

Resilience and Affective Well-Being in Digital Environments

This study examines the relationship between resilience as a multidimensional resource for adaptive coping and emotional regulation and affective well-being in online work and learning environments. The sample included 375 participants engaged in online work and online learning activities (age range 18–48 years; $M = 21.42$, $SD = 3.93$). Resilience was assessed using the Connor–Davidson Resilience Scale (Connor & Davidson, 2003), while affective well-being was measured through a linguistically adapted Bulgarian version of the IWP Multi-Affect Indicator (Warr, 2016). Factor analysis supported the multidimensional structure of both instruments. In the Bulgarian adaptation, anxiety and depressive affect converged into a broader Emotional Distress dimension alongside Enthusiastic Affect and Calm Emotional States. Regression analyses demonstrated that resilience dimensions were differentially associated with emotional functioning. Adaptive coping capacity emerged as the strongest predictor across emotional outcomes, highlighting the role of emotional flexibility and self-regulation in online adaptation. Goal-directedness was positively associated with enthusiastic affect, whereas social support demonstrated protective associations with emotional distress. External reliance was positively related to emotional distress and negatively associated with calm emotional states. The findings support multidimensional approaches to affective well-being and underline the practical relevance of resilience-focused interventions in online educational and occupational settings.

Keywords: *resilience; affective well-being; digitally mediated environments; online work and learning; emotional distress.*

Introduction

The transition toward online work and online learning has transformed not only the organization of everyday activities, but also the emotional conditions under which psychological functioning unfolds. Online environments frequently require sustained concentration, continuous cognitive engagement, and increased self-regulation, while simultaneously limiting many forms of direct interpersonal interaction and emotional feedback. As occupational and academic responsibilities increasingly overlap with personal space and everyday routines, individuals are often expected to maintain emotional involvement and productivity under conditions of prolonged mental activation. These developments have intensified scholarly interest in the psychological consequences of online environments and in the factors that may support adaptive emotional functioning within them (Akese, 2024; Nuñez et al., 2022).

Recent empirical evidence suggests that online environments may contribute to emotional fatigue, motivational decline, and difficulties maintaining psychological balance over time. Akese (2024) observed that remote learning settings were associated with heightened anxiety, emotional stress, and reduced social connectedness, particularly among younger individuals who experienced challenges separating academic demands from personal life. Similarly, Nuñez et al. (2022)

1 reported that students engaged in prolonged remote learning demonstrated
2 fluctuations in emotional well-being, attentional difficulties, and lower levels of
3 academic engagement. Together, these findings indicate that emotional experiences
4 in online contexts may gradually become more cumulative, internally sustained, and
5 psychologically difficult to separate, increasing the importance of adaptive personal
6 resources in maintaining emotional stability.

7 Within this framework, resilience has increasingly been conceptualized as a
8 dynamic process of psychological adaptation rather than as a fixed personality
9 characteristic. Connor and Davidson (2003) describe resilience as a multidimensional
10 capacity that enables individuals to preserve or restore effective functioning under
11 conditions of adversity, uncertainty, or prolonged stress exposure. Their
12 conceptualization emphasizes several interconnected psychological resources,
13 including adaptive coping, emotional regulation, goal orientation, perceived control,
14 and social connectedness. Importantly, resilience is viewed as modifiable and context-
15 sensitive, suggesting that adaptive functioning depends not only on stable individual
16 characteristics but also on the interaction between personal resources and
17 environmental demands. This perspective appears particularly relevant in online
18 settings, where emotional adaptation often occurs under conditions of sustained
19 cognitive and emotional activation.

20 Several contemporary studies provide empirical support for the protective role of
21 resilience in online educational and occupational environments. Ge (2025)
22 demonstrated that resilience among university students engaged in online learning
23 was positively associated with emotional engagement, persistence, and self-regulated
24 learning strategies, while simultaneously reducing emotional exhaustion during
25 prolonged digital participation. Similarly, Shi et al. (2025) found that psychological
26 resilience among online teachers was strongly related to self-efficacy, digital
27 competence, and adaptive professional functioning under conditions of continuous
28 technological demands. These findings suggest that resilience in online settings
29 appears to involve considerably more than resistance to stress alone and includes the
30 capacity to sustain emotional balance, motivation, and psychological engagement
31 despite ongoing cognitive strain.

32 The emotional dynamics of online environments have also contributed to
33 increasing interest in multidimensional models of affective well-being. Warr's
34 circumplex model proposes that emotions differ simultaneously in valence and
35 activation level, resulting in qualitatively distinct emotional experiences such as
36 anxiety, enthusiasm, calmness, and depressive affect (Warr, 2016). This framework
37 challenges traditional approaches that conceptualize well-being primarily through
38 broad positive versus negative affect distinctions. In online settings, individuals may
39 experience emotional exhaustion together with cognitive activation, or psychological
40 tension alongside productivity and engagement. Such emotional patterns are difficult
41 to capture through generalized indicators of stress or well-being alone and require
42 more differentiated conceptual approaches capable of reflecting the complexity of
43 emotional adaptation in online environments.

44 Recent evidence increasingly supports the importance of distinguishing between
45 different forms of emotional experience when examining occupational and
46 educational functioning. Madrid et al. (2014), for example, demonstrated that

1 activated positive affect was specifically associated with proactive work behavior and
2 psychological engagement, whereas other affective states showed substantially
3 weaker or qualitatively different relationships with occupational outcomes. Likewise,
4 Zhang (2025) emphasized that emotional well-being in online educational contexts is
5 strongly associated with emotional self-confidence and the ability to sustain
6 psychological involvement during prolonged digital interaction. These findings
7 suggest that activated positive emotions, calm emotional states, and emotional distress
8 may reflect distinct adaptive mechanisms rather than opposite poles of a single
9 emotional continuum.

