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# A Qualitative Analysis of Construct Measurement Techniques Used in Industrial/Organizational Research

Benjamin Michael btm67@uakron.edu

Andrea F. Snell asnell@uakron.edu

Katie Rosneck kr157@uakron.edu

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# Honors Research Project

## A Qualitative Analysis of Construct Measurement Techniques Used in I/O Research

## Benjamin T. Michael

Student

College of Arts and Sciences

The University of Akron

#### Dr. Andrea Snell

Professor

Department of Psychology

College of Arts and Sciences

The University of Akron

#### **Abstract**

This project aims to challenge the appropriateness of the methodological strategies and tools utilized within psychological research. We will look at the types of statistical modeling used and the context in which they are used, such as measurement modeling, confirmatory factor analysis, hierarchical confirmatory factor analysis, and bifactor analysis within survey development. The objective of this research is to search for and recognize patterns from the content of some of the top journal articles in the field of industrial and organizational psychology. The information gained from analyzing the content of the journal articles will be used to create a qualitative assessment regarding when and how tools and strategies are being used in the IO field at the moment, and to provide considerations for future research methodologies.

Keywords: Confirmatory Factor Analysis, Bifactor Measurement Model, Hierarchical Construct Factor Analysis, Unit Weighted Composite Score, Industrial / Organizational Psychology, Measurement Model

#### Introduction

There are an abundant number of statistical tools utilized within the field of Industrial / Organizational (I/O) Psychology. The proper usage of these tools is vital in understanding the research being conducted and understanding the strengths of the relationships between constructs. This paper aims to showcase the appropriateness of the statistical tools being utilized within current I/O research, when they are being used, and if they are being used in proper context. The motivation that prompted this research was the desire to better examine the measurement tools that are currently being used in psychological research that may not be matching the correct research questions with their corresponding research designs. The objective for this research is to complete a targeted analysis of some of the top IO journals to qualitatively assess the different designs being used to determine when and how the tools are currently being used and provide recommendations for future psychological researchers. At the end of this literary review, we believe there will be a significant trend in the use of statistical modeling within the current I/O research.

### Methodology

To analyze the type of statistical modeling being used within I/O research, a sample size of 42 journal articles submitted to the Journal of Applied Psychology (JAP) were reviewed. The methodology sections within each journal entry were analyzed by looking for the type of statistical modeling being utilized and the context in which the modeling was being applied. The types of statistical modeling that were of interest when analyzing the JAP articles were Unit

Weighted Composite Models, Exploratory Factor Analysis (EFA), First Order Confirmatory Factor Analysis (CFA), Hierarchical Confirmatory Factor Analysis (HCFA), and Bifactor Analysis models (**Figure 1**).

Prior to analyzing the methodology being used in the JAP articles, the current reasoning for using each method was broken down into a pros and cons list to determine potential reasoning behind utilizing specific types of methodology (**Figure 1**). Each journal article was searched for the use of these statistical models and in the specific case of CFA, whether a path analysis was conducted afterwards to determine if the CFA was utilized properly. This type of qualitative analysis allowed us to draw general conclusions from our observations with our analysis of the journal articles. Not relying on a specific theory to formulate a hypothesis gave us flexibility with our research compared to a deductive approach, which uses existing theory to formulate a hypothesis. This methodology allowed us to develop theory and recognize patterns when studying the content of the JAP articles.

Figure 1

Pros and Cons of Statistical Models of Interest

Statistical Model	Pros: Cons:	
Unit Weighted Composite	Easy, simple to utilize	Assumes equal relevancies
		for all items
Exploratory Factor Analysis	Exploratory and empirical	Weights are not Stable
Confirmatory factor Analysis	Well known, allows for	Ignores general factor
(First Order)	differential weights, and is a	variance.
	priori	
Hierarchical Confirmatory	Theoretically supported for	Requires additional model
Factor Analysis	most theoretical constructs.	parameters.
	Subfactor structure	Takes more effort
Bifactor Analysis	Models both a general factor	Can alter a priori factor
	and more specific factors.	structure.

Note. Pros and Cons of Unit Weighted Composite, Exploratory Factor Analysis, Confirmatory factor Analysis (First Order), Hierarchical Confirmatory Factor Analysis, and Bifactor Analysis.

The type of methodology used in each article was highlighted and reported into two sets of data tables for review once all the articles had been analyzed (**Figure 2 & 3**). Once all the data regarding the methodology was reported into the data table, the data was analyzed further. The types of models used, frequency of use, and context of use were all analyzed into the data table, providing a greater perspective on the type of modeling frequently being used in I/O.

