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Understanding the Association of Fatigue with Other Symptoms of Fibromyalgia: Development of a Cluster Model

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Abstract

Objective—The study's purpose was to develop symptoms cluster model that can describe factors of FMS associated with fatigue severity as reported by the sample. The study will also explore FMS clinical symptom sub-clusters based on varying symptom intensities.

Methods—FMS individuals (n = 120; 82% between 31–60 years of age, 90% women, 59% Caucasian) diagnosed with the 1990 or 2010 American College of Rheumatology diagnostic criteria were enrolled. Participants completed multiple validated self-report questionnaires to measure fatigue, pain, depression, anxiety, pain catastrophizing, daytime sleepiness, cognitive function, and FMS-related polysymptomatic distress. Cluster analysis using SPSS 19.0 and Structural Equation Modeling using AMOS 17.0 were used.

Results—Final Structural Equation Modeling symptoms cluster model showed good fit and revealed that FMS fatigue was associated with widespread pain, symptoms severity, pain intensity, pain interference, cognitive dysfunction, catastrophizing, anxiety, and depression ($\chi^2 = 121.72$, $df = 98$, $p > 0.05$, $\chi^2/df = 1.242$, CFI = 0.982, RMSEA = 0.045). Two distinct clinical symptom sub-clusters emerged; sub-cluster 1 (78% of total subjects) defined by widespread pain, unrefreshed waking, and somatic symptoms and sub-cluster 2 (22% of total subjects) defined by fatigue and cognitive dysfunction with pain being a less severe and less widespread complaint.

Conclusion—Overall, sub-cluster 1 had more intense symptoms than sub-cluster 2. FMS symptoms may be categorized into two clinical sub-clusters. These findings have implications for an illness whose diagnosis and management are symptom-dependent. A longitudinal study capturing the variability in symptom experience of FMS subjects is warranted.

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INTRODUCTION

Fibromyalgia syndrome (FMS) is characterized by a widespread chronic bodily pain, profound fatigue, and sleep disturbance and appears to represent the end of the spectrum of polysymptomatic distress (1). Based on the 2010 diagnostic criteria, FMS affects 6% of the United State population (2). In some patients with FMS, fatigue interferes with the performance of daily activities, as much or more than bodily pain. Because most studies have primarily investigated the mechanisms and treatment of FMS-related pain, less is known about other FMS symptoms including fatigue (3–6). Patients with FMS are often unemployed and have high medical utilization rates related to their fatigue symptoms (7).

Fatigue is defined as a subjective sense of persistent tiredness that interferes with the performance of daily life activities and is not relieved by rest (6). The etiology of fatigue is unknown; however, studies agree on its multidimensionality (3, 8). Fatigue is categorized into peripheral and central components (9). In FMS and chronic fatigue syndrome, peripheral fatigue (physical fatigue) has been associated with the reduction of muscle contraction from impaired energy resources (10), while central fatigue (mental fatigue) has been associated with cognitive impairment (9, 10). In FMS, fatigue manifests within a cluster of symptoms that includes pain, sleep disorders, depression, difficulty with concentration, and worsening memory (11). Previous FMS studies reported that pain and depression are strongly associated with fatigue, while sleep quality has moderate and inverse association with fatigue (6). Psychobehavioral symptoms reported by FMS patients include depression, anxiety, and catastrophizing (12). The mutual relationship between behavioral symptoms, specifically depression and pain, have been demonstrated indicating that both share common biological pathways and neurotransmitters (13). The association between depression and fatigue has also been reported in patients who have cancer (14), arthritis (15), or FMS (16). One study found that among 839 FMS patients, fatigue was significantly associated with depression while pain was associated with anxiety (17). Pain catastrophizing was also found to be correlated with pain intensity (18). Our recent review demonstrated that catastrophizing has a large impact on fatigue severity (19); however, only one study has explored the associations among catastrophizing, pain, and fatigue, altogether in FMS patients (20).

The multidimensional model of fatigue was investigated in one study of patients with rheumatoid arthritis (RA) using structural equation modeling (SEM) and results from that study suggested that fatigue is significantly linked to disease activity, psychological factors, and sleep (21). FMS is a polysymptomatic condition that is characterized by manifest (e.g. fatigue intensity) and latent variables (e.g. cognitive vulnerabilities). The SEM approach is advantageous to use in this complex, polysymptomatic condition because it tests interrelationships among observable and latent variables. Compared to other cluster analytical strategies, the SEM approach is the only technique that can do complete and simultaneous analyses of the relationships between these variables (22).

