- NFC and burnout in teachers A replication and extension study
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6 Abstract

The prevalence of burnout has been rising for years, not just due to the increasing demands during the Covid-19 pandemic. While it is known that burnout primarily affects employees in social jobs, less is known about the personality traits that promote or protect against burnout. One of these traits is Need for Cognition (NFC), the stable intrinsic 10 motivation to seek out and enjoy effortful cognitive activities. In the present study, we 11 analyzed questionnaire data of N = 180 teachers that had been collected in spring of 2020. 12 Firstly, we aimed to replicate results by Grass et al. (2018), who showed that the 13 association of NFC and the burnout aspect of reduced personal efficacy was mediated by habitual use of reappraisal, but not by habitual suppression or self-control. With our data, 15 self-control became a significant mediator when teaching experience was being taken into account, but neither reappraisal nor suppression mediated between NFC and reduced personal efficacy. Secondly, we computed a structural equation model to investigate 18 whether NFC and burnout were associated via different ratios of demands and personal 19 resources, and included other variables in an exploratory approach. The results indicated 20 that teachers with higher NFC and more self-control have lower burnout because they 21 experience their resources as fitting to the demands. 22

23 Keywords: mediation, resources, demands, structural equation modelling, Covid-19

24 Word count: 5600

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NFC and burnout in teachers - A replication and extension study

Introduction

Need for Cognition (NFC) is a stable intrinsic motivation to seek out and especially 27 to enjoy effortful cognitive activities (Cacioppo & Petty, 1982). As it bridges the gap 28 between cognition and motivation, NFC is considered to be an investment trait (Stumm & 29 Ackerman, 2013), and has come to the fore of psychological research in the last years. NFC can easily be assessed using the Need for Cognition Scale (NCS), a self-report questionnaire 31 with 18 to 34 items (Cacioppo et al., 1984; Cacioppo & Petty, 1982). While many studies have found positive associations of NFC with academic performance (Cazan & Indreica, 33 2014; Elias & Loomis, 2002; Grass et al., 2017; Lavrijsen et al., 2021; Zheng et al., 2020), recent investigations have also looked at NFC as a personal resource in academic and work 35 contexts. Individuals high in NFC have more positive emotions at the end of the work day (Rosen et al., 2020), higher work motivation, perceive their roles as less ambiguous (Nowlin et al., 2017), are less likely to drop out of college (Grass et al., 2017; Klaczynski & Fauth, 1996), and have less anxiety regarding their course work (Karagiannopoulou et al., 2020). These findings suggest that individuals high in NFC might be less prone to experience adverse effects of work stress, which range from physical (Dragano et al., 2017; Steptoe & Kivimäki, 2013) to psychological (Madsen et al., 2017; Maslach & Leiter, 2016; Wiesner et al., 2005). 43

One of these psychological consequences is burnout, a state of exhaustion and cynicism caused by long-term overstimulation in the workplace, which results in employees being dissatisfied, being sick more often, and performing poorly (Schaufeli & Salanova, 2014). Burnout is especially prevalent in social jobs such as healthcare or teaching because the worker is always in conflict between advocating for their client and meeting the goals set by the employer (Gray-Stanley & Muramatsu, 2011; Lloyd et al., 2002). Lackritz (2004) found that university teachers' burnout scores were higher the more students they had, the

- higher their teaching load was, and the more time they spent grading students' work.

 Burnout is most often assessed using the Maslach Burnout Inventory (MBI) (Maslach et al., 1997), a self-report questionnaire with three subscales: Emotional exhaustion,

 depersonalisation, and reduced personal efficacy.
- Individuals with high burnout scores are often passive copers, high in neuroticism, 55 low in self-esteem, and have an external locus of control (Schaufeli & Salanova, 2014). NFC on the other hand is negatively associated with those variables (Double & Birney, 2016; 57 Fleischhauer et al., 2019; Ghorbani et al., 2004; Grass et al., 2018; Osberg, 1987), suggesting that people high in NFC are less prone to experience burnout. This is 59 supported by findings that NFC is negatively associated with burnout scores in adults (Fleischhauer et al., 2019), students (Fleischhauer et al., 2019; Naderi et al., 2018), and 61 teacher trainees (Grass et al., 2018). However, the associations of NFC with the sum score and the subscales of the MBI are not always consistent between these studies. This is likely 63 not caused by inaccurate measurement, since the validity of both NCS (Bless et al., 1994; Osberg, 1987; Tolentino et al., 1990) and MBI (Brady et al., 2021; Kantas & Vassilaki, 1997; Schaufeli et al., 2001; Valdivia Vázquez et al., 2021) has been demonstrated in multiple studies. What is more likely is the influence of one or more other variables, moderating or mediating the association of NFC and burnout. Grass et al. (2018) investigated such a mediation and found that the relation of NFC and the MBI subscale reduced personal-efficacy was fully mediated by reappraisal, active coping, and passive coping, but not by suppression or self-control. Reappraisal and suppression are two 71 emotion regulation strategies, which refer to the cognitive reassessment of a stressor and the inhibition of emotional reactions, respectively (Gross, 1998). The findings by Grass et al. (2018) suggest that individuals high in NFC experience a weaker decline in personal efficacy in response to long-term stress because they actively reassess the situation in a way that reinforces their sense of self-efficacy and don't avoid dealing with the stressor. One goal of this paper was to replicate the findings of Grass et al. (2018) using a multiple

mediation model on cross-sectional self-report data of teachers. We expected NFC to be
negatively associated with reduced personal efficacy via higher reappraisal scores, but not
via suppression, via self-control, or directly.

Furthermore, we extended the analysis to other possible mediators. These mediators 81 were motivated by our own recent survey of the literature on NFC and well-being, which suggested that individuals high in NFC might not only have a high level of personal resources but also overestimate their own resources to a certain degree (Zerna et al., 2021). Only a balance of resources and demands results in personal well-being, while an imbalance threatens well-being, regardless of whether this imbalance is in favour of resources or demands (Dodge et al., 2012). Following the framework of Hobfoll (1989), resources can be 87 objects with practical or status purpose, conditions like marriage or tenure, personality aspects like coping style, and energies such as time, money, or knowledge. In the case of NFC, resources are from the categories personality and energies: Personality, because NFC is a trait, encompassing a curious, analytic, and passionate approach to challenges, and 91 energies, because individuals high in NFC have been coping actively all their life, which enriches their level of experience and knowledge in approaching challenges (Cacioppo et al., 1996). These personal resources matter with regards to stress assessment (how the situation is appraised) and with regards to both coping and recovery (Salanova et al., 2006). We therefore investigated whether the association of NFC and burnout was mediated by different ratios of demands and resources; demands that are too high to be dealt with using one's personal resources (DTH), demands that are too low for one's personal resources (DTL), and a balanced fit of demands and resources (DRF). Using the same data as for the replication, we computed a structural equation model (SEM) to assess the influence of these mediators. Since individuals high in NFC are confident in their abilities (Bye & Pushkar, 2009; Ghorbani et al., 2004; Heppner et al., 1983; Klaczynski & 102 Fauth, 1996), we expected NFC to be negatively associated with DTH, positively 103 associated with DTL, and positively associated with DRF. And since burnout results from 104

constant unpleasant activation by high demands, we expected it to be positively associated 105 with DTH and negatively associated with DRF. However, we had no hypothesis regarding 106 the association of DTL and burnout, because even though DTL is akin to the concept of 107 boredom and the consequences of boredom and burnout are very similar, burnout is a state 108 with even lower activation and even more negative affect than boredom (Schaufeli & 100 Salanova, 2014). It has already been shown that the Covid-19 pandemic has exacerbated 110 the rising prevalence of burnout (Fröbe & Franco, 2021), so we incorporated the degree of 111 feeling burdened by the pandemic in an exploratory approach. 112