10 Another important direction in recent literature concerns the role of interpersonal
11 connectedness and emotional support in maintaining psychological well-being within
12 online environments. Abbasi (2025) found that emotional resilience and online social
13 connectedness significantly contributed to digital well-being by reducing emotional
14 disengagement and feelings of isolation. Similarly, Bagdžiūnienė et al. (2023)
15 reported that emotional resilience resources were associated with greater
16 psychological well-being, emotional flexibility, and adaptive functioning among
17 professionals exposed to sustained occupational demands. Their findings highlight the
18 importance of perceived support and emotional connectedness as mechanisms that
19 may buffer the effects of chronic psychological strain in online settings. These
20 observations support broader theoretical perspectives suggesting that resilience-
21 related processes are deeply embedded within interpersonal and contextual
22 experiences rather than functioning solely as isolated individual characteristics.

23 Recent studies have also begun to examine whether resilience-related capacities
24 can be strengthened through targeted online interventions and supportive digital
25 practices. Research focusing on online resilience-building programs has reported
26 improvements in adaptive coping, emotional regulation, and reductions in anxiety and
27 depressive symptoms following structured interventions. Such findings are important
28 because they position resilience as a modifiable psychological process rather than a
29 static personal disposition. In parallel, Medina-Garrido et al. (2023) demonstrated that
30 emotional well-being in flexible digital work arrangements was closely associated
31 with the ability to maintain balance between occupational and personal demands.
32 Together, these findings suggest that emotional adaptation in online contexts may
33 depend not only on individual coping capacities, but also on the broader
34 organizational and interpersonal conditions under which online functioning takes
35 place.

36 Despite substantial recent advances, several conceptual and methodological gaps
37 remain insufficiently explored. Much of the existing literature continues to rely on
38 generalized indicators of stress, burnout, or global well-being, often overlooking both
39 the multidimensional structure of resilience and the differentiated nature of emotional
40 experience. In addition, relatively limited attention has been devoted to culturally
41 sensitive adaptations of resilience and affective well-being measures within online
42 contexts. The present research addresses these limitations by integrating
43 multidimensional models of resilience and emotional functioning within a Bulgarian
44 sample engaged in online work and online learning. By examining distinct resilience-
45 related resources alongside differentiated emotional states, the study aims to

1 contribute to a more nuanced and context-sensitive understanding of psychological
2 adaptation in contemporary digital environments.

3 4 5 **Method**

6
7 Data were collected through an online survey administered via the Google Forms
8 platform. Participants were informed that the study aimed to examine psychological
9 resilience and emotional functioning in digitally mediated work and learning
10 environments. They were also informed that participation was voluntary, anonymous,
11 and that all collected data would be used solely for scientific purposes. Prior to each
12 questionnaire and scale, participants received brief information regarding the purpose
13 of the instrument as well as instructions for completion. At the end of the survey, a
14 demographic section was included assessing participants' socio-demographic
15 characteristics. The collected data were processed and statistically analyzed using
16 IBM SPSS (v. 29).

17 18 *Participants*

19
20 The sample included 375 Bulgarian participants engaged in online work and
21 online learning activities. Participants ranged in age from 18 to 48 years ($M = 21.43$,
22 $SD = 3.93$). The gender distribution consisted of 140 males (37.3%) and 235 females
23 (62.7%). Data were collected through an anonymous online psychological survey.
24 Participation was voluntary, and all respondents were informed about the aims of the
25 study, the confidentiality of the collected data, and the scientific purposes of the
26 research prior to participation. The online format allowed the inclusion of individuals
27 actively involved in digitally mediated educational and occupational environments,
28 reflecting the everyday psychological conditions associated with prolonged online
29 engagement.

30 31 *Instruments*

32
33 For the purposes of the present study, a psychological assessment battery
34 including the following questionnaires and scales was administered.

35 The first instrument was the Connor–Davidson Resilience Scale (CD-RISC;
36 Connor & Davidson, 2003), designed to assess psychological resilience as a
37 multidimensional adaptive capacity associated with emotional regulation, persistence,
38 coping flexibility, and functioning under conditions of stress and adversity. The
39 questionnaire is self-report and consists of 25 items.

40 In the original version, participants evaluate each statement using a five-point
41 Likert scale ranging from “0 – not true at all” to “4 – true nearly all the time.” In the
42 present study, a modified response format was applied using a seven-point agreement
43 scale ranging from “1 – strongly disagree” to “7 – strongly agree.” The original
44 instrument includes five resilience-related dimensions: “Personal Competence, High
45 Standards, and Tenacity,” “Trust in One’s Instincts, Tolerance of Negative Affect,

1 and Strengthening Effects of Stress,” “Positive Acceptance of Change and Secure
2 Relationships,” “Control,” and “Spiritual Influences.”

3 The second instrument was the IWP Multi-Affect Indicator developed by Warr
4 (2016), designed to assess affective well-being in occupational and educational
5 contexts through emotions differing in valence and activation level. The original
6 version includes 16 affective descriptors representing four emotional domains:
7 Anxiety, Depression, Enthusiasm, and Comfort. During the linguistic adaptation of
8 the questionnaire, two descriptors from the original version (“dejected” and “laid-
9 back”) were excluded due to semantic overlap with other retained emotional states
10 and difficulties identifying conceptually precise Bulgarian equivalents. Several
11 descriptors were additionally refined in order to improve semantic clarity and
12 contextual comprehensibility within the Bulgarian cultural context. As a result, the
13 final adapted version consisted of 15 affective descriptors. In the original version, the
14 emotional states are evaluated using a six-point frequency scale. In the present study,
15 a modified five-point response scale ranging from “1 – never” to “5 – all the time”
16 was applied in order to facilitate clearer differentiation and more consistent
17 responding within the online research context.