No studies investigated the association of fatigue with other symptoms experienced by individuals with FMS using the SEM approach. This study developed a symptoms model describing the symptoms experience of FM patients based on existing literature of the

relationship among polysymptomatic distress experienced by FM patients are attributed to specific symptoms such as pain, depression, anxiety, catastrophizing, cognitive dysfunction, daytime sleepiness, and fatigue. Then we utilizes the statistical approach of the previous RA study (21), to address its purposes which was to develop a symptoms cluster model that can describe the associations of fatigue with other FMS symptoms and to explore FMS symptom sub-clusters based on varying symptom intensities reported by the sample.

MATERIALS AND METHODS

Participants

This study is part of a prospective, longitudinal, observational study from an Institutional Review Board protocol. Participants diagnosed with FMS using the 1990 (self-report history of widespread pain with at least 11 out of 18 tender sites on exam) or the 2010 American College of Rheumatology criteria (Widespread Pain Index (WPI) $\geq 7/19$ and Symptom Severity Scale (SSS) $\geq 5/12$, or WPI = 3 – 6/19 and the SS $\geq 9/12$) were included in the analyses. Based on the hypothesized model for SEM, 2 latent variables (variable that contains subscales) and 10 observed variables (variables that can be directly measured) were tested. To achieve statistical power at level of 0.80 with 0.05 significant level, sample size was calculated using a-priori sample size calculator for structural equation modeling software (23). The minimum number participant needed for model structure is 100.

Design

Participants' demographic information and the symptoms scores were obtained on one initial outpatient visit. No intervention was provided in this cross-sectional study. Patients were receiving a wide array of different therapeutic interventions obtained from community physicians and practitioners at the time of the study visit.

Measures

Demographic data (gender, age, marital status, educational level and employment status) were obtained from the participants' medical charts. Participants' symptom experiences were assessed using methods described below.

Polysymptomatic distress was measured by the sum of the widespread pain index (WPI) and symptoms severity scale (SSS) scores. The WPI measures the number of bodily areas (total = 19) that a patient has had pain in over the past week. The SSS is a 4-item, 0–3 rating scale (total = 12) to measure severity of unrefreshed waking, cognitive problems, fatigue and other somatic symptoms (e.g., irritable bowel syndrome, numbness/tingling, dizziness, depression, constipation, nausea, nervousness, chest pain, blurred vision, fever, dry mouth, itching, wheezing, Raynaud's phenomenon, hives/welts, ringing in ears, heartburn, oral ulcers, loss of/change in taste, seizures, dry eyes, shortness of breath, loss of appetite, sun sensitivity, hearing difficulties, easy bruising, hair loss, and frequent urination). Higher scores for both instruments indicate widespread painful bodily locations and more severe symptoms, respectively (24). These questionnaires have been validated in previous studies and are currently used as part of the 2010 FMS diagnostic criteria (24, 25).

Number of tender points reported (conducted by applying < 4 kilogram on 18 bodily areas) and the participant's pain threshold (the average kilogram tolerated on the 18 bodily areas) were measured using dolorimetry, a reliable tool to measure tenderness in FMS (26).

The Brief Pain Inventory-Short Form (BPI-SF) measures pain intensity (4 items) and pain interference (7 items) using a numeric rating scale of 0 (no pain / interference) to 10 (pain as bad as you can imagine / complete interference) (27). The internal consistency of BPI-SF as measured by Cronbach's alpha for pain intensity is 0.88 and pain interference is 0.87 (28).

Fatigue was measured by the Multidimensional Fatigue Inventory (MFI), a 20-item self-report questionnaire composed of five subscales: general fatigue, physical fatigue, reduced activity, reduced motivation, and mental fatigue (29). Each of the five subscales is measured with 4 items using a rating scale of 1 (completely true) to 5 (no, not true), which have been found to have internal consistency reliability of Cronbach's alpha greater than 0.80 (29).