113 Methods

We report how we determined our sample size, all data exclusions (if any), all manipulations, and all measures in the study (Simmons et al., 2012). Our preregistration, the data, and the R Markdown document used to analyze the data and write this manuscript are available at https://osf.io/36ep9/.

118 Participants

Teachers were recruited via social media, emails to colleagues of N.E., and to Saxon 119 schools with the request to pass on the information. All teachers were eligible, no payment 120 was issued. Of the N=278 participants, who started filling out the online survey, N=180121 (72.20% female, aged 20 to 67 years) data sets were complete and those participants 122 indicated to have answered truthfully. All of them were currently teaching at a primary, 123 secondary, comprehensive, or vocational school. Data was collected between the 12th of June and the 24th of July 2020. At this point, schools had been switching between digital 125 and hybrid forms of teaching for at least three months due to the Covid-19 pandemic, 126 causing additional stress for many teachers. 127

128 Material

All questionnaires were used in their German form. Burnout was assessed using the 129 21-item Maslach Burnout Inventory (MBI) (Büssing & Perrar, 1992), NFC using the 130 16-item Need for Cognition Scale (NCS) (Bless et al., 1994), self-control using the 13-item 131 Self-Control Scale (SCS) (Bertrams & Dickhäuser, 2009), reappraisal and suppression using 132 the 10-item Emotion Regulation Questionnaire (ERQ) (Abler & Kessler, 2009), and work 133 satisfaction using the Allgemeine Arbeitszufriedenheit questionnaire (Fischer & Lück, 134 2014). Eleven items were created to assess each participant's current burden by the 135 Covid-19 pandemic, such as whether they belong to a risk group or whether they currently 136 had a higher workload. The translated Covid-19 items can be found in the Supplementary 137 Material S1. The survey also included the Subjective Wellbeing Index of the World Health 138 Organization (Brähler et al., 2007), which we will not analyze. Due to a technical error 139 during survey setup, the coping style data of the Erfurter Belastungsinventar 140 (Böhm-Kasper et al., 2001) could not be used, so we could not replicate the mediation of 141 NFC and burnout by active and passive coping.

143 Procedure

The questionnaires were provided online using SoSci Survey (Leiner, 2019).

Participants were informed about aims and duration of the study and data security,

provided demographic information, answered the questionnaires, and could optionally enter

their email address to be informed about the results of the analysis of N.E.'s thesis.

48 Data analysis

We used *R Studio* (R Core Team, 2020; RStudio Team, 2020) with the main packages lavaan (Rosseel, 2012) and psych (Revelle, 2021) for all our analyses. Data were checked for multivariate normality using Mardia's coefficient. To account for non-linear

relationships, correlations were computed using Spearman's rank coefficient rather than
Pearson's product moment correlation. Internal consistencies were assessed with
Cronbach's Alpha and MacDonald's Omega. Since Cronbach's Alpha has been criticized
for being insensitive to violations of internal consistency (Dunn et al., 2014; Taber, 2018),
the additional computation of MacDonald's Omega has the purpose of ensuring a more
reliable estimation.

Replication of Grass et al. (2018). Items were reverse coded according to the 158 scale manuals. NFC and self-control were computed as the sum scores of the NCS and the 159 SCS, respectively. Reduced personal efficacy was computed using the sum of the MBI 160 subscale, and reappraisal and suppression were computed using the sum of each ERQ 161 subscale. NFC was entered as the independent variable, having a direct and multiple 162 indirect effects on MBI via self-control, reappraisal, and suppression as mediators. 163 Following Grass et al. (2018), the results of the model were appraised by using $N=2{,}000$ 164 bootstrap samples for confidence intervals. Multiple indices were used to evaluate model fit 165 as recommended by Hu and Bentler (1999): the Chi-square test statistic, which measures 166 the fit compared to a saturated model, the Comparative Fit Index (CFI), which compares 167 the fit to the baseline model, the Standardized Root Mean Square Residual (SRMR), which compares the residuals of the observed and predicted covariance matrix, and the Root Mean Square Error of Approximation (RMSEA), which does the same as the latter but takes degrees of freedom and model complexity into account. 171

Demand-resource-ratio model. All items, apart from those making up the
demand-resource-ratios, were reverse coded according to the scale manuals. The latent
factor NFC was computed by subjecting the NCS items to a parcelling procedure following
Grass et al. (2019), a method that is used in SEM when only relations between but not
within constructs are of interest. Principal component analysis was used to determine the
factor loadings of each NCS item onto the first component. Then, the items were randomly
divided into four parcels and the average item loading per parcel was computed. This was

repeated 10,000 times to find the parcelling choice with the smallest difference in average 179 item loadings between parcels. The latent factor MBI was computed using the three 180 subscales as indicators. For the demand-resource-ratios, we used three items from the work 181 satisfaction scale each. The latent factor DTH was indicated by items 4, 8, and 9, DTL by 182 the recoded items 12, 26, and 27, and DRF by items 17, 22, and 24. The items can be 183 translated as follows: 4) "There is too much pressure on me." 8) "There is often too much 184 being demanded of us at work." 9) "I often feel tired and weary because of my work." 12) 185 "I can realize my ideas here." 17) "I take pleasure in my work." 22) "Does your place of 186 work give you the opportunity to do what you do best?" 24) "Does your place of work give 187 you enough opportunities to use your skills?" 26) "Are you happy with your promotion 188 prospects?" and 27) "Are you happy with your position when comparing it to your skills?" 189 Model parameters were estimated using the maximum likelihood method with robust standard errors. Model fit was evaluated by looking at the Chi-square test statistic, CFI, SRMR, and RMSEA. 192

Exploratory analyses. We preregistered two exploratory analyses. Firstly, we repeated the SEM with the subscale reduced personal efficacy in place of the MBI score, since this subscale has shown higher correlations with NFC than the other subscales (Grass et al., 2018; Naderi et al., 2018). And secondly, we included a Covid-19 burden score into the SEM, computed as the sum of the Covid-19 items.