18 19 Factor Structure of the Connor–Davidson Resilience Scale

20 The factorial structure of the Bulgarian adaptation of the Connor–Davidson
21 Resilience Scale (CD-RISC; Connor & Davidson, 2003) was examined through
22 exploratory factor analysis using Principal Component Analysis with Varimax
23 rotation. The analysis aimed to investigate the latent organization of resilience-related
24 characteristics within the Bulgarian sample and to evaluate the structural adequacy of
25 the adapted instrument. Sampling adequacy indicators demonstrated excellent
26 suitability of the data for factor analysis (Kaiser–Meyer–Olkin measure = .935;
27 Bartlett’s Test of Sphericity: $\chi^2(300) = 4153.95$, $p < .001$). The extracted factors
28 explained 59.75% of the total variance, indicating satisfactory structural coherence
29 and psychometric adequacy of the scale.

30 The obtained factorial solution revealed a multidimensional structure reflecting
31 several interconnected psychological resources involved in adaptation, emotional
32 regulation, and coping under stress. Although the original factorial structure proposed
33 by Connor and Davidson was only partially replicated, the extracted dimensions
34 demonstrated clear conceptual interpretability and satisfactory internal consistency.
35 Rather than reproducing the original factor configuration, the Bulgarian adaptation
36 yielded psychologically meaningful clusters of items corresponding to distinct aspects
37 of resilience functioning in digitally mediated contexts. Consequently, revised factor
38 labels were introduced in order to more accurately reflect the semantic content and
39 psychological meaning of the retained item groupings (see Table 1).

40
41

1 **Table 1.** *Rotated Factor Matrix of the Bulgarian Adaptation of the Connor–Davidson*
 2 *Resilience Scale (CD-RISC; Connor & Davidson, 2003).*

Item	Adaptive Coping Capacity	Goal-Directedness	Personal Agency	External Reliance	Social Support
Able to adapt to change	.781				
Can deal with whatever comes	.758				
Coping with stress strengthens me	.742				
Tend to bounce back after hardship	.731				
Under pressure, stay focused	.706				
Think clearly under pressure	.691				
Can handle unpleasant feelings	.648				
Not easily discouraged by failure	.617				
Achieve goals despite obstacles		.817			
Strong sense of purpose		.801			
Work to attain goals		.783			
Give best effort no matter what		.751			
Like challenges		.710			
Past success gives confidence		.694			
Things happen for a reason		.602			
Pride in achievements		.588			
Think of self as strong person			.774		
Feel in control of life			.756		
Make difficult decisions			.691		
Prefer taking the lead			.643		
Sometimes fate or God can help				.779	
Have to act on a hunch				.713	
Trust instincts under stress				.645	
Close and secure relationships					.812
Know where to turn for help					.748
Eigenvalue (after rotation)	5.214	3.882	2.311	1.911	1.620
Explained Variance (%)	20.86	15.52	9.24	7.64	6.48
Cumulative Variance (%)	20.86	36.39	45.63	53.27	59.75
Cronbach's α	.846	.876	.768	.618	.723

3 *Note.* Extraction Method: Principal Component Analysis. Rotation Method: Varimax with Kaiser
 4 Normalization. Only factor loadings above .40 are presented.

5
 6 The first factor, Adaptive Coping Capacity (8 items; $\alpha = .846$), included
 7 statements associated with emotional flexibility, stress tolerance, behavioral
 8 adaptation, and the ability to recover under pressure. The factor reflects the capacity
 9 to maintain psychological functioning despite emotionally demanding or uncertain
 10 situations. Goal-Directedness (8 items; $\alpha = .876$) represented persistence,
 11 motivational orientation, and the ability to sustain focus and commitment to personal
 12 goals despite obstacles and difficulties. Personal Agency (4 items; $\alpha = .768$) captured

1 perceptions of self-efficacy, decisiveness, confidence in personal capabilities, and
2 perceived influence over challenging situations.

3 The factor labeled External Reliance (3 items; $\alpha = .618$) included items
4 associated with reliance on intuition, external sources of meaning, faith, or broader
5 beliefs supporting psychological stability under stress. Unlike the original “Spiritual
6 Influences” factor proposed by Connor and Davidson, the obtained structure
7 suggested a broader orientation toward external forms of guidance, meaning-making,
8 and psychological support rather than exclusively spiritual beliefs. The fifth factor,
9 Social Support (2 items; $\alpha = .723$), reflected perceived emotional connectedness and
10 the availability of supportive interpersonal relationships during stressful experiences.
11 Although brief in structure, the factor demonstrated satisfactory internal consistency
12 and conceptual clarity.

13 The revised factor labels were selected to improve conceptual clarity and
14 interpretability while maintaining theoretical consistency with contemporary
15 resilience research. This renaming approach follows common psychometric practices
16 in cross-cultural adaptation studies, where factor nomenclature is determined not only
17 by theoretical expectations, but also by the empirical composition and psychological
18 coherence of the extracted dimensions. Overall, the obtained structure supports the
19 multidimensional conceptualization of resilience and suggests that resilience-related
20 psychological functioning in digitally mediated environments involves several
21 differentiated but interconnected adaptive resources.