#### **Findings**

Once all JAP articles were reviewed and analyzed for their methodology, the observational data were inputted into the data table below to organize observations for further analysis. The first data table represents the observations from at the time of this publication, the latest 2 issues from the most recent volume of JAP (**Figure 2**). The second data table represents a larger sample of observations from JAP articles published within the past year in Volume 107 (**Figure 3**).

Within the 18 JAP articles analyzed in Volume 108, 16 of the articles were quantitative studies, while the remaining 4 were qualitative / review studies (**Figure 2**). Within the 14 quantitative studies, there was a trend of the authors either only utilizing CFA for their methodology or a lack of mentioning the type of modeling being used. Of the 14 JAP articles in Volume 108 analyzed, 8 of the articles utilized CFA as their statistical model. The remaining 6 quantitative studies had no mention of any of the statistical models of interest. From this data table there was a significant trend in the use of CFA as the model of choice within the articles. If CFA was not utilized there was a trend within the methodology reports of lacking clarity within

the articulation of the methodological process behind the utilization of the specific statistical tools used in the studies. Of the 8 articles that utilized CFA in their methodology, 6 of them conducted a path analysis following their use of CFA.

To see if the high usage of CFA and lack of written clarity within the methodology sections in JAP were consistent, articles published within the past year in volume 107 were analyzed for similar trends. In total, 24 journal articles were reviewed for trends in Volume 107 of JAP spanning across Issues 4,5,8 and 9 (**Figure 3**). In a similar manner to how the articles in Volume 108 were treated, the observations pertaining to the modeling used in the articles in Volume 107 were inputted into a data table and were analyzed for trends after the data for all the articles were inputted (**Figure 3**).

Of the 24 articles reviewed across Volume 107, 22 of the articles were quantitative studies, and 2 were qualitative / review articles. Of the 22 quantitative studies, CFA was explicitly mentioned in the methodology sections of 11 articles. There was a slight alteration within the previous trend of CFA being the sole statistical model being used. In 3 of the articles reviewed within Volume 107, other statistical models of interest were explicitly stated within the methodology sections such as: Exploratory factor analysis, Bifactor analysis, and Hierarchical confirmatory factor analysis. The remaining 8 articles reviewed provided no explicit explanations pertaining to the models of interest in this review article, providing further evidence that the articulation of the models being used in studies is not being explicitly stated, causing the appropriateness of these models in these contexts to be challenged. Of the 11 articles in Volume 107 that explicitly stated their process of using CFA, 5 of the articles conducted a path analysis or multi-level path analysis along with the use of CFA to conduct a deeper understanding of the relationships between variables.

**Figure 2.**Data table for JAP Articles in Volume 108

Article Title with link	Type of study	Use of models
Volume 108 Issue 3		
Multiple, Speeded Assessments Under Scrutiny: Underlying Theory, Design Considerations, Reliability, and Validity.	Quantitative	Therefore, we averaged instructor ratings into an overall performance measure (see Viswesvaran et al., 2005). A confirmatory factor analysis via Mplus 7.4 (Muthén & Muthén, 1998–2015) using the maximum likelihood parameter (MLR)estimator provided support for a one-factor model, $\chi 2(df) = 4.45(2)$ , $p = .108$ , comparative fit index (CFI) = .984, standardized root mean square residual (SRMR) = .024, although the RMSEA was poor (.113, 90% CI [.000–.259]).
Makeup Calls in Organizations: An Application of Justice to the Study of Bad Calls	Quantitative	No mention of Path Analysis  No mention of models of interest
Feeling Possessive, Performing Well? Effects of Job-Based Psychological Ownership on Territoriality, Information Exchange, and Job Performance	Quantitative	Stage 4 uses confirmatory factor analysis (CFA) to confirm the distinctiveness of territorial expanding relative to prior measures.  We applied path analysis to test the overall model
Walking on Eggshells: A Self-Control Perspective on Workplace Political Correctness	Quantitative	We then ran a confirmatory factor analysis to confirm the distinctiveness of our study variables (other orientation, self-concern, political correctness, cognitive resource depletion, and angry and withdrawn marital behavior).  We used group-mean centering
		for the daily predictors and modeled hypothesized paths with random slopes (w hile using