198 Results

During visual inspection of correlation plots we noticed an unexpected outlier with very high MBI scores and very low NFC scores. A Q-Q-plot contrasting Mahalanobis D^2 against expected Chi Square values confirmed the outlier. To adhere to the preregistration, we report the results containing the outlier in this section and the results excluding the outlier in the Supplementary Material S2.

204 Descriptive statistics

Basic metric descriptives of the questionnaire scores and subscales are listed in Table
1. Only the ERQ sum score and its reappraisal subscale followed a multivariate normal
distribution, so the results of the models should be interpreted with some caution and with
a focus on indices that are robust against violation of normality, such as the
Satorra-Bentler or Yuan-Bentler-scaled test statistics (Rosseel, 2012).

Table 1
Descriptive statistics of the questionnaire scores.

Variable	Minimum	Maximum	Mean	SD	Normality	Skewness	Kurtosis
MBI	27	101	52.93	13.06	No	0.35	0.02
MBI EE	12	52	27.99	8.87	No	0.19	-0.59
MBI DP	5	24	9.72	3.26	No	0.82	0.86
MBI RPE	7	28	15.22	3.43	No	0.42	1.11
ERQ	16	63	39.18	7.82	Yes	-0.16	0.45
ERQ S	4	26	12.59	4.85	No	0.14	-0.73
ERQ R	9	42	26.59	6.29	Yes	-0.05	0.01
SCS	-19	23	7.79	8.42	No	-0.39	-0.22
NFC	-34	48	20.37	14.04	No	-0.59	0.56
DTH	-6	6	0.49	2.65	No	-0.15	-0.56
DTL	-6	6	-2.22	2.24	No	0.46	0.28
DRF	-4	6	3.63	1.79	No	-0.91	1.75
COV	14	33	24.53	4.28	No	-0.14	-0.70

Note: MBI = Maslach Burnout Inventory, MBI EE = Emotional exhaustion subscale, MBI DP = Depersonalisation subscale, MBI RPE = Reduced personal efficacy subscale, ERQ = Emotion Regulation Questionnaire, ERQ S = Suppression subscale, ERQ R = Reappraisal subscale, SCS = Self-Control Scale, NFC = Need for Cognition, DTH = Demands Too High, DTL = Demands Too Low, DRF = Demand-Resource-Fit, COV = Covid-19 Burden, SD = Standard deviation. N = 180.

Correlations and internal consistencies are displayed in Table 2. For this descriptive
analysis, the variables DTH, DTL, and DRF were computed as a sum of their item scores,
not weighted as in the structural equation model. Using traditional cut-off values
(Nunnally & Bernstein, 1994), the Cronbach's Alpha of the three demand-resource-ratios
can be considered acceptable. The more robust MacDonald's Omega (Dunn et al., 2014)

did not deviate much from Cronbach's Alpha and indicated acceptable to good internal consistency. As expected, the MBI score was positively correlated with DTH ($r_s = .67$, p < .001) and negatively with DRF ($r_s = -.55$, p < .001), large associations according the classification scheme by Gignac and Szodorai (2016). Surprisingly, the correlation between the MBI score and DTL was positive and also large ($r_s = .44$, p < .001). The NFC score correlated negatively with the MBI sum score and about equally with all subscales, contrary to some previous observations in other studies.

Table 2
Spearman correlations and internal consistencies of the questionnaire scores.

	1	2	3	4	5	6	7	8	9	10	11	12	13
1. MBI	0.90(0.91)												
2. MBI EE	0.925***	0.91(0.92)											
3. MBI DP	0.748***	0.535***	0.68(0.69)										
4. MBI RPE	0.669***	0.434***	0.480***	0.79(0.79)									
5. ERQ	-0.058	-0.059	0.043	-0.099	0.73(0.62)								
6. ERQ S	0.053	-0.000	0.166*	0.076	0.592***	0.75(0.79)							
7. ERQ R	-0.101	-0.058	-0.061	-0.197**	0.715***	-0.075	0.84(0.84)						
8. SCS	-0.342***	-0.283***	-0.368***	-0.185*	-0.034	-0.121	0.050	0.85(0.86)					
9. NFC	-0.248***	-0.196**	-0.219**	-0.213**	-0.008	-0.176*	0.158*	0.216**	0.89(0.89)				
10. DTH	0.665***	0.722***	0.348***	0.365***	0.029	0.054	-0.006	-0.207**	-0.148*	0.73(0.73)			
11. DTL	0.444***	0.358***	0.379***	0.431***	0.007	0.158*	-0.136	-0.191*	-0.162*	0.409***	0.73(0.75)		
12. DRF	-0.545***	-0.457***	-0.410***	-0.531***	-0.005	-0.096	0.097	0.177*	0.241**	-0.420***	-0.561***	0.77(0.77)	
13. COV	0.241**	0.324***	0.083	0.016	-0.028	0.019	-0.065	-0.040	0.125	0.447***	0.095	-0.130	0.77(0.81)

Note: MBI = Maslach Burnout Inventory, MBI EE = Emotional exhaustion subscale, MBI DP = Depersonalisation subscale, MBI RPE = Reduced personal efficacy subscale, ERQ = Emotion Regulation Questionnaire, ERQ S = Suppression subscale, ERQ R = Reappraisal subscale, SCS = Self-Control Scale, NFC = Need for Cognition, DTH = Demands Too High, DTL = Demands Too Low, DRF = Demand-Resource-Fit, COV = Covid-19 Burden. N = 180. * p < .05. *** p < .01. *** p < .001. Diagonal is Cronbach's Alpha and (in brackets) MacDonald's Omega.

222 Replication of Grass et al. (2018)

In order to replicate findings by Grass et al. (2018) we computed a multiple 223 mediation model to investigate whether the association of NFC and reduced personal 224 efficacy was partially mediated by self-control and habitual use of reappraisal and 225 suppression, respectively. The baseline model did not fit the data ($\chi^2(10, N = 180)$) 226 49.64, p < .001). Applying the cutoffs by Hu and Bentler (1999) to the fit indices of 227 $CFI=1,\,TLI=1.14,\,SRMR=0.02,\,\mathrm{and}\,\,RMSEA=0.00,\,95\%\,\,CI$ [0,0.09], suggested 228 good fit of the proposed model throughout all indices. Standardized estimates are 220 displayed in Figure 1, total, direct, and indirect effects are listed in Table 3. We could 230 replicate a positive association of NFC and self control ($\beta = 0.27$, p = 0.00), and a negative 231 association of habitual reappraisal and reduced personal efficacy ($\beta = -0.17$, p = 0.01). 232 However, we could neither replicate the effect of NFC on reappraisal ($\beta = 0.12$, p = 0.10), 233 nor the indirect effect of NFC on reduced personal efficacy via reappraisal ($\beta = -0.02$, p =0.15). Furthermore, even though NFC and reduced personal efficacy were both associated 235 with self-control, the indirect effect of NFC on reduced personal efficacy via self control did 236 not reach significance ($\beta = -0.05$, p = 0.09). Additionally, NFC was negatively associated 237 with habitual use of suppression ($\beta = -0.18$, p = 0.01), which was not the case in the study 238 by Grass et al. (2018). 239