22 Factor Structure of the IWP Multi-Affect Indicator

23 The factorial structure of the Bulgarian adaptation of the short version of the IWP
24 Multi-Affect Indicator (Warr, 2016) was also examined through exploratory factor
25 analysis using Principal Component Analysis with Varimax rotation. The instrument
26 assesses affective well-being through emotional states differing in emotional valence
27 and activation level. While Warr’s original circumplex model conceptualizes
28 emotional experience through four affective quadrants — anxiety, depression,
29 enthusiasm, and comfort — cross-cultural adaptations may demonstrate alternative
30 empirical structures depending on linguistic and contextual characteristics.

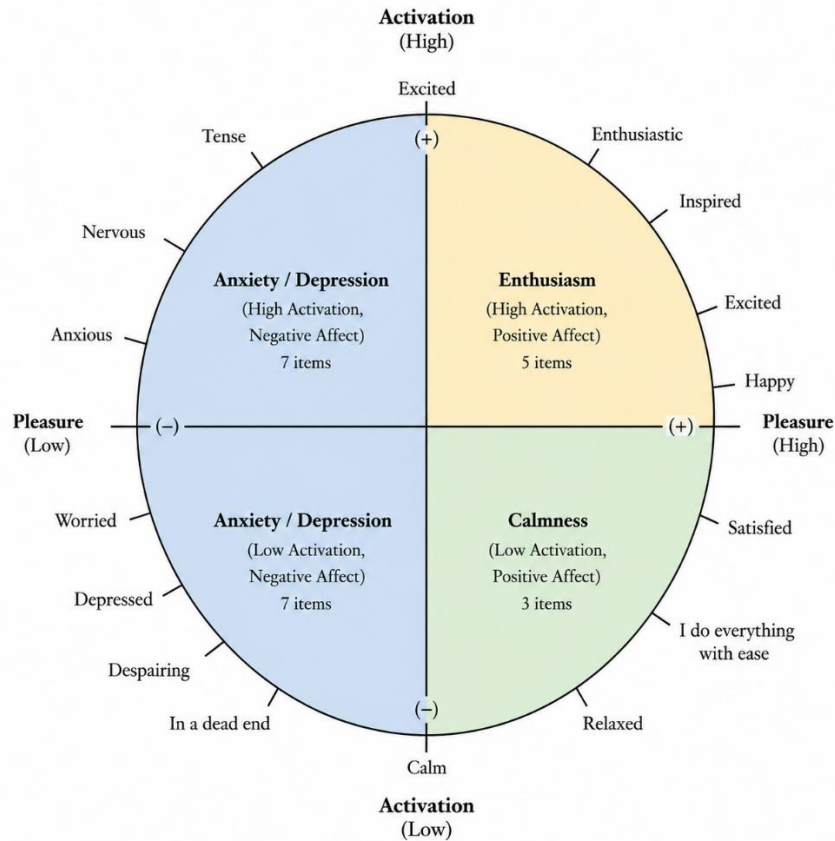
31 Figure 1 presents the circumplex structure of the Bulgarian adaptation of the IWP
32 Multi-Affect Indicator based on the dimensions of emotional valence (pleasure) and
33 activation. The figure illustrates the organization of emotional states into three
34 empirically derived affective domains: Emotional Distress, Enthusiastic Emotions,
35 and Calm Emotions. In the Bulgarian adaptation, anxiety- and depression-related
36 emotions converged into a unified Emotional Distress dimension, while positive
37 affect remained differentiated into high-activation enthusiastic emotions and low-
38 activation calm emotional states.

39 A total of 15 affective descriptors were included in the analysis: enthusiastic,
40 nervous, calm, depressed, happy, anxious, satisfied, despairing, inspired, tense,
41 relaxed, excited, worried, “I do everything with ease,” and “in a dead end.” Figure 1
42 presents the circumplex structure of the Bulgarian adaptation of the IWP Multi-Affect
43 Indicator based on the dimensions of emotional valence (pleasure) and activation. The
44 figure illustrates the organization of emotional states into three empirically derived
45 affective domains: Emotional Distress, Enthusiastic Emotions, and Calm Emotions.
46 In the Bulgarian adaptation, anxiety- and depression-related emotions converged into

1 a unified Emotional Distress dimension, while positive affect remained differentiated
 2 into high-activation enthusiastic emotions and low-activation calm emotional states.

3

4 **Figure 1.** *Circumplex Structure of the Bulgarian Adaptation of the IWP Multi-Affect*
 5 *Indicator*



6

7 Sampling adequacy indicators demonstrated excellent suitability of the data for
 8 factor analysis (Kaiser–Meyer–Olkin measure = .916; Bartlett’s Test of Sphericity:
 9 $\chi^2(105) = 3292.623, p < .001$). Three components with eigenvalues greater than 1
 10 were extracted, explaining 67.41% of the total variance. The rotated solution revealed
 11 a psychologically meaningful and interpretable structure (see Table 2).

12

13 **Table 2.** *Rotated Factor Matrix of the Bulgarian Adaptation of the IWP Multi-Affect*
 14 *Indicator (Warr, 2016).*

Item	Emotional Distress	Enthusiastic Emotions	Calm Emotional States
In a dead end	.816		
Despairing	.774		
Depressed	.767		
Anxious	.704		
Worried	.680		
Nervous	.663		
Tense	.661		

Excited		.791	
Enthusiastic		.779	
Satisfied		.740	
Inspired		.732	
Happy		.700	
Relaxed			.774
Calm			.702
I do everything with ease			.572
Eigenvalue (after rotation)	4.091	3.379	2.642
Explained Variance (%)	27.27	22.53	17.61
Cumulative Variance (%)	27.27	49.80	67.41
Cronbach's α	.916	.856	.722

1 *Note.* Extraction Method: Principal Component Analysis. Rotation Method: Varimax with Kaiser
 2 Normalization. Only factor loadings above .40 are presented.