		fixed slopes for controls to reduce model complexity; Wangetal.,2011). As is the standard for these analyses (e.g., Hill et al., 2021; Jennings, Lanaj, & Koopman, et al., 2022; Lennard et al., 2019), and preferred to listwise deletion (Newman, 2014), missing data—which are allowable only on either outcome variables or lagged controls in Mplus—were handled using the full information maximum—likelihood estimator (which is Mplus' default setting in these instances).
Toward a Holistic Perspective of Congruence Research With the Polynomial Regression Model	Qualitative	
Stopping Surface-Acting Spillover: A Transactional Theory of Stress Perspective	Quantitative	We also conducted a multilevel confirmatory factor analysis (CFA) to analyze measurement fit. Given the unidimensional structure of all our study constructs, we conducted a CFA with each of the items of our core variables serving as first-order factors and each construct serving as a second order factor. This model adequately fit the data, χ2 = 1660.02, df = 384, comparative fit index (CFI) = .92, Tucker–Lewis index (TLI)= .91, root-mean-square error of approximation (RMSEA) = .05, standardized root-mean-square residual [SRMR] (within) = .06.  Specifically, we conducted a multilevel path analysis with fixed effects in MPlus 8.7 (Múthen & Múthen, 2021). Given our theoretical focus on our moderators, we followed past work in specifying our full model at the within-person level, and therefore not controlling for the intervention condition at Level 2 (Schabram & Heng, 2022)

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		To test our hypothesized relationships. Specifically, we conducted parametric bootstrapping using the estimated coefficients from our analyses to test the significance of our hypothesized indirect effects. This approach allowed us to estimate the sampling distribution for the f irst-, second-, and third-stage coefficients using a Monte Carlo simulation (20,000 replications).
The Organizational Psychology of Gig Work: An Integrative Conceptual Review	Qualitative	
Toward a Better Understanding of the Causal Effects of Role Demands on Work– Family Conflict: A Genetic Modeling Approach	Quantitative	No mention of models of interest
Volume 108 Issue 2		
Understanding Racism in the Workplace	Qualitative	
When Thriving Requires Effortful Surviving: Delineating Manifestations and Resource Expenditure Outcomes of Microaggressions for Black Employees	Quantitative	To first assess the fit of our hypothesized measurement model, we conducted confirmatory factor analysis (CFA) in MPlus (Muthén& Muthén,1998–2017). To test study hypotheses, we then conducted observed variable path analysis using structural equation modeling (SEM) in MPlus with maximum likelihood estimation (Kelloway, 2014). We chose this approach as "the goal of path analysis is to test a 'structural' model, that is, a model comprising theoretically based statements of relationships among constructs"
The Cost of Managing Impressions for Black Employees: An Expectancy Violation Theory Perspective	Quantitative	Step1 involves the evaluation of the measurement model. We sought to establish convergent and discriminant validity of our measures by running a confirmatory factor analysis  To examine if the moderated mediation hypotheses were full mediation or partial mediation, we included direct paths from employee race to performance

		rating,P-Ofit, and developmental i-deals to our hypothesized
		model in Figure 1.
The Missing Middle: Asian Employees' Experience of Workplace Discrimination and Pro-Black Allyship	Quantitative	We conducted confirmatory factor analyses to confirm our measures were unique from one another. Analyses revealed that the four-factor model had significantly better fit than nested three-factor models or the one-factor model ( $\Delta \chi 2 = 226.61-1080.64$ ), supporting the distinctiveness of the four measures (see Supplemental Materials, for details).
		Finally, for completeness, we conducted a path analysis testing all of our hypotheses (Hypotheses 1–5) in one model, with workplace discrimination as the independent variable, group similarity with Black employees as the mediator, pro-Black allyship as the outcome variable, employee race as a first-stage moderator, and zero-sum beliefs as a second-stage moderator (Model 21 in process macro using SPSS; Hayes, 2017).
Challenging Racism as a Black Police Officer: An Emergent Theory of Employee Anti-Racism	Qualitative	, , , ,
Asians Don't Ask? Relational Concerns, Negotiation Propensity, and Starting Salaries	Quantitative	No mention of Models of Interest
A Matter of When, Not Whether: A Meta-Analysis of Modesty Bias in East Asian Self-Ratings of Job Performance	Quantitative	No mention of models of interest.
The Ethics of Diversity Ideology: Consequences of Leader Diversity Ideology on Ethical Leadership Perception and Organizational Citizenship Behavior	Quantitative	Confirmatory factor analysis revealed that items that were reverse-coded exhibited lower factor loadings and/or highly correlated error terms see Schmitt & Stults(1985) for a discussion of this common problem. Following previous research in which a similar issue was identified (Piccolo&Colquitt,2006) we omitted reverse-coded items from the ethical leadership(6-items) and institutional discrimination(1-item)