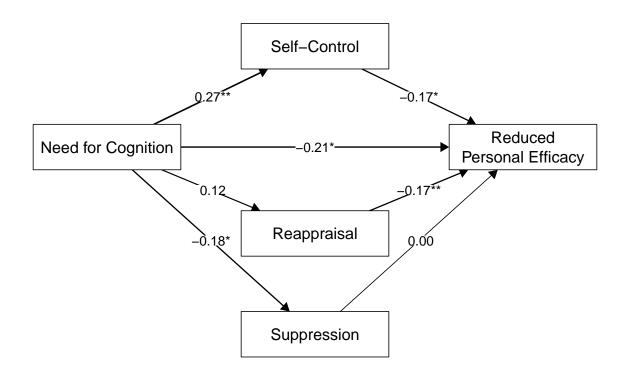


Figure 1. Standardized regression coefficients in the replication of Grass et al. (2018). * p < .05, ** p < .01.

Grass et al. (2018) controlled for age and a-level grade in their analysis, which we did 240 not consider when preregistering this analysis. Since grade was not assessed in this sample, and age was assessed as a categorical variable, we instead incorporated how many years 242 each participant had spent teaching at the point of assessment. We placed this variable as 243 an independent variable influencing self control, as the latter was the only variable in the 244 model that showed a partial correlation with years spent teaching. As it was not preregistered, this was an exploratory analysis. Again, the baseline model did not fit the data $(\chi^2(14, N = 180) = 60.41, p < .001)$, and the fit indices of CFI = 1, TLI = 1.19, SRMR = 0.02, and RMSEA = 0.00, 95% CI [0,0.04], suggested good fit of the proposed model throughout all indices. Standardized estimates, total, direct, and indirect effects are 249 displayed and listed in Supplementary Material S3. The associations between NFC, self

Table 3
Results of the replication of Grass et al. (2018).

Path	В	SE	z-value	<i>p</i> -value	CI Lower	CI Upper	β
Direct Effects							
NFC on Self Control	0.162	0.051	3.154	0.002	0.055	0.258	0.271
NFC on Reappraisal	0.055	0.034	1.619	0.105	-0.011	0.120	0.123
NFC on Suppression	-0.063	0.025	-2.524	0.012	-0.113	-0.017	-0.182
Self Control on RPE	-0.069	0.030	-2.318	0.020	-0.126	-0.009	-0.169
Reappraisal on RPE	-0.094	0.036	-2.652	0.008	-0.159	-0.023	-0.173
Suppression on RPE	0.002	0.051	0.043	0.966	-0.094	0.106	0.003
NFC on RPE	-0.051	0.021	-2.473	0.013	-0.089	-0.008	-0.208
Indirect Effects							
NFC on RPE via Self Control	-0.011	0.007	-1.695	0.090	-0.026	-0.001	-0.046
NFC on RPE via Reappraisal	-0.005	0.004	-1.429	0.153	-0.013	0.001	-0.021
NFC on RPE via Suppression	0.000	0.004	-0.039	0.969	-0.008	0.006	-0.001
Total Effect							
Total Effect	-0.067	0.023	-2.957	0.003	-0.111	-0.021	-0.276

Note: B = unstandardized regression coefficient, beta = standardized regression coefficient, CI = confidence interval, NFC = Need for Cognition, RPE = reduced personal efficacy subscale of the Maslach Burnout Inventory, SE = standard error.

control, reappraisal, suppression, and reduced personal efficacy were almost identical to the first model. However, because of the positive association of years spent teaching and self control ($\beta = 0.22$, p < .001), the indirect path leading from NFC and years spent teaching via self control to reduced personal efficacy reached significance in this model ($\beta = -0.09$, p = 0.05). Therefore, the total effect also increased slightly, compared to the first model ($\beta = -0.32$, p = 0.00).

57 Demand-Resource Model

Next we looked at how different ratios of subjective demands and resources affect the association of NFC and burnout. The parcelling procedure for the indicators of the latent factor NFC resulted in four parcels with a summed difference in average loadings of 0.00.
The first parcel contained item 4, 6, 8, and 9, the second parcel item 2, 14, 15, and 16, the third parcel item 7, 11, 12, and 13, and the fourth parcel item 1, 3, 5, and 10. Standardized path coefficients of the demand-resource model are illustrated in Figure 2, total, direct, and

indirect effects are listed in Table 4. The robust Chi-square statistic of $\chi^2 = 399.08$ 264 (p < .001) did not indicate good model fit. However, since it was in the range of 4 265 $*df < \chi^2 < 5 *df$ the lack of good fit might have been due to the underlying assumption of 266 multivariate normality (Hu & Bentler, 1999; Schumacker & Lomax, 2012), which was 267 violated here. This also held true for the CFI of 0.78, the SRMR of 0.17, and the RMSEA 268 of 0.13, 95% CI [0.12,0.14]. Overall, the fit indices did not support the proposed model, 269 and not all proposed paths were significant. NFC showed no direct association with the 270 MBI score ($\beta = 0$, p = 0.99), even though it was negatively correlated with the sum score 271 and all subscales. Instead, NFC showed indirect negative associations with the MBI score 272 via lower scores in the latent variable DTH ($\beta = -0.20$, p = 0.03) and via higher scores in 273 the latent variable DRF ($\beta = -0.13$, p = 0.03). The latent variable DTL was neither related 274 to NFC ($\beta = -0.18$, p = 0.13) nor to the MBI score ($\beta = 0.11$, p = 0.20).

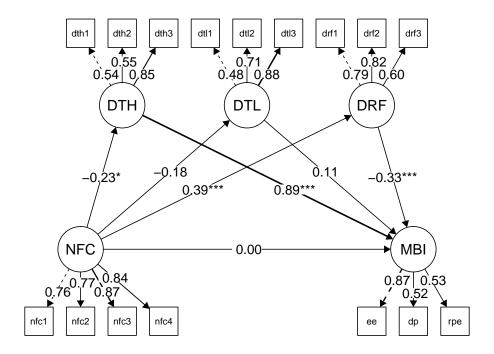


Figure 2. Standardized path coefficients in the mediation of NFC and burnout by demand-resource-ratios. * p < .05, ** p < .01, *** p < .001. NFC = Need for Cognition, DTH = demands too high, DTL = demands too low, DRF = demand resource fit, MBI = Maslach Burnout Inventory, nfc1-4 = item parcels, dth/dtl/drf1-3 = item indicators, ee = emotional exhaustion, dp = depersonalisation, rpe = reduced personal efficacy.