3

4 The first factor, Emotional Distress (7 items), integrated descriptors traditionally
 5 associated with both anxiety and depressive affect, including “in a dead end,”
 6 “despairing,” “depressed,” “anxious,” “worried,” “nervous,” and “tense.” Rather than
 7 forming two separate emotional domains, high-activation anxiety and low-activation
 8 depressive affect converged into a broader dimension of psychological distress. This
 9 merged structure suggests a substantial semantic and experiential overlap between
 10 tension, emotional exhaustion, discouragement, and depressive affect within the
 11 Bulgarian adaptation and the examined online context.

12 The second factor, Enthusiastic Emotions (5 items), included descriptors such as
 13 excited, enthusiastic, inspired, happy, and satisfied. These emotional states reflect
 14 activated positive affect characterized by energy, motivational involvement,
 15 optimism, and psychological engagement. The third factor, Calm Emotions (3 items),
 16 included calm, relaxed, and “I do everything with ease,” reflecting low-activation
 17 positive emotional states associated with emotional balance, relaxation, and subjective
 18 psychological comfort. Although relatively brief, both subscales demonstrated
 19 satisfactory internal consistency and conceptual coherence.

20 Overall, the obtained structure partially supports Warr's theoretical model while
 21 also demonstrating cultural and contextual specificity in the organization of affective
 22 experiences. Negative emotional states appeared less differentiated into distinct
 23 anxiety and depressive dimensions and were instead experienced as a broader
 24 emotional distress construct. In contrast, positive affect remained differentiated into
 25 activated enthusiastic emotions and low-activation calm emotional states. The results
 26 support the construct validity and applicability of the Bulgarian short version of the
 27 IWP Multi-Affect Indicator for research examining emotional functioning,
 28 occupational well-being, stress, and resilience in digitally mediated educational and
 29 occupational environments.

30

1 **Results**

2

3 *Descriptive Statistics for the Study Variables*

4

5 Descriptive statistics were calculated for all resilience and affective well-being
6 variables in order to examine the distributional characteristics of the data and the
7 overall levels of the studied psychological constructs. As presented in Table 3, the
8 variables demonstrated acceptable distributional properties, with skewness and
9 kurtosis values remaining within recommended ranges for parametric analyses.
10 Overall, participants reported relatively higher levels of resilience-related resources
11 and moderate levels of emotional well-being in digitally mediated work and learning
12 environments.

13

14 **Table 3.** *Descriptive Statistics for the Study Variables*

Variable	M	SD	Skewness	Kurtosis	Min	Max
Adaptive Coping Capacity	3.81	.76	-.50	.01	1.13	5.00
Goal-Directedness	3.98	.75	-.74	.45	1.25	5.00
Personal Agency	3.79	.86	-.42	-.53	1.25	5.00
External Reliance	3.86	.92	-.65	-.02	1.00	5.00
Social Support	3.85	.97	-.79	.07	1.00	5.00
Emotional Distress	2.87	.99	.16	-.70	1.00	5.00
Enthusiastic Emotions	3.36	.83	-.33	.06	1.00	5.00
Calm Emotional States	3.12	.86	.03	-.46	1.00	5.00

15 *Note.* *M* = Mean; *SD* = Standard Deviation.

16

17 *Correlations Between Resilience Dimensions and Affective Well-Being Variables*

18

19 Pearson correlation analysis was conducted to examine the relationships between
20 resilience dimensions and affective well-being variables in online work and learning
21 environments. The obtained results (see Table 4) demonstrated a coherent pattern of
22 associations supporting the multidimensional relationship between resilience and
23 emotional functioning.

24

25

1 **Table 4.** *Correlations Between Resilience Dimensions and Affective Well-Being*
 2 *Variables*

Variable	Emotional Distress	Enthusiastic Emotions	Calm Emotional States
Adaptive Coping Capacity	-.533**	.523**	.510**
Goal-Directedness	-.442**	.543**	.375**
Personal Agency	-.341**	.429**	.301**
External Reliance	-.069	.181**	.072
Social Support	-.415**	.436**	.277**

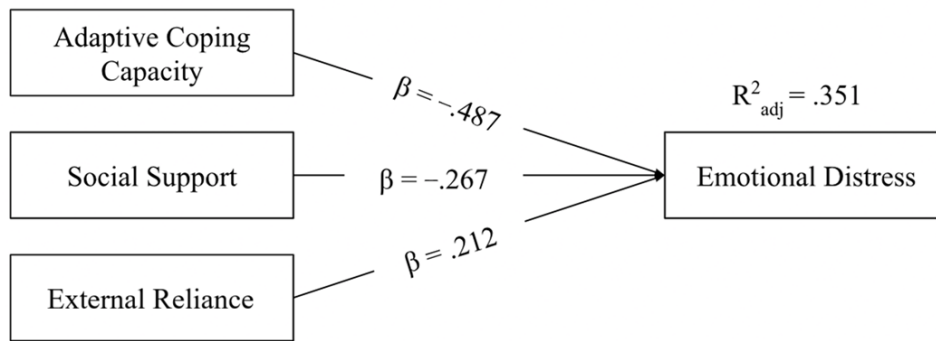
3 *Note.* Pearson correlation coefficients are presented. ** $p < .01$ (two-tailed).
 4

5 Higher levels of adaptive coping, goal-directedness, personal agency, and social
 6 support were consistently associated with lower emotional distress and more positive
 7 emotional experiences. In particular, adaptive coping capacity and goal-directedness
 8 demonstrated the strongest associations with enthusiastic and calm emotional states,
 9 suggesting the importance of emotional flexibility, persistence, and self-regulatory
 10 functioning for positive emotional adaptation in digitally mediated contexts.