Anxiety and depression were measured by the 14-item, two-subscale, self-report Hospital Anxiety and Depression Scale (HADS) (30). Each item is rated on a 4-point Likert scale and each subscale (anxiety and depression) has a score that ranges from 0 to 21. High scores indicate greater anxiety, and depressive symptoms. Both subscales have internal consistency with Cronbach's alpha of 0.89 (30, 31).

Catastrophizing was measured using the Pain Catastrophizing Scale (PCS), a 13-item, self-report questionnaire consisting of three subscales: rumination, magnification and helplessness. Participants were asked to rate their thoughts and feelings on a 0 (not at all) to 4 (all the time) numeric scale. The internal consistency of PCS showed a Cronbach's alpha of 0.87 (32, 33).

Self-perceived cognitive difficulties were measured by the Multiple Ability Self-Report Questionnaire (MASQ). This instrument evaluated five domains of cognitive difficulties based on neuropsychological evaluation that include language, visual-perceptual ability, verbal memory, visual-spatial memory, and attention/ concentration (34). MASQ contains 38 items on a 1–5 Likert rating scale. The total score for each domain ranges from 8–40, except for visual-perceptual ability, which ranges from 6–30 (35). Higher MASQ scores indicate greater perception of cognitive difficulties. The internal consistency reliability of MASQ showed a Cronbach's alpha that ranged from 0.72–0.74 (36).

Daytime sleepiness was measured by the 8-item Epworth Sleepiness Scale (ESS). Each item was rated on a 4-point (0–3) Likert scale, with total scores ranging from 0 to 24, with higher ESS sum scores mean higher daytime sleepiness (37). The internal consistency of ESS showed a Cronbach's alpha of 0.70 (38).

Data analysis

Subjects were grouped based on their symptom scores. The agglomerative hierarchical cluster analysis with ward's methods and squared Euclidean distances were performed using the SPSS 19.0 program. SEM analysis was used to test and estimate the relations among polysymptomatic distress to pain severity and interference, depression, anxiety, catastrophizing, cognitive dysfunction, daytime sleepiness, and fatigue. The hypothesized

symptoms cluster model (Figure 1) was tested using the AMOS™ 17.0 program (39). Prior to testing the hypothesized symptoms cluster model, data were screened for normal distribution. Missing data were managed using the expectation maximization (EM) method, by finding the maximum log-likelihood parameters for the missing data (40). The hypothesized symptoms cluster model was assessed for multiple goodness of fit criteria to include: model chi-squared goodness of fit statistic, the ratio of chi squared to the degree of freedom (χ^2/df), the Comparative Fit Index (CFI), and the root mean square error of approximation (RMSEA). The criterion of non-significant chi-squared statistic ($p > 0.05$) was suggested as a good fit between data and the tested symptoms cluster model (41). Other indications for the goodness of fit included the χ^2/df less than 2 and a CFI of greater than .95 (42). The RMSEA value of less than .06 was considered acceptable to minimize type I and II errors (43). Symptoms cluster model modifications were performed to interpret the model fit. Using a previous approach (44), relationship paths that were theoretically justifiable and empirically explainable based on existing literature were added and the non-significant paths were dropped.

To identify the number of distinct sub-clusters from the data, the maximum percentage change in the agglomeration coefficient recorded between successive sub-cluster profiles was used. Following the formation of sub-clusters, discrimination function analysis using the SPSS19.0 program was conducted to investigate the relative weight of each predictive variable in discriminating between the sub-clusters. Multiple analyses of variance (MANOVAs) were performed to investigate differences in symptom experience between sub-clusters and to explore whether the goodness of fit of the symptoms cluster model remained unchanged between the identified FMS sub-clusters. Multiple sample structural equation modeling analyses were used to examine the differences between FMS symptoms sub-clusters 1 and 2. The chi-square difference statistic was used to test differences between the constrained model where regression weights of all relationship parameters were constrained to be equal between the two sub-clusters and the unconstrained model, where regression weights of all relationship parameters were allowed to vary. If the chi-square results revealed a significant difference between the constrained and unconstrained models, then the SEM models for sub-clusters 1 and 2 were considered unique for each FMS sub-cluster. Then, each relationship path was tested to determine if relationships between variables in that path had equal regression weights in the model for both sub-clusters by allowing one relationship path to be unequal between the two sub-clusters. Significant chi-square differences between the unconstrained and constrained models indicated that the specific relationship path significantly differed between the two FMS sub-clusters.