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		awareness scales. In addition,
		given the length of our scales
		relative to the sample size, and
		that our primary interest was in
		the interrelations of constructs
		rather than the interrelations of
		items within constructs
		(Littleetal., 2002), we created
		parcels from our remaining
		items to be used as manifest
		indicators of our constructs
		(Russelletal.,1998).
		No mention of Path Analysis
		In addition, future research could
		examine other pathways that link
		leader diversity ideologies to
		outcomes of interest, focusing,
		for example, on how followers'
		trust in the leader (Ng &
		Feldman, 2015) relates to
		followers' reactions to leader
		ideologies.
Are Leaders Still Presumed White by Default? Racial Bias in Leader	Quantitative	No mention of models of interest
<u>Categorization Revisited</u>		
Challenging Conclusions About Predictive Bias Against Hispanic Test Takers in	Quantitative	Meta-analysis, no mention of
Personnel Selection		models of interest

**Figure 3.**Data table for JAP articles in Volume 107

Article With Link	Type of Study	Use of Statistical Models	
Volume 107			
Issue 8			
<u>A</u>	Quantitative	No mention of models of interest	
Comprehensive			
Examination of			
the Cross-			
Validity of			
Pareto-Optimal			
Versus Fixed-			
Weight			
Selection			
Systems in the			
<b>Biobjective</b>			
Selection			
Context			

Multiple Qualitative	
Lobholding	
Jobholding Motivations and	
Experiences: A	
Typology and	
<u>Latent Profile</u>	
<u>Analysis</u>	
Beyond Targets   Quantitative   We also conducted confirmatory factor analysis to examine the distinctivenes	s of descriptive
and Instigators: and injunctive incivility norm perceptions in Samples 2 and 3. Results showed	d that a two-
Examining factor model fit better than a one-factor model in both samples ( $p < .01$ ). Resu	ılts are
Workplace available upon request.	
Incivility in	
Dyads and the	
Moderating Role	
of Perceived	
Incivility Norms	1
Too Much to Quantitative Table 2 displays the means, standard deviations (within- and between-person)	
<u>Know? The</u> intercorrelations of the variables. Before testing our hypotheses, we first cond	
<u>Cognitive</u> multilevel confirmatory factor analysis (CFA) to confirm the distinctiveness of	
<u>Demands of</u> variables (coworker contact quality, knowledge seeking, knowledge overload,	
Daily learning, resource depletion, and goal attainment), as well as controls (time pr	essure at work
Knowledge and negative affect). We tested the hypothesized eight-factor model by loadin	g items on their
Seeking and the respective latent factors. Results showed that the hypothesized model fit the d	
Buffering Role 426.88,df = 245,CFI = .98,TLI = .98,RMSEA = .02, SRMR within = .03 and 3	
of Coworker = .04). Before proceeding, we tested several alternative models. Table 3 display	
Contact Quality descriptive statistics of these alternative CFA models' f it indices. First, to ass	
potential existence of common variance attributable to simultaneous measurer	
three alternative CFA models that collapsed latent factors based on the timing	
· · · · · · · · · · · · · · · · · · ·	
measurement. One alternative model specified items of coworker contact qual	
pressure (measured at baseline) to load on the same factor, one alternative mo	dei specified
items of knowledge seeking, knowledge overload, and	
We thus modeled the positive noth of Irneviledes scaling on deily goal attains	mant via
We thus modeled the positive path of knowledge seeking on daily goal attains	
perceived learning and subsequent resource depletion. Given the potential sim	
existence of negative and positive paths that link knowledge seeking and daily	
attainment, it seems worthwhile to explore whetherhetotal indirect effect is po	
negative. This analysis sheds light on the relative strength of the opposing ind	
and clarifies whether the positive path from knowledge seeking via perceived	
resource depletion or the negative path via knowledge overload and resource	depletion is
more influential for daily goal attainment. This document is copyrighted by the	
Psychological Association or one of its allied publishers. This article is intend	
the personal use of the individual user and is not to be disseminated broadly. I	
Question: Is the total indirect effect of an employee's knowledge seeking on of	
attainment positive or negative, considering the negative path via knowledge of	
resource depletion and positive path via perceived learning and resource depletion	
Automated Quantitative Regression Estimates Predicting Academic Outcomes:Sample 2 Cross-Sample	
<u>Video Interview</u> Scores (Interviewer-ReportModels). Hierarchical regression conducted but no	CFA, or
Personality Bifactor methodology	
Personality Assessments: Reliability, Bifactor methodology	