276 Exploratory analyses

The first exploratory analysis concerned a modification of the demand-resource-model in which the subscale reduced personal efficacy would be used in place of the MBI sum score. Path coefficients, total, direct, and indirect effects are displayed and listed in Supplementary Material S4. Similar to the previous model, this model's indices did not indicate good fit, with a Chi-square statistic of $\chi^2 = 247.82$ (p < .001), a CFI of 0.83, a SRMR of 0.17, and a RMSEA of 0.12, 95% CI [0.10,0.13]. NFC showed no direct association with reduced personal efficacy ($\beta = -0.05$, p = 0.55), but an indirect one via

Table 4
Results of the demand-resource-ratio model.

Path	В	SE	z-value	<i>p</i> -value	CI Lower	CI Upper	β
Direct Effects							
NFC on DTH	-0.042	0.020	-2.154	0.031	-0.081	-0.004	-0.228
NFC on DTL	-0.023	0.015	-1.522	0.128	-0.052	0.007	-0.180
NFC on DRF	0.070	0.020	3.488	0.000	0.031	0.110	0.386
NFC on MBI	0.002	0.144	0.014	0.989	-0.281	0.285	0.001
DTH on MBI	10.624	2.229	4.767	0.000	6.256	14.991	0.892
DTL on MBI	1.838	1.428	1.287	0.198	-0.960	4.637	0.106
DRF on MBI	-4.036	1.080	-3.736	0.000	-6.153	-1.918	-0.332
Indirect Effects							
NFC on MBI via DTH	-0.451	0.203	-2.221	0.026	-0.848	-0.053	-0.203
NFC on MBI via DTL	-0.042	0.033	-1.270	0.204	-0.107	0.023	-0.019
NFC on MBI via DRF	-0.284	0.127	-2.236	0.025	-0.533	-0.035	-0.128
Total Effect							
Total Effect	-0.775	0.258	-3.003	0.003	-1.280	-0.269	-0.349

Note: B = unstandardized regression coefficient, beta = standardized regression coefficient, CI = confidence interval, DTH = Demands Too High, DTL = Demands Too Low, DRF = Demand Resource Fit, MBI = Maslach Burnout Inventory, NFC = Need for Cognition, SE = standard error.

higher scores in the latent variable DRF (β = -0.22, p = 0.00). And again, NFC was associated with lower scores in the latent variable DTH (β = -0.22, p = 0.03), but the latter did not mediate the relationship between NFC and reduced personal efficacy (β = -0.03, p = 0.24) as it did with the MBI score in the previous model. The latent variable DTL was neither related to NFC (β = -0.19, p = 0.10) nor to the MBI score (β = 0.11, p = 0.20).

The second exploratory analysis concerned the incorporation of the Covid burden score into the model. We based the development of this model on the partial correlations of all variables, which provide an indication of how closely or remotely related variables might be in a path model. Then we modified the structure of the model using the modincides()-function in lavaan in order to increase the goodness-of-fit indices within the framework of contentually meaningful variable relationships. The final model is illustrated in Figure 3, the total, direct, and indirect effects are listed in Supplementary Material S5.

All fit indices suggested that the proposed model had good fit while the baseline model did

 $\mathrm{not}~(\chi^2=130.13~(p<.001),~CFI=0.95,~RMSEA=0.07~(95\%~CI~[0.05,0.08]),~SRMR=0.001,~CFI=0.001,~CF$ 297 0.06). Neither the ERQ sum score, nor its subscales, nor the depersonalisation subscale of 298 the MBI contributed significantly to the explained variance and were therefore not included 299 in the final model. Years spent teaching was associated with higher self control ($\beta = 0.21$, 300 p = 0.00) and higher Covid burden ($\beta = 0.17, p = 0.02$) but not with NFC. NFC covaried 301 with self control ($\sigma_{NFC,scs} = 0.31$, p = 0.01) and Covid burden ($\sigma_{NFC,covb} = 0.19$, p = 0.02), 302 but not with years spent teaching (p = 0.72). In turn, NFC was associated with higher 303 DRF scores ($\beta = 0.34$, p = 0.00) and lower DTH scores ($\beta = -0.21$, p = 0.01) but not 304 directly with any of the two MBI subscales. DRF scores fully mediated the negative 305 association of NFC and self-control with reduced personal efficacy (indirect effect $\beta = -0.29$, 306 p = 0.00), which was also true for DTH scores and emotional exhaustion, but DTH also 307 partially mediated between Covid burden and emotional exhaustion (indirect effect $\beta =$ -0.18, p = 0.01). Covid burden was not associated with DRF or reduced personal efficacy.

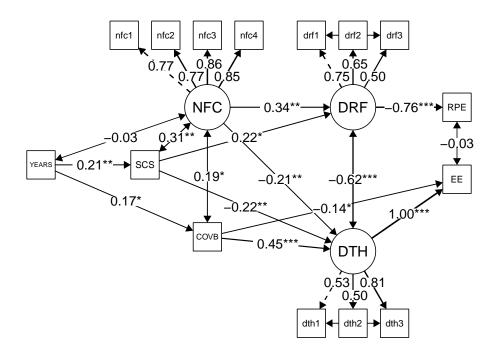


Figure 3. Standardized path coefficients in the exploratory analysis of variable relations. * p < .05, ** p < .01, *** p < .001. Years = years spent teaching, SCS = Self-Control Scale, COVB = Covid burden, NFC = Need for Cognition, DTH = demands too high, DRF = demand resource fit, nfc1-4 = item parcels, dth/drf1-3 = item indicators, EE = emotional exhaustion, RPE = reduced personal efficacy.

Discussion

The present study aimed to replicate findings of mediators between Need for
Cognition and burnout in teachers, as well as to extend the analysis to the role of different
ratios of demands and resources in burnout using latent variable models. In an exploratory
approach, we investigated the influence of the burden that the Covid-19 pandemic has
placed on teachers. Previous studies have indicated a protective effect of NFC against
burnout, but the associations with the burnout subscales were inconsistent, suggesting that
there are more variables influencing this relationship.

Replication of Grass et al. (2018)

While the mediation model had good fit, not all patterns were similar to the original study: NFC and self-control were positively associated, and reappraisal and reduced personal efficacy were negatively related, but there was no association between NFC and reappraisal. There was, however, a positive association between self-control and reduced personal efficacy, and a negative one between NFC and suppression.