11 Social support was also positively related to positive emotional experiences and
 12 negatively associated with emotional distress, highlighting the protective role of
 13 perceived interpersonal connectedness even within predominantly online
 14 environments. In contrast, external reliance demonstrated comparatively weaker and
 15 less consistent associations with emotional functioning, suggesting that reliance on
 16 external sources of meaning or regulation may represent a more context-dependent
 17 resilience-related resource. Overall, the correlation patterns support the assumption
 18 that resilience-related psychological resources are differentially connected to both
 19 positive and negative dimensions of affective well-being.
 20

21 *Resilience Components as Predictors of Emotional Distress*

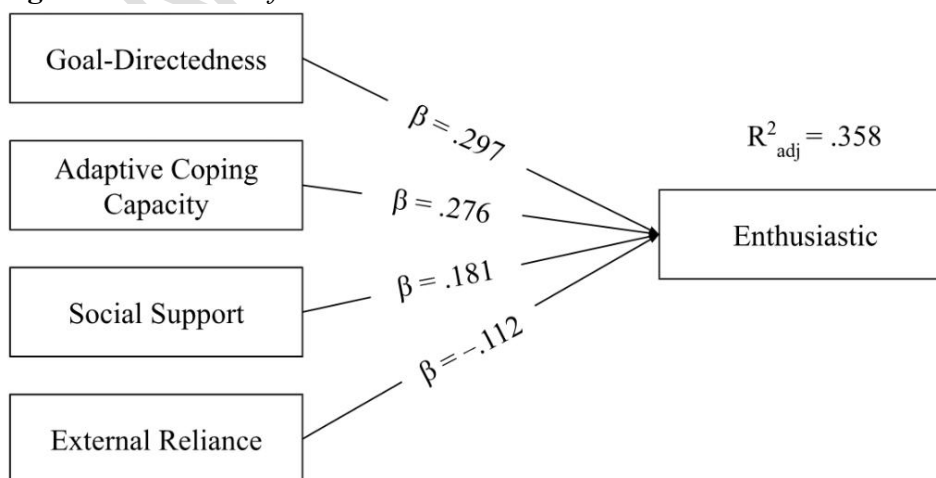
22
 23 A stepwise multiple regression analysis was conducted to examine the predictive
 24 role of resilience dimensions for Emotional Distress in digitally mediated work and
 25 learning environments. The final regression model was statistically significant, $F(3,$
 26 $371) = 68.35, p < .001$, explaining 35.1% of the variance in Emotional Distress (adjusted
 27 $R^2 = .351$). Adaptive Coping Capacity emerged as the strongest negative predictor ($\beta =$
 28 $-.487, p < .001$), indicating that greater emotional flexibility and adaptive self-
 29 regulation were associated with lower levels of emotional distress. Social Support also
 30 demonstrated a significant negative association with Emotional Distress ($\beta = -.267, p$
 31 $< .001$), suggesting a protective role of perceived interpersonal connectedness. In
 32 contrast, External Reliance positively predicted Emotional Distress ($\beta = .212, p < .001$),
 33 indicating that stronger reliance on external sources of regulation or stability was
 34 associated with higher levels of psychological strain. Overall, the model suggests that
 35 emotional distress in online environments is shaped by both adaptive self-regulatory
 36 capacities and interpersonal psychological resources (see Fig. 2).
 37

1 **Figure 2.** *Predictors of Emotional Distress.*

2
3 *Note.* All β coefficients were statistically significant at $p < .001$.

4
5 *Predictors of Enthusiastic Affect Among Resilience Components*

6
7 A second stepwise multiple regression analysis examined the predictors of
8 Enthusiastic Affect in digitally mediated work and learning environments. The final
9 regression model was statistically significant, $F(4, 370) = 53.03$, $p < .001$, explaining
10 35.8% of the variance in Enthusiastic Affect (adjusted $R^2 = .358$). Goal-Directedness
11 emerged as the strongest positive predictor ($\beta = .297$, $p < .001$), suggesting that
12 motivational persistence and future-oriented engagement are closely associated with
13 activated positive emotional experiences in online contexts. Adaptive Coping
14 Capacity also demonstrated a substantial positive association ($\beta = .276$, $p < .001$),
15 indicating the importance of emotional flexibility for maintaining enthusiastic affect.
16 Social Support positively predicted enthusiastic emotions as well ($\beta = .181$, $p < .001$),
17 supporting the role of perceived connectedness in sustaining positive emotional
18 engagement. By contrast, External Reliance showed a negative association with
19 Enthusiastic Affect ($\beta = -.112$, $p < .001$). Overall, the model suggests that enthusiastic
20 emotional functioning in online environments is associated with a combination of
21 motivational persistence, adaptive coping, and interpersonal psychological resources
22 (see Fig. 3).

23
24 **Figure 3.** *Predictors of Enthusiastic Emotions*

25
26 *Note.* All β coefficients were statistically significant at $p < .001$.