Validity, and Generalizability		
Investigations  A Multirater Perspective on Personality and Performance: An Empirical Examination of the Trait— Reputation— Identity Model	Quantitative	Analyses for all TRI models were conducted using confirmatory factor analysis (CFA) and structural equation modeling (SEM) in Mplus version 7.4 (Muthén & Muthén, 1998–2012). The TRI Model is an application of a bifactor structure (Holzinger & Swineford, 1937; Reise, 2012) in which each item simultaneously loads on the trait factor and one source factor (see Figure 1).  The TRI Model is an application of a bifactor structure (Holzinger & Swineford, 1937; Reise, 2012) in which each item simultaneously loads on the trait factor and one source factor  All CFA and SEM models were specified to include relevant constraints where implied.  CONNELLY, MCABEE, OH, JUNG, AND JUNG  Figure 1  Trait  CONSELLY, MCABEE, OH, JUNG, AND
		B).

		Model fit estimates for all CFA and SEM models were subject to these adjustments where		
		implied .  Table 2  Model Fit Statistics for I-SAT Adjusted Trait—Reputation—Identity (TRI) Models (CFA = Confirmatory Factor Analysis; SEM = Structural Equation Modeling)		
		Big Five Trait/ $Y - B \chi^2_{ADJ}$ $Y - B \chi^2_{A$		
		CFA 296.77 (109) .965 .064 [.054, .073] 1591.90 SEM 1074.42 (585) .954 .045 [.034, .054] 3974.79 Agreeableness CFA 374.58 (145) .937 .061 [.051, .070] 1974.11 SEM 1139.44 (661) .946 .041 [.030, .051] 4493.56		
		Conscientiousness CFA 414.98 (143) 9.45 .067 [.057, .076] 2020.25 SEM 1177.34 (659) .950 .043 [.032, .053] 4537.21 Neuroticism CFA 411.69 (111) .878 .080 [.070, .090] 1701.07		
		SEM 109.11 (587) .934 .045 [.034, .055] 3986.76  Openness  CFA 787.02 (189) .879 .087 [.076, .097] 2713.92  SEM 1679.75 (744) .909 .055 [.045, .064] 5513.44		
		Note. $N = 422$ , I-SAT = interchangeable saturated model (Olsen & Kenny, 2006), $Y = B_{XMD}^2$ adjusted Yuan-Bentler scaled $\chi^2$ (Yuan & Bentler, 2000). CFI = comparative fit index; RMSEA = root mean square error of approximation [90% CI]; SABIC = sample adjusted Bayesian information criterion. The $Y = B\chi^2_{AD}$ statistic, $d\chi_{AD}$ and other fit statistics based on the $Y - B\chi^2$ and $df$ have been adjusted by subtracting out the corresponding values estimated in the I-SAT model for each trait. Data were clustered within training groups ( $k = 27$ ); all models were estimated in Mplus v. 7.4 (Muthén & Muthén, 1998–2012) using procedures for clustered survey data and robust maximum likelihood estimation. Small negative factor loadings were constrained to zero for all TRI models.		
Volume 107 Issue 9				
The Role of Fairness Perceptions in Patient and Employee Health: A Multilevel, Multisource Investigation	Quantitative	A confirmatory factor analysis demonstrated that the two factors were indeed separate (the two-factor model with Comparative Fit Index (CFI) = 0.99, Tucker Lewis Index (TLI) = 0.98, Root Mean Squared Error of Approximation (RMSEA) = 0.06, Standardized Root Mean Square Residual (SRMR) = 0.02), and in particular better fit than a one-factor model [CFI = 0.79, TLI = 0.69, RMSEA = 0.22,SRMR = 0.09;difference inchi-squared = 37,303 (1df); p < .001]. In addition, discriminant validity was confirmed using Fornell and Larcker's (1981) test, which showed average variance extracted of 0.56 and 0.66 for psychological safety and voice, respectively, clearly exceeding the squared correlation of 0.40. Table 1 also includes the aggregated version of organizational fairness, as this was used in some of the analyses with individual health as an outcome.		
You Are What You Eat: How and When Workplace Healthy Eating Cultivates Coworker Perceptions and Behaviors	Quantitative	We conducted a confirmatory factor analysis to assess the discriminant validity of perceived selfcontrol, citizenship behavior, and social undermining. We utilized an item parceling technique (Little et al., 2002; Sass & Smith, 2006)due to the high number of items we used to measure each construct (see Little et al., 2013), which has been done previously by scholars testing models with trait self-control (e.g., Fehr et al., 2017). This technique involves pairing the highest and lowest loading items together to form parcels, which is appropriate when the relationships between (rather than within) constructs are of interest (Little et al., 2013).		
201111111111111111111111111111111111111		We present the results of our path analysis in Table4. The path analysis revealed that our model accounted for an additional 10%, 12%, and 7% of the individual-level variance inself-control, citizenship behavior, and social undermining, respectively.		
Managing My Shame: Examining the Effects of Parental Identity Threat and	Quantitative	Second, we conducted a multilevel CFA with six factors within person (work-to-family conflict, family-to-work conflict, parental identity threat, shame, work productivity, investment in parenting) and eight factors between person (work-to-family conflict, emotional stability).		