NFC had a direct and negative effect on reduced personal efficacy, but this 324 relationship was not mediated by any other variable. Only when the amount of teaching experience was included as a predictor of self-control next to NFC, an indirect effect via self-control reached significance, indicating that teachers with high NFC and more years of teaching experience have higher self-control and, consequently, lower reduced personal efficacy. The higher self-control that comes with more teaching experience is in line with 320 findings of fluctuations in self-control in young adults, reaching a low point between the age of 15 and 19 (Oliva et al., 2019). The participants in the study by Grass et al. (2018) 331 were teacher trainees with a mean age of 25.5 years, while the majority of the current 332 sample was between 40 and 59 years old. Therefore, it is likely that not only the teaching 333 experience itself but also higher age might be associated with higher self-control. However, 334 one could argue that more experience provides the teacher with a bigger repertoire of 335 coping strategies to enable an efficient exertion of self-control, especially for teachers high 336 in NFC who are intrinsically motivated to find and apply such strategies. 337

We could replicate the relation between the two emotion regulation strategies
reappraisal and suppression with reduced personal efficacy, but not their association with
NFC. There is ample evidence that reappraisal is associated with positive outcomes for
students (Haga et al., 2007; Levine et al., 2012; Schmidt et al., 2010) and teachers alike
(Jiang et al., 2016; Moè & Katz, 2020; Tsouloupas et al., 2010), so it is suprising that
reappraisal did not mediate between NFC and reduced personal efficacy. Reappraisal did

correlate with NFC, as it should appease the preference for cognitive effort in individuals
with high NFC, but it was not a mediator in this model. One possible explanation could be
that the ways by which reappraisal can be achieved, such as taking the role of an
uninvolved observer, are less feasible for teachers in retaining their sense of efficacy in the
classroom than the self-control needed to structurally manage students and situations.

Hence, the mediation of NFC and reduced personal efficacy by self-control when taking the
years spent teaching into account.

Demand-resource-ratio model

Despite not having good fit indices, the model suggested a complete mediation of 352 NFC and burnout via DTH and DRF but not DTL. Specifically, individuals with higher 353 NFC had lower burnout scores through perceiving demands as fitting to and not exceeding 354 their own resources. Interestingly, the correlation between NFC and burnout, which can be 355 classified as medium according to Gignac and Szodorai (2016), disappeared in the context 356 of the demand-resource-ratios as mediators. The mediator that did not reach significance 357 was the perception of own resources exceeding the job demands. As this latent variable was 358 conceptualized as boredom at work, we could not confirm the positive association of 350 boredom and burnout found by Reijseger et al. (2013.). The fact that the items that made up the demand-resource-ratios were about the subjective perception and not about 361 objective measures, supports the idea that the individual appraisal of one's own 362 circumstances plays a crucial role in the development of burnout. This individual appraisal 363 has been emphasized as the cause for the ambiguous impact of demands on psychological well-being before, in the form of challenge demands and hindrance demands (Lazarus & Folkman, 1984; Lepine et al., 2005; Podsakoff et al., 2007). Challenge demands such as time pressure, responsibility, and workload (Podsakoff et al., 2007) are being positively valued due to their potential to increase personal growth, positive affect, and 368 problem-focused coping (Lepine et al., 2005). In contrast, hindrance demands such as

inadequate resources, role conflict, and organisational politics (Podsakoff et al., 2007) are perceived as negative because they harm personal growth, trigger negative emotions, and 371 increase passive coping (Lepine et al., 2005). Ventura et al. (2015) found that hindrance 372 but not challenge demands were positively related to burnout in teachers, and teachers who 373 reported high challenge and low hindrance demands also reported higher engagement. 374 Whether and to what extent a circumstance is perceived as a challenge or hindrance 375 demand is highly influenced by a person's level of self-efficacy (Bandura, 1997), so much so 376 that a reduction in self-efficacy is considered to be a precurser of burnout, not necessarily a 377 symptom (Cherniss, 1993; M. Vera et al., 2012). Self-efficacy and self-control are closely 378 entwined (Przepiórka et al., 2019; E. M. Vera et al., 2004; Yang et al., 2019) and both are 379 positively associated with NFC (Bertrams & Dickhäuser, 2012; Holch & Marwood, 2020; 380 Naderi et al., 2018; Xu & Cheng, 2021). Cacioppo et al. (1996) even proposed that higher levels of NFC might develop as a result of a high need for structure or control in those who have the skill, ability, and inclination to do so. These associations would imply that teachers with high levels of NFC report lower levels of burnout because their higher (desire for) self-control motivates them to appraise demands as a chance for personal growth, 385 thereby meeting their passion for thinking and problem-solving. Nevertheless, appraisal is no universal remedy for circumstances that threaten well-being, as there certainly are 387 circumstances that one cannot get any benefit out of. It remains an open question whether 388 a high desire for control and high NFC might cloud one's judgement in this case, by 389 encouraging to invest one's own insufficient resources in order to meet these high external 390 demands. Such behavioural tendencies would threaten personal well-being in the long 391 term, as the demands cannot be met, self-efficacy declines, and stress increases. 392

Exploratory analyses

Demand-resource-ratio model with subscale. The demand-resource-ratio model with the subscale reduced personal efficacy in place of the MBI score did not have

good fit indices. Compared to the confirmatory demand-resource-ratio model, the 396 mediation of NFC and reduced personal efficacy via DTH did not reach significance, but 397 both the mediation via DRF and the total effect remained significant. Overall, this pattern 398 did not resemble those from previous studies in which NFC had the strongest relation with 399 this subscale of the MBI (Grass et al., 2018; Naderi et al., 2018). Teachers with high NFC 400 appear to retain their sense of personal efficacy to a higher degree, because they experience 401 a fit of demands and resources, which allows them to complete tasks and reinforce their 402 self-efficacy in return. However, while this association was similar in the confirmatory and 403 the exploratory demand-resource-ratio model, the mediation via DTH was not significant 404 with this subscale, suggesting that the large association of DTH and MBI in the 405 confirmatory model was driven by a different subscale. To explore this, we built a second 406 exploratory model.

Exploratory model with Covid burden. Due to the complete freedom in 408 setting up the structure of this model, it had good fit indices. Interestingly, the third MBI 409 subscale depersonalisation and the latent variable DTL did not explain any variance in the 410 model, so they were removed. Once again, NFC and self-control were positively related, 411 but NFC was also positively related to Covid burden. One possible explanation is that 412 teachers with higher NFC show higher consideration of the consequences and progression of 413 the pandemic, thereby anticipating that it will take a long time until normal teaching can 414 resume, which heightens their feeling of being burdened. Although NFC has been shown to 415 be related to more reflective thinking and unrelated to rumination, which are considered 416 healthy and unhealthy thinking styles, respectively (Nishiguchi et al., 2018; Vannucci & Chiorri, 2018), a higher perceived Covid burden itself cannot indicate whether it stems 418 from a realistic view on the pandemic or a feeling of being overwhelmed. Teachers with more years of experience also reported higher Covid burden, presumably because older 420 people are less comfortable with technology (Hauk et al., 2018) and therefore stressed by 421 the prospect of online teaching. Teachers with higher self-control and higher NFC reported

a stronger fit of demands and resources, which was associated with a strong decrease in reduced personal efficacy. Higher self-control, higher NFC, and lower Covid burden was in 424 turn associated with a lower DTH score, so teachers with those characteristics felt less 425 overwhelmed and consequently less emotionally exhausted. The degree of association 426 between DTH and emotional exhaustion indeed suggested a congruence between the two, 427 indicating that emotional exhaustion in burnout is caused by excessive demands that 428 cannot be met with one's resources, while reduced personal efficacy in burnout is caused by 420 a lack of opportunities to utilize one's resources at work. Curiously, higher Covid burden 430 also showed a small negative association with emotional exhaustion. It could be that for 431 some teachers, remote teaching was experienced as a relief from the strain of dealing with a 432 group of over twenty students each day, who are more likely to misbehave in a classroom 433 setting than when they are studying at home. So while those teachers did feel the pandemic burden, they also felt less emotionally exhausted. 435