RESULTS

Participant characteristics

The sample included a total of 120 participants with 12 males (10%) and 108 females (90%) ages 21 to 82 years (mean age = 46.30 ± 11.00). Majority of these participants were Caucasians (59%), married (37%), and college educated (41%). About 40% were employed and 25% were on disability (Table 1).

Structural equation modeling results

Hypothesized model—Based on the goodness of fit criteria, the hypothesized model shown in Figure 1 showed a poor fit to the data ($\chi^2 = 259.6$, degrees of freedom = 113, probability level < 0.01, the $\chi^2/df = 2.6$, CFI = 0.89, RMSEA = 0.10). The hypothesized model was modified based on the theoretical and statistical plausibility of the data using the modification indices suggested by the AMOS™ program. The daytime sleepiness variable was dropped from the FMS symptoms cluster model because it was not associated with fatigue and any of the other variables. Only statistically significant paths ($p < 0.05$) as shown in Figure 2 and Table 3 were included in the model. The modified model had a better fit to the data ($\chi^2 = 121.7$, $df = 98$, $p > 0.05$, $\chi^2/df = 1.24$, CFI = 0.98 and RMSEA = 0.04) than did the hypothesized model. The most striking observation from the FMS symptoms cluster model was the negligible impact of pain symptoms on the severity of fatigue, cognitive dysfunction, anxiety, and depression of FMS.

Cluster analysis and discriminant function analysis—Two FMS sub-clusters were identified as being a good fit to the data based on the largest agglomeration coefficient difference of 1431.6. The two FMS sub-clusters had approximately equal percentages of male (9.6% in sub-cluster 1 and 11.5% in sub-cluster 2) and were similar ($p > 0.05$) in age, BMI, daytime sleepiness, and self-perceived cognitive dysfunction on visual-perceptual ability, verbal memory, visual-spatial memory, and attention/concentration. Sub-cluster 1 included 94 FMS subjects (78%) and sub-cluster 2 had 26 subjects (22%). Comparing symptom severities, sub-cluster 1 subjects complained of more intense symptoms compared to sub-cluster 2 and these symptoms include higher ($p < 0.05$) Widespread Pain Index (mean \pm SD sub-cluster 1 = 14.18 ± 2.5 ; sub-cluster 2 = 5.77 ± 2.4), higher tender point count (sub-cluster 1 = 14.77 ± 4.2 ; sub-cluster 2 = 12.50 ± 5.4), and lower pain threshold as measured by dolorimetry (sub-cluster 1 = 2.69 ± 1.4 ; sub-cluster 2 = 3.60 ± 1.8 ; Table 1).

The discriminant function analysis results shown in Table 2 indicated that the two FMS sub-clusters were significantly distinct from each other ($\chi^2 (df) = 130.990 (5)$, $p < 0.001$), suggesting that each sub-cluster has its own distinct symptoms characteristics. Widespread pain and somatic symptoms were important in separating the 2 FMS sub-clusters. Based on function loading (table 2), FMS sub-cluster 1 was distinguished from FMS sub-cluster 2 by widespread pain, unrefreshed waking, and somatic symptoms, while FMS sub-cluster 2 had fatigue and cognitive symptoms coupled with less intense and widespread pain that was distinct from FMS sub-cluster 1. About 94% of sub-cluster 1 subjects (widespread pain cluster) and 42% of sub-cluster 2 subjects (fatigue cluster) met the 2010 FMS diagnostic criteria of WPI ≥ 7 and SSS ≥ 5 , while no sub-cluster 1 subject and about 23% of sub-cluster 2 subjects met the 2010 FMS diagnostic criteria of WPI = 3–6 and SSS ≥ 9 . A small proportion of patients (6% in sub-cluster 1, 35% in sub-cluster 2) met only the 1990 criteria, reminding that clinically substantial tenderness can occur in the absence of self-reported widespread pain when measured in a clinical setting.