	T	
Emotional		We tested our hypotheses via a multi-level path analysis model with observed variables in
Stability on		Mplus8.4(Muthén&Muthén,19982017). Level-1predictorsweregroup-
<u>Work</u>		meancenteredandLevel-2 predictors were grand-mean centered(Hofmannetal.,2000).By
Productivity and		group-mean centering Level-1predictors,we are able to effectively control for possible
<u>Investment in</u>		between-person methodological confounds such as social desirability and assess purely
<u>Parenting</u>		within-person relationships (Enders&Tofighi,2007).
I Know How I	Quantitative	No mention of models of interest
Feel but Do I		
Know How You		
Feel?		
<u>Investigating</u>		
Metaperceptions		
to Advance		
Relationship-		
Based		
Leadership		
Approaches		
Inclined but	Quantitative	No mention of models of interest
Less Skilled?	Quiumiziumizi v	
Disentangling		
Extraversion,		
Communication		
Skill, and		
Leadership		
Emergence		
When Leader	Quantitative	We conducted multilevel confirmatory factor analyses to investigate the factor
	Quantitative	distinctiveness of our variables. At Level 1, we modeled the items for positive affect,
Self-Care Begets		negative affect, leader identity, task-related helping, personal helping, leader competence,
Other Care:		i negative affect, leader identity, task-related helbing, bersonal helbing, leader combetence.
Landon Dala		
<u>Leader Role</u>		and leader civility, all loading on their respective factors. Factors were allowed to covary, as
Self-		
Self- Compassion and		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.
Self-		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4
Self- Compassion and		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable
Self- Compassion and		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader
Self- Compassion and		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence =
Self- Compassion and		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we
Self- Compassion and		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and
Self- Compassion and Helping at Work		and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we
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Self-Compassion and Helping at Work  Volume 107:	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and
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Self-Compassion and Helping at Work  Volume 107: Issue 5 Disaster or	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations
Self-Compassion and Helping at Work  Volume 107: Issue 5 Disaster or Opportunity?	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations
Self-Compassion and Helping at Work  Volume 107: Issue 5  Disaster or Opportunity? How COVID-19-Associated	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations
Volume 107: Issue 5 Disaster or Opportunity? How COVID- 19-Associated Changes in	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations
Volume 107: Issue 5 Disaster or Opportunity? How COVID- 19-Associated Changes in Environmental	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations
Volume 107: Issue 5 Disaster or Opportunity? How COVID- 19-Associated Changes in Environmental Uncertainty and	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations
Volume 107: Issue 5 Disaster or Opportunity? How COVID- 19-Associated Changes in Environmental Uncertainty and Job Insecurity	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations
Volume 107: Issue 5 Disaster or Opportunity? How COVID- 19-Associated Changes in Environmental Uncertainty and	Quantitative	and leader civility, all loading on their respective factors. Factors were allowed to covary, as is the default in Mplus.  We tested all hypotheses simultaneously using multilevel path analyses in Mplus 7.4 (Muthén & Muthén, 1998–2013). A null model revealed that there was a considerable amount of variance at the within-person level for all our endogenous variables (leader identity = 42.4%, task-related helping = 56.8%, personal helping = 52.0%, competence = 64.1%, and civility = 67.9%), supporting the use of multilevel modeling. In our analyses, we person-mean centered our Level-1 predictors, which removes between-person variance and allows for appropriate interpretation of within-person associations