Limitations and future implications

The data used in this study had been collected for another purpose, so there were 437 several aspects that would have improved the investigation of our research questions but 438 were not feasible. Firstly, collecting coping style data would have enabled a full replication 439 of the mediation model of Grass et al. (2018). Secondly, longitudinal data would have 440 facilitated more definitive conclusions about causal relations, as well as about 441 inter-individual differences in the perception of demands and resources as the pandemic progresses. Furthermore, the latent variables for the demand-resource-ratios were item groups chosen from the work satisfaction questionnaire and had not been validated for this use before. However, as two of them showed meaningful relations with self-control, NFC, and two of the three MBI subscales, pursuing this concept further seems promising. Especially because we worked with pre-existing data, we preregistered all analyses and clearly differentiated between confirmatory and exploratory models in order to make the

results as reliable as possible. Applied to real-life teaching practise, our results suggest that
a healthy work environment should offer ample opportunities to make use of one's abilities,
without creating demands that are too high. As a consequence, experiences and sense of
self-efficacy will increase, which in turn heightens confidence in one's skills to deal with
future demands that are higher, preventing loss of personal efficacy and burnout in the
long term.

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Supplementary Material

758 S1: Items used to assess Covid burden

- 1. How burdened do you currently feel by the measures associated with Covid-19?
- 2. Are you in a Covid-19 risk group?
- 3. Do you have or have you had a Covid-19 infection?
- 4. Are or were family members or other people close to you infected with Covid-19?
- 5. Do you feel more burdened at work?
- 6. Are your worried more?
- 765 7. Do you feel restricted in your current day-to-day life?
- 8. Do you currently have additional responsibilities?
- 9. How much time do you currently spend on leisure activities?
- 10. Do you currently spend more/less time on work-related activities (e.g. preparing lessons, reading literature, attending trainings for digital teaching)?
- 11. Did the current demands within your job change?
- For each response scale, please refer to Excel file with the full list of items and response types on OSF https://osf.io/36ep9/.

- 773 S2: Results when excluding the outlier with very high MBI scores and very low
- NFC scores

Table S.1 Spearman correlations and internal consistencies of the questionnaire scores.

	1	2	3	4	5	6	7	8	9	10	11	12	13
	-			-			•					- -	
1. MBI	0.90(0.91)												
2. MBI EE	0.924***	0.91(0.91)											
3. MBI DP	0.743***	0.528***	0.68(0.66)										
4. MBI RPE	0.663***	0.425***	0.471***	0.79(0.78)									
5. ERQ	-0.051	-0.052	0.052	-0.092	0.73(0.62)								
6. ERQ S	0.053	-0.001	0.167*	0.076	0.593***	0.75(0.79)							
7. ERQ R	-0.094	-0.050	-0.054	-0.191*	0.713***	-0.075	0.84(0.84)						
8. SCS	-0.331***	-0.271***	-0.357***	-0.171*	-0.043	-0.121	0.041	0.85(0.85)					
9. NFC	-0.235**	-0.182*	-0.205**	-0.200**	-0.016	-0.177*	0.151*	0.203**	0.89(0.88)				
10. DTH	0.660***	0.717***	0.337***	0.354***	0.038	0.054	0.002	-0.194**	-0.134	0.73(0.72)			
11. DTL	0.436***	0.347***	0.369***	0.422***	0.015	0.158*	-0.129	-0.178*	-0.149*	0.399***	0.73(0.76)		
12. DRF	-0.537***	-0.447***	-0.400***	-0.523***	-0.013	-0.096	0.090	0.163*	0.228**	-0.410***	-0.554***	0.77(0.75)	
13. COV	0.231**	0.316***	0.072	0.004	-0.022	0.019	-0.059	-0.028	0.139	0.441***	0.084	-0.120	0.77(0.81)

Note: MBI = Maslach Burnout Inventory, MBI EE = Emotional exhaustion subscale, MBI DP = Depersonalisation subscale, MBI RPE = Reduced personal efficacy subscale, ERQ = Emotion Regulation Questionnaire, ERQ S = Suppression subscale, ERQ R = Reappraisal subscale, SCS = Self-Control Scale, NFC = Need for Cognition, DTH = Demands Too High, DTL = Demands Too Low, DRF = Demand-Resource-Fit, COV = Covid-19 Burden. N = 179. * p < .05. *** p < .01. *** p < .001. Diagonal is Cronbach's Alpha and (in brackets) MacDonald's Omega.

Table S.2
Results of the replication of Grass et al. (2018).

Path	В	SE	z-value	<i>p</i> -value	CI Lower	CI Upper	β
Direct Effects							
NFC on Self Control	0.132	0.047	2.804	0.005	0.042	0.226	0.217
NFC on Reappraisal	0.052	0.039	1.353	0.176	-0.021	0.127	0.112
NFC on Suppression	-0.068	0.027	-2.519	0.012	-0.121	-0.016	-0.188
Self Control on RPE	-0.055	0.029	-1.910	0.056	-0.112	0.001	-0.137
Reappraisal on RPE	-0.093	0.034	-2.707	0.007	-0.156	-0.020	-0.177
Suppression on RPE	0.011	0.051	0.209	0.834	-0.089	0.111	0.016
NFC on RPE	-0.039	0.020	-1.994	0.046	-0.076	0.000	-0.160
Indirect Effects							
NFC on RPE via Self Control	-0.007	0.005	-1.403	0.161	-0.019	0.000	-0.030
NFC on RPE via Reappraisal	-0.005	0.004	-1.217	0.224	-0.014	0.002	-0.020
NFC on RPE via Suppression	-0.001	0.004	-0.191	0.848	-0.009	0.006	-0.003
Total Effect							
Total Effect	-0.052	0.021	-2.518	0.012	-0.090	-0.010	-0.212

Note: B = unstandardized regression coefficient, beta = standardized regression coefficient, CI = confidence interval, NFC = Need for Cognition, RPE = reduced personal efficacy subscale of the Maslach Burnout Inventory, SE = standard error, N = 179.

Table S.3 Results of the demand-resource-ratio model.