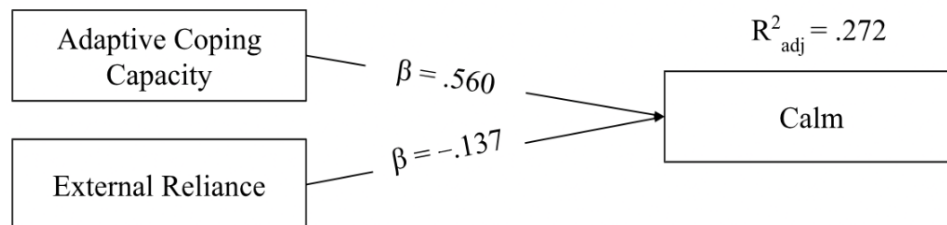
1 *Predictors of Calmness Among Resilience Dimensions*

2

3 The third stepwise multiple regression analysis explored the predictive role of
 4 resilience dimensions for Calm Emotional States in digitally mediated work and
 5 learning environments. The final regression model was statistically significant, $F(2,$
 6 $372) = 70.80, p < .001$, explaining 27.2% of the variance in Calm Emotional States
 7 (adjusted $R^2 = .272$). Adaptive Coping Capacity demonstrated the strongest positive
 8 association with Calmness ($\beta = .560, p < .001$), highlighting the central importance of
 9 emotional self-regulation and psychological flexibility for maintaining low-activation
 10 positive emotional states in online contexts. External Reliance, in contrast,
 11 demonstrated a significant negative association with Calm Emotional States ($\beta =$
 12 $-.137, p < .001$), suggesting that stronger dependence on external sources of stability
 13 may be related to reduced emotional equilibrium and recovery capacity. Overall, the
 14 model indicates that calm emotional functioning in online environments is strongly
 15 associated with adaptive self-regulatory processes rather than with external forms of
 16 psychological orientation (see Fig. 4).

17

18 **Figure 4.** *Predictors of Calm Emotions.*



19

20 *Note.* All β coefficients were statistically significant at $p < .001$.

21

22

23 **Discussion**

24

25 The linguistically adapted version of the IWP Multi-Affect Indicator retained the
 26 theoretical foundations of Warr's (2016) circumplex model while introducing several
 27 modifications aimed at improving semantic clarity and cultural appropriateness within
 28 the Bulgarian context. The original instrument includes 16 affective descriptors
 29 representing four affective domains differing in emotional valence and activation
 30 level: Anxiety, Depression, Enthusiasm, and Comfort. During the adaptation process,
 31 two descriptors from the original version ("dejected" and "laid-back") were excluded
 32 due to substantial semantic overlap with other retained items and difficulties in
 33 identifying conceptually precise and culturally natural equivalents in Bulgarian.
 34 Several additional descriptors were linguistically refined in order to improve
 35 comprehensibility and contextual relevance without altering the conceptual meaning
 36 of the original constructs. As a result, the final adapted version consisted of 15
 37 affective descriptors reflecting both positive and negative emotional states varying in
 38 activation intensity.

39

40 The adaptation process was guided simultaneously by theoretical and
 41 psychometric considerations. Particular attention was given to preserving the
 conceptual integrity of Warr's circumplex framework while ensuring that the

1 emotional descriptors remained psychologically meaningful within the Bulgarian
2 linguistic context. The excluded items appeared less distinguishable from neighboring
3 low-activation emotional states such as depressed, relaxed, and at ease, increasing the
4 likelihood of semantic redundancy during participant interpretation. The retained
5 descriptors demonstrated satisfactory variability and conceptual coherence, producing
6 a stable factorial structure organized around broader dimensions of Emotional
7 Distress, Enthusiastic Affect, and Calmness/Comfort. Such selective adaptation
8 procedures are consistent with established practices in cross-cultural psychometric
9 research, where abbreviated or linguistically refined versions are often developed to
10 improve interpretability and contextual applicability while preserving the theoretical
11 foundations of the original instrument.

12 The obtained results highlight the importance of examining emotional
13 functioning in digitally mediated environments through multidimensional rather than
14 generalized approaches to well-being. Contemporary online educational and
15 occupational contexts frequently involve prolonged cognitive activation, sustained
16 attentional demands, reduced opportunities for psychological recovery, and
17 fragmented interpersonal interaction (Akese, 2024; Shi et al., 2025). Under such
18 conditions, emotional experiences may become more internally regulated,
19 cumulative, and difficult to differentiate. Within this framework, resilience appeared
20 not as a single global protective factor, but as a differentiated system of psychological
21 resources associated with distinct emotional experiences. Adaptive coping capacity
22 demonstrated the most stable associations across emotional outcomes, suggesting that
23 emotional flexibility and the ability to reorganize psychological functioning under
24 changing demands may represent central mechanisms supporting adaptation in
25 digitally mediated environments. Similar observations were reported by Ge (2025),
26 who found that resilience among students engaged in online learning was associated
27 with reduced emotional exhaustion alongside stronger emotional engagement and
28 self-regulated learning behavior.

29 The differentiated relationships between resilience dimensions and affective
30 states additionally support multidimensional conceptualizations of emotional well-
31 being. Warr's (2016) circumplex framework proposes that emotional experiences
32 differ simultaneously in emotional valence and activation level, allowing for a more
33 nuanced understanding of psychological functioning than traditional positive-
34 negative affect distinctions. Within the obtained models, goal-directedness
35 demonstrated the strongest association with enthusiastic affect, suggesting that
36 motivational persistence and future-oriented engagement may play an especially
37 important role in sustaining activated positive emotional states in online
38 environments. These findings correspond with research by Madrid et al. (2014), who
39 demonstrated that activated positive affect was specifically related to proactive
40 engagement and adaptive occupational functioning. More recent findings by Zhang
41 (2025) similarly indicate that emotional engagement and self-confidence contribute
42 significantly to maintaining psychological involvement during prolonged online
43 educational activities.