T.1 ('C' ('		<u> </u>			
<u>Identification</u>					
and Performance					
The Push-and-	Quantitative	No mention of models of interest			
Pull of					
Frenemies:		To address the fact that the study design involved individuals nested within dyads and			
When and Why		account for the nonindependence of observations that arose because of interactions in those			
Ambivalent			hypotheses in a path model in which we clustered stand		
Relationships			onindependence (Angrist & Pischke, 2008). In this path		
		-		·	
Lead to Helping			oded variables using indicator coding: the first had pos		
and Harming			a 1, and the ambivalent and negative conditions coded		
<u>Organizational</u>	Quantitative	We used confirmatory	factor analysis (AMOS 23, Arbuckle, 2014) to exami	ne the structure	
Political		of the multiple-item su	urvey measures (Anderson & Gerbing, 1988). Table 4	shows that most	
Affiliation and		items loaded strongly	on their respective scales. In addition, the model demo	onstrated an	
Job Seekers: If I			p = 727.35 (p < .01), comparative fit index; CFI = .94,		
Don't Identify			mean square error of approximation; RMSEA = $.06$ , s		
•			= .05. Table 5 presents fit statistics for several alterna		
With Your					
Party, Am I Still		_	del, a single-factor model, and a model that constrains		
Attracted?		between liking and IT	P to unity (to test the discriminant validity of these two	o correlated	
		variables). These mod	els did not fit as well as our hypothesized model.		
		Table 4			
		Standardized Factor Loadings	From Confirmatory Factor Analysis of the Measurement Model (Study 3)		
		Construct	Item	Factor loading	
		Political identification	When someone criticizes the Democratic (Republican) party, it feels like a personal insult.	.85	
			I am very interested in what others think about the Democratic (Republican) party.  When I talk about the Democratic (Republican) party, I usually say "we" rather than "they."	.52 .81	
			When someone praises the Democratic (Republican) party, it feels like a personal compliment. If a story in the media criticized the Democratic (Republican) party, I would feel embarrassed.	.90 .77	
		Delitical disidentification			
		Political disidentification	I would be embarrassed to be part of the Democratic (Republican) party.  The Democratic (Republican) party does shameful things.	.90 .86	
			I find the Democratic (Republican) party to be disgraceful.  I want people to know that I disagree with how the Democratic (Republican) party behaves.	.92 .84	
			I have been ashamed of what goes on in the Democratic (Republican) party.	.86	
		Overall similarity	are similar in terms of our outlook and perspective.	.82	
		HBA organization and I	see things in much the same way think alike in many ways.	.88 .87	
			trink alike in many ways are alike in a number of areas.	.87	
		Organizational liking	have similar values.  I like this organization.	.73 .80	
		O-gammanonin manig	Working at this organization would be a pleasure.	.94	
			I would enjoy working at this organization.  I would likely get along well with the people of this organization.	.94 .69	
		Intentions to pursue employment	I think I would make good friends at this organization.  I would accept a job offer from this company.	.62 .79	
		mendons to pursue employment	I would make this company one of my first choices as an employer.	.83	
			If this company invited me for a job interview, I would go.  I would exert a great deal of effort to work for this company.	.58 .69	
			I would recommend this company to a friend looking for a job.	.71	
A 3 #	0	NT .	1 1 01 150	1 1	
A Meta-Analytic	Quantitative		xploratory analyses of the proposed D factor structure	• •	
Investigation of		dark traits(Moshagene	etal.,2018).Our analyses are exploratory because Mosh	agenetal.(2018)	
the Antecedents,		proposed a bifactor m	odel, where items load simultaneously on the latent D	factor and the	
Theoretical			are unable to test a bifactor model since we do not ha		
Correlates, and			em-level data. Instead, we specified a CFA model whe		
·		_	•	ac the dark traits	
Consequences of		loaded onto a latent,D	Tactor.		
<u>Moral</u>					
<u>Disengagement</u>		As shown in Table8,	noral disengagement is negatively related to moral aw	areness ( $\rho$ =24)	
at Work			o=49). Path analyses using the correlation matrix in T		
	1	(	,		