Path	В	SE	z-value	<i>p</i> -value	CI Lower	CI Upper	β
Direct Effects							
NFC on DTH	-0.035	0.020	-1.789	0.074	-0.074	0.003	-0.183
NFC on DTL	-0.020	0.015	-1.287	0.198	-0.050	0.010	-0.152
NFC on DRF	0.060	0.020	2.942	0.003	0.020	0.100	0.318
NFC on MBI	0.024	0.151	0.161	0.872	-0.272	0.320	0.010
DTH on MBI	11.464	2.117	5.416	0.000	7.316	15.612	0.912
DTL on MBI	1.951	1.565	1.247	0.212	-1.115	5.018	0.106
DRF on MBI	-3.565	1.020	-3.495	0.000	-5.564	-1.566	-0.280
Indirect Effects							
NFC on MBI via DTH	-0.403	0.230	-1.754	0.079	-0.853	0.047	-0.167
NFC on MBI via DTL	-0.039	0.034	-1.134	0.257	-0.106	0.028	-0.016
NFC on MBI via DRF	-0.215	0.104	-2.070	0.038	-0.418	-0.011	-0.089
Total Effect							
Total Effect	-0.632	0.253	-2.498	0.012	-1.128	-0.136	-0.262

Note: B = unstandardized regression coefficient, beta = standardized regression coefficient, CI = confidence interval, DTH = Demands Too High, DTL = Demands Too Low, DRF = Demand Resource Fit, MBI = Maslach Burnout Inventory, NFC = Need for Cognition, SE = standard error, N = 179.

S3: Replication of Grass et al. (2018) when including years spent teaching

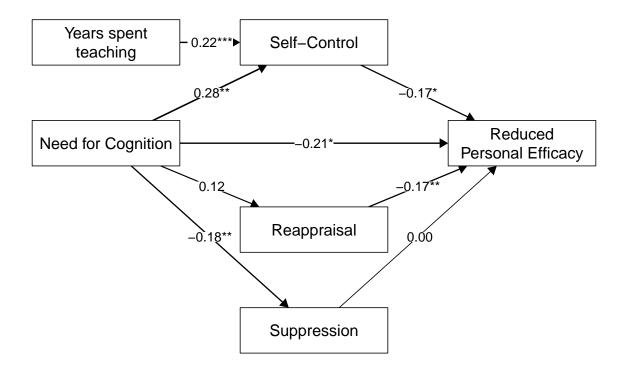


Figure S.1. Standardized regression coefficients in the replication of Grass et al. (2018) when including years spent teaching. * p < .05, ** p < .01.

Table S.4 Results of the replication of Grass et al. (2018) when including years spent teaching.

Path	В	SE	z-value	<i>p</i> -value	CI Lower	CI Upper	β
Direct Effects							
NFC on Self Control	0.168	0.052	3.258	0.001	0.064	0.267	0.280
Years spent teaching on	0.145	0.044	3.299	0.001	0.054	0.230	0.223
Self Control							
NFC on Reappraisal	0.055	0.036	1.519	0.129	-0.016	0.125	0.123
NFC on Suppression	-0.063	0.024	-2.602	0.009	-0.109	-0.014	-0.182
Self Control on RPE	-0.069	0.030	-2.271	0.023	-0.127	-0.010	-0.169
Reappraisal on RPE	-0.094	0.036	-2.618	0.009	-0.164	-0.022	-0.173
Suppression on RPE	0.002	0.049	0.044	0.965	-0.093	0.101	0.003
NFC on RPE	-0.051	0.020	-2.491	0.013	-0.089	-0.010	-0.208
Indirect Effects							
NFC and years spent	-0.021	0.011	-1.965	0.049	-0.045	-0.002	-0.085
teaching on RPE via Self							
Control							
NFC on RPE via	-0.005	0.004	-1.325	0.185	-0.014	0.002	-0.021
Reappraisal							
NFC on RPE via	0.000	0.003	-0.041	0.968	-0.008	0.006	-0.001
Suppression							
Total Effect							
Total Effect	-0.078	0.025	-3.164	0.002	-0.124	-0.027	-0.315

Note: B = unstandardized regression coefficient, beta = standardized regression coefficient, CI = confidence interval, NFC = Need for Cognition, RPE = reduced personal efficacy subscale of the Maslach Burnout Inventory, SE = standard error.

S4: Demand-resource-ratio model with the MBI subscale reduced personal efficacy

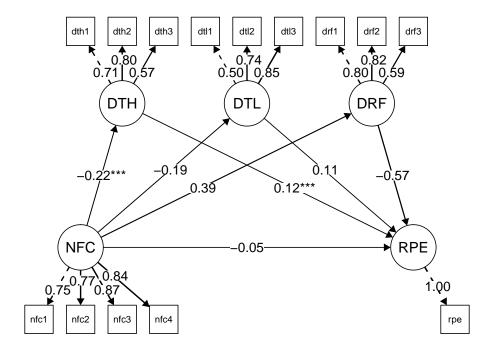


Figure S.2. Standardized path coefficients of the demand-resource-ratio model with the MBI subscale reduced personal efficacy. * p < .05, ** p < .01, *** p < .001. NFC = Need for Cognition, DTH = demands too high, DTL = demands too low, DRF = demand resource fit, nfc1-4 = item parcels, dth/dtl/drf1-3 = item indicators, RPE = reduced personal efficacy.

778 S5: Exploratory model with all relevant variables

Table S.5
Results of the exploratory model with Covid burden.

Path	В	SE	z-value	<i>p</i> -value	CI Lower	CI Upper	β
Direct Effects							
Years on COVB	0.055	0.024	2.327	0.020	0.009	0.102	0.168
Years on SCS	0.137	0.045	3.037	0.002	0.049	0.226	0.212
COVB on DTH	0.061	0.014	4.352	0.000	0.034	0.089	0.449
SCS on DTH	-0.015	0.005	-3.069	0.002	-0.025	-0.005	-0.217
NFC on DTH	-0.038	0.014	-2.646	0.008	-0.065	-0.010	-0.210
SCS on DRF	0.015	0.006	2.540	0.011	0.003	0.026	0.223
NFC on DRF	0.057	0.018	3.162	0.002	0.022	0.093	0.336
DTH on EE	14.985	2.111	7.098	0.000	10.847	19.124	1.004
COVB on EE	-0.294	0.136	-2.161	0.031	-0.560	-0.027	-0.144
DRF on RPE	-4.686	0.634	-7.387	0.000	-5.930	-3.443	-0.760
Indirect Effects							
NFC and Years on RPE via SCS and DRF	-0.279	0.084	-3.319	0.001	-0.443	-0.114	-0.291
NFC and Years on EE via SCS, COVB, and DTH	-0.543	0.206	-2.633	0.008	-0.947	-0.139	-0.181
Total Effect							
Total Effect	-0.821	0.256	-3.212	0.001	-1.322	-0.320	-0.472

Note: B = unstandardized regression coefficient, beta = standardized regression coefficient, CI = confidence interval, COVB = Covid Burden, DTH = Demands Too High, DRF = Demand Resource Fit, MBI = Maslach Burnout Inventory, NFC = Need for Cognition, SCS = Self Control Scale, SE = standard error, Years = Years spent teaching.