44 At the same time, adaptive coping capacity demonstrated particularly strong
45 associations with calm emotional states, suggesting that low-activation positive affect
46 may depend less on motivational activation and more on emotional flexibility,

1 recovery processes, and self-regulatory stability. This distinction appears theoretically
2 important because calmness in digitally mediated environments may not simply
3 reflect the absence of emotional strain, but rather the ability to preserve psychological
4 equilibrium under continuous cognitive demands. Recent research supports this
5 interpretation. Bagdžiūnienė et al. (2023), for example, reported that emotional
6 resilience resources were associated with emotional flexibility and psychological well-
7 being among professionals exposed to prolonged occupational stress, while Abbasi
8 (2025) demonstrated that emotional resilience and online social connectedness
9 contributed to greater digital well-being and reduced emotional disengagement.
10 Together, these findings suggest that emotionally adaptive functioning in online
11 contexts may involve the capacity to sustain internal emotional balance under prolonged
12 activation rather than merely minimizing negative affective experiences.

13 A particularly noteworthy result concerns the convergence of anxiety and
14 depressive affect into a unified Emotional Distress factor within the Bulgarian
15 adaptation of the instrument. In Warr's (2016) original theoretical model, anxiety and
16 depression represent qualitatively distinct emotional experiences differing in
17 activation level. In the obtained structure, however, descriptors associated with
18 tension, worry, discouragement, emotional exhaustion, and depressive affect
19 clustered within a broader dimension of psychological distress. Several factors may
20 contribute to this convergence. Previous cross-cultural research has shown that the
21 circumplex differentiation between anxiety and depression is not always consistently
22 reproduced across linguistic contexts, particularly in shorter adapted affective
23 measures. In addition, occupational and chronic stress environments frequently
24 involve the simultaneous presence of anxious activation and emotional depletion,
25 making these experiences subjectively difficult to separate. Within digitally mediated
26 contexts characterized by prolonged attentional demands and blurred boundaries
27 between academic, occupational, and personal functioning, emotional tension may
28 gradually evolve into broader states of exhaustion and discouragement rather than
29 remaining clearly differentiated forms of affect. Similar tendencies have been
30 indirectly reflected in studies demonstrating overlapping experiences of anxiety,
31 emotional fatigue, cognitive exhaustion, and motivational depletion in remote
32 learning and online occupational settings (Nuñez et al., 2022; Akese, 2024).

33 The role of social support within the obtained models further emphasizes the
34 importance of interpersonal connectedness for emotional adaptation in online
35 environments. Social support demonstrated negative associations with emotional
36 distress and positive associations with enthusiastic affect, suggesting that perceived
37 emotional connectedness may function as a stabilizing psychological resource even
38 under conditions of reduced face-to-face interaction. Contemporary evidence
39 increasingly indicates that the subjective experience of social connectedness remains
40 psychologically significant in digitally mediated contexts despite the reduced
41 immediacy of interpersonal communication. Abbasi (2025), for instance, demonstrated
42 that online social connectedness contributed significantly to emotional resilience and
43 digital well-being by reducing emotional disengagement and feelings of isolation.
44 Similarly, Bagdžiūnienė et al. (2023) observed that perceived emotional support was
45 associated with greater emotional flexibility and adaptive functioning under prolonged
46 occupational demands. These findings suggest that resilience-related processes are

1 deeply interconnected with interpersonal emotional experiences rather than
2 functioning solely as intrapersonal coping capacities.

3 The associations involving external reliance require more cautious interpretation.
4 Within the obtained models, external reliance positively predicted emotional distress
5 while simultaneously demonstrating negative associations with calm emotional states.
6 Although the construct may reflect meaning-making tendencies, intuitive orientation,
7 or reliance on external sources of psychological stability, stronger external reliance in
8 digitally mediated contexts may also indicate uncertainty regarding autonomous
9 emotional regulation and reduced confidence in independent coping capacities.
10 Online environments often require individuals to organize tasks independently,
11 regulate attentional focus, and manage emotional demands without immediate
12 external structure or direct interpersonal feedback. Under such conditions, greater
13 external reliance may coexist with heightened emotional instability and reduced
14 psychological certainty. At the same time, the emotional meaning of external reliance
15 may vary across sociocultural and linguistic contexts, making culturally sensitive
16 interpretation particularly important.

17 These observations also carry important practical implications for psychological
18 assessment and intervention in online educational and occupational settings. The
19 obtained models suggest that multidimensional approaches to emotional functioning
20 may provide more informative and clinically meaningful perspectives than
21 generalized indicators of stress or global well-being alone. Distinguishing between
22 emotional distress, enthusiastic affect, and calm emotional states allows for a more
23 differentiated understanding of emotional adaptation under prolonged online
24 participation. In addition, resilience-related dimensions such as adaptive coping
25 capacity, goal-directedness, and social support may represent meaningful targets for
26 preventive and supportive interventions aimed at strengthening emotional functioning
27 in digitally mediated environments. Existing evidence already suggests that resilience-
28 related capacities can be enhanced through structured online interventions associated
29 with improvements in emotional regulation and reductions in anxiety and depressive
30 symptoms (Building Personal Resilience Following an Online Intervention, 2022).

31 Several limitations should nevertheless be acknowledged. The reliance on self-
32 report measures may introduce subjective interpretation biases, while the cross-
33 sectional design limits conclusions regarding causal relationships between resilience
34 dimensions and emotional states. At the same time, the multidimensional structure of
35 both instruments allowed for a considerably more differentiated examination of
36 emotional adaptation than is typically achieved through generalized stress or well-
37 being measures alone. The culturally sensitive adaptation of the affective measure
38 additionally contributes to understanding emotional functioning within Bulgarian
39 digitally mediated contexts, where emotional experiences may demonstrate culturally
40 specific semantic and experiential patterns. Taken together, the obtained results
41 highlight the importance of examining resilience and emotional functioning through
42 context-sensitive and multidimensional psychological frameworks capable of
43 capturing the complexity of emotional adaptation in contemporary digital
44 environments.

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