	Π	
		moral awareness mediates the effect of moral disengagement on misconduct
The Nieuth	Onertitet	(ρ=.02,95%CI=[.01,.03]),supporting Hypothesis15a.
The Nonlinear	Quantitative	No mention of models of interest
Relationship		
Between		
Atypical		
Applicant		
Experience and		
Hiring: The Red		
Flags		
<u>Perspective</u>		
A Meta-Analytic	Quantitative	Meta analysis was conducted.
Review of		No mention of models of interest
<u>Identification at</u>		
Work: Relative		
Contribution of		
Team,		
Organizational,		
and Professional		
<u>Identification</u>		
Conquering	Quantitative	No mention of CFA, but mention of path analysis
Unwanted		
Habits at the		Specifically, for these hypotheses' tests, we used multilevel path analysis (Preacher et al.,
Workplace:		2010) in which we specified the hypothesized paths both at the day and the person level. This
Day-Level		approach of partitioning all day-level variables into within-person and between-person
Processes and		component implies that the day-level variables are implicitly centered at the person level
Longer Term		(Preacher et al., 2010, p. 210), removing all between-person variance from the within person
Change in Habit		part of the model.
Strength		
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Issue 4		
Variable Work	Quantitative	I used Hierarchical Linear Modeling (HLM), treating units as Level 2 and unit-month
Schedules, Unit-	Quantitutive	observations as Level 1. Also, I included a Level 3 variable to account for the nesting of
Level Turnover,		units within U.S. states. For Hypotheses 4a and 4b testing, I used discontinuous growth curve
and Performance		modeling (Bliese & Lang, 2016; Singer & Willett, 2003) to test temporal trajectories of unit
Before and		performance over different phases of the pandemic.
During the		portormance over different phases of the pandenne.
COVID-19		
Pandemic		
	Quantitative	As such, we performed a more rigorous test of the measurement properties using a series of
Indexing Dynamic	Quantitative	nested confirmatory factor analyses (CFAs) of the MTMM correlations shown in Table 2
-		• • • • • • • • • • • • • • • • • • • •
Collective Constructs		following Widaman's (1985) study. Note that Table 2 includes correlations with some team
		demographic composition variables for completeness that we do not consider further. We show a summary of our model comparisons in Table 3.
Using Computer Aided		snow a summary of our model comparisons in Table 5.
Computer-Aided Tout Analysis		
Text Analysis:		
<u>Construct</u>		
<u>Validity</u>		
Evidence and		
<u>Illustrations</u>		

Featuring Team		
Processes		
Work-Leisure	Qualitative	
Blending: An		
<u>Integrative</u>		
Conceptual		
Review and		
Framework to		
Guide Future		
Research		
It's Not Only	Quantitative	We conducted a CFA (Hu & Bentler, 1999) to establish the distinctiveness of our variables.
What You Do,		We used random item parceling to reduce the number of indicators to three parcels for
But Why You		motives factor that included more than five items (Little et al., 2013). In addition, we used
Do It: How		the scale means of the four justice facets as indicators of justice (e.g., Liao, 2007; Sherf et al.,
<u>Managerial</u>		2019). For Study 1a, the 7-factor model (justice, prosocial [faceted], self-interest [faceted],
<u>Motives</u>		power, effecting compliance, identity maintenance, and establishing fairness) fit the data well $\chi 2$ [N = 146, df = 356] = 561.98, p < .001, CFI = .93, RMSEA = .06, SRMR = .06) and
Influence Employees'		all indicators loaded significantly on their intended factors.
Fairness		an indicators loaded significantly on their intended factors.
Judgments		
Vocational	Quantitative	To investigate how subgroup differences on predictors vary across primary interest job types
Interests and	Quantitutive	and application ratios for each predictor, we first obtain estimates of the relationships
Adverse Impact:		between race and each predictor while accounting for range restriction (Sackett & Yang,
How Attraction		2000). We are treating vocational interests as the application restriction variable, z, where
and Selection on		individuals enter the applicant pool on the basis of their vocational interests. The relevant
Vocational		range restriction scenario is Thorndike's Case 3 (Sackett & Yang, 2000; Thorndike, 1949).
Interests Relate		Newman and Cottrell (2015, p. 148) adapted the indirect range restriction formula by
to Adverse		combining it with Dobson's (1988) formula for variance of a truncated normal distribution
<b>Impact Potential</b>		(Schmidt et al., 1976) and formula for the ordinate (height) of a normal curve, to yield the
		equations:

### **Implications**

The trends found in this scholarly review conducted on 42 JAP articles published within the past year have shown that there is a significant lack of written clarity and reliance on less complex structures for their simplicity rather than their appropriateness found in the studies of top journal entries in I/O. This lack of detail causes conceptual confusion amongst the readers and leaves the process between the desired construct measurements and the actual models being

used to measure those constructs unclear. Of the articles that did implement what was considered thorough detail and clarity within their methodology sections, such as those that utilized CFA, most of these studies implemented CFA on a first order factor level instead of implementing more complex structures such as a hierarchical or bifactor analysis. Outside of CFA, some articles in our findings implemented a bifactor analysis within their methodology, however this was only present in a few of the articles analyzed. This finding suggests that the field seems to be reluctant to utilize more complex structures in their methodology even though they may be more appropriate to use in their studies. In our findings, after reviewing and analyzing the methodology in over 40 JAP articles within I/O, we encourage future researchers to provide more written clarity within their methodology section that provides a more detailed explanation of why they utilized the type of statistical modeling that they did in their studies. On top of providing a clear explanation of the structures used within studies, we suggest that researchers also consider in more detail which statistical model to utilize based on best fit for their study, rather than its simplicity.

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