Multiple-sample SEM analysis—To examine the conformity of the FMS symptoms cluster model for the two identified FMS sub-clusters, a multiple-sample SEM analysis was conducted to compare the fully constrained FMS symptoms cluster model (all parameter

estimates in the model between the two samples were the same) and the unconstrained FMS symptoms cluster model (all parameter estimates in the model between two samples varied). The results showed a significant difference between the fully constrained and unconstrained models ($\chi^2 = 44.3$, $df = 28$, $p = 0.03$), indicating that each FMS sub-cluster has a distinct symptom sub-cluster model.

The significant paths that differed between the 2 sub-clusters using multiple sample SEM analyses are presented in Figure 3. The paths between pain intensity and pain interference (sub-cluster 1: $\beta = 0.54$, $p < 0.05$, sub-cluster 2: $\beta = 1.06$, $p < 0.05$) and between symptoms severity and mental fatigue (sub-cluster 1: $\beta = 0.40$, $p < 0.05$, sub-cluster 2: $\beta = 0.95$, $p < 0.05$) were significantly different in FMS sub-cluster 2, compared to the FMS sub-cluster 1 model. These FMS symptoms sub-cluster models suggest that pain appears to only influence physical fatigue, and only in FMS patients whose symptom experience is primarily defined by pain. The paths between catastrophizing and total fatigue and between pain interference and physical fatigue became non-significant ($p > 0.05$) in FMS sub-cluster 2.

DISCUSSION

The FMS symptoms cluster model developed in this study using structural equation modeling showed that fatigued FMS subjects have high pain severity, cognitive dysfunction, depression, anxiety, and catastrophizing, a consistent finding from previous studies (44, 45). Further investigation of this FMS symptoms cluster model revealed clinical sub-clusters of FMS patients. Previous studies identified heterogeneous clinical subgroups of patients with FMS based on their symptoms using the Multidimensional Pain Inventory (46), the Short Form 36 Health Survey (47), and the Visual Analog subscale of the Fibromyalgia Impact Questionnaire (48). One study categorized FMS subjects into three subtypes based on mood, cognition, and hyperalgesia (49). Our study reported 2 FMS clinical sub-clusters, sub-cluster 1 (defined by pain) and sub-cluster 2 (defined by fatigue). This provides further evidence that FMS is best considered an illness of polysymptomatic distress rather than a primary pain disorder.

One distinguishing feature between the FMS sub-cluster models points to how increasing pain intensity had stronger impact on pain interference in FMS sub-cluster 2 subjects (sub-cluster defined by fatigue) compared to FMS sub-cluster 1 subjects (defined by pain), suggesting that FMS sub-cluster 1 subjects may have adapted to their daily, persistent and more intense widespread pain compared to FMS sub-cluster 2 subjects. Further, FMS sub-cluster 2 subjects also complain of worst mental fatigue with increasing symptom severity, compared to FMS sub-cluster 1 subjects, confirming that fatigue is a more bothersome symptom for FMS sub-cluster 2 subjects with increasing symptom severity, compared to FMS sub-cluster 1 subjects.

The observation that 35% of the subjects in sub-cluster 2 did not meet the 2010 FMS diagnostic criteria reminds that the 2010 FMS diagnostic criteria is most sensitive in capturing moderate to severe widespread pain symptoms. These results also demonstrate that a minority of persons can demonstrate the substantial widespread tenderness, as indicated by meeting the 1990 ACR criteria, despite not having substantial widespread pain

within the last week. While there is a close relationship between clinical pain reporting and tenderness, it is not absolute.

Our FMS symptom sub-cluster models revealed that high pain intensity was not significantly associated with fatigue severity, but high depression was, which is consistent with previous findings in FMS (45) and other chronic conditions (e.g., rheumatoid arthritis, cancer) (14, 50). This important observation suggests that pain is not the main driver of the other symptoms experienced by FMS subjects. In fact, we also found in this study that depression is significantly associated with total fatigue, pain interference with physical fatigue, and cognitive dysfunction with mental fatigue. Previous studies in FMS and chronic fatigue syndrome reported a similar significant association between cognitive dysfunction and fatigue (51). The influence of cognition and the perception of fatigue support the proposed central mechanism of fatigue, suggesting that the sensation of fatigue is controlled by the combination of the afferent feedback from the periphery, current knowledge of the external environment, and prior experiences (51).

The study findings cannot be generