         0.0534
                                                    2.1759
                                                               0.0300
                                                                          0.0113
                                                                                     0.2213
   ##
                   3.7402
                              0.1163
189
   ##
                   4.2063
                              0.0269
                                         0.0594
                                                    0.4535
                                                               0.6503
                                                                         -0.0897
                                                                                     0.1435
190
   ##
191
```

Moderator value(s) defining Johnson-Neyman significance region(s):

193 ## Value % below % above
194 ## 3.8196 55.6338 44.3662

195 ##

Conditional effect of focal predictor at values of the moderator:

130		001101101101101	p.					
197	##	overall.resource	effect	se	t	р	LLCI	ULCI
198	##	1.0149	0.6389	0.3003	2.1276	0.0338	0.0491	1.2288
199	##	1.2078	0.6020	0.2809	2.1433	0.0325	0.0503	1.1536
200	##	1.4006	0.5650	0.2615	2.1608	0.0311	0.0514	1.0785
201	##	1.5935	0.5280	0.2421	2.1807	0.0296	0.0524	1.0035
202	##	1.7863	0.4910	0.2228	2.2034	0.0280	0.0533	0.9287
203	##	1.9791	0.4540	0.2037	2.2293	0.0262	0.0540	0.8540
204	##	2.1720	0.4170	0.1846	2.2592	0.0243	0.0545	0.7796
205	##	2.3648	0.3801	0.1657	2.2937	0.0222	0.0546	0.7055
206	##	2.5577	0.3431	0.1470	2.3336	0.0200	0.0543	0.6318
207	##	2.7505	0.3061	0.1287	2.3791	0.0177	0.0534	0.5588
208	##	2.9434	0.2691	0.1108	2.4292	0.0154	0.0515	0.4867
209	##	3.1362	0.2321	0.0937	2.4784	0.0135	0.0482	0.4161
210	##	3.3290	0.1951	0.0778	2.5085	0.0124	0.0423	0.3479
211	##	3.5219	0.1582	0.0641	2.4667	0.0139	0.0322	0.2841
212	##	3.7147	0.1212	0.0543	2.2306	0.0261	0.0145	0.2279
213	##	3.8196	0.1011	0.0515	1.9642	0.0500	0.0000	0.2021
214	##	3.9076	0.0842	0.0507	1.6605	0.0974	-0.0154	0.1838
215	##	4.1004	0.0472	0.0545	0.8662	0.3867	-0.0599	0.1543
216	##	4.2933	0.0102	0.0644	0.1589	0.8738	-0.1163	0.1368
217	##	4.4861	-0.0267	0.0782	-0.3421	0.7324	-0.1803	0.1268
218	##	4.6790	-0.0637	0.0941	-0.6773	0.4985	-0.2485	0.1211

```
##
                    4.8718
                              -0.1007
                                          0.1112
                                                    -0.9054
                                                                0.3656
                                                                          -0.3192
                                                                                      0.1178
219
   ##
220
   ## Data for visualizing the conditional effect of the focal predictor:
221
         overall.hindrance overall.resource
   ##
                                                   stress
222
   ##
                     1.6667
                                       3.2983
                                                   2.6985
223
   ##
                     2.2894
                                       3.2983
                                                   2.8237
224
   ##
                     3.2416
                                       3.2983
                                                   3.0151
225
                     1.6667
                                       3.7402
                                                  2.7039
   ##
226
                     2.2894
                                       3.7402
                                                  2.7763
   ##
227
                     3.2416
                                       3.7402
                                                  2.8871
   ##
228
                     1.6667
                                       4.2063
                                                  2.7096
   ##
229
                     2.2894
   ##
                                       4.2063
                                                  2.7264
230
                     3.2416
                                       4.2063
   ##
                                                   2.7520
231
   ##
      ************ ANALYSIS NOTES AND ERRORS ****************
233
   ##
234
   ## Level of confidence for all confidence intervals in output: 95
   ##
236
   ## W values in conditional tables are the 16th, 50th, and 84th percentiles.
237
```

#Results

238

A moderated regression including hindrances, resources, and the interaction between them was done using PROCESS, version 4.1.1. First, the overall regression model including mean hindrances, mean resources, and the interaction between the two variables was significant, F(3, 564) = 3.29, p = .020. The interaction between hindrance and resources (uncentered) revealed that the relationship between hindrances and stress was conditional on resources, F(3, 564) = 3.51, p = .061. As can be seen in Figure 1, those with fewer

resources show a much stronger positive relationship between hindrances and stress than
those with more resources. Upon exploring the interaction further, it was evident that this
moderated effect happened at lower, but not higher levels of resources.

Next steps: 1) Make a prettier graph using ggplot. Note the percentiles that we graphed. 2) Or, could generate Johnson-Neyman Technique – shows specifically the range of W values that are significant. See p. 272 for visual and r-script. Table example on p. 286: includes variables, symbol, coeff, SE, t, p; below that, rsquared, MSE and f-string.

252 Discussion

253 References

262

263

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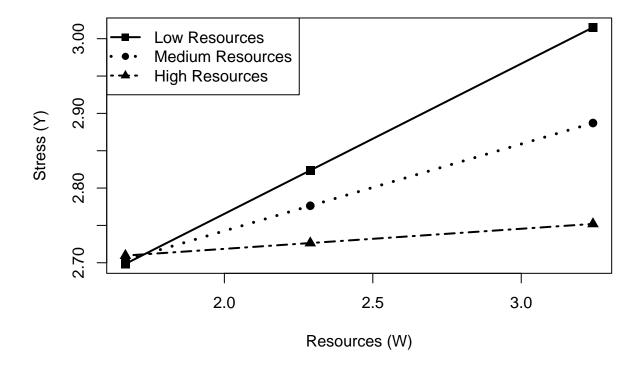
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Table 1
Results from a regression analysis examining the moderation of resources on the relationship between hindrance demands and stress

Component	coeff	SE	t	р
Constant	1.27	1.01	1.26	0.21
Hindrance (X)	0.83	0.40	2.07	0.04
Resource (W)	0.33	0.25	1.32	0.19
Hindrance x Resource	-0.19	0.10	-1.87	0.06

Note. R^2 etc here



Figure~1. Interaction between hindrances and resources as predictors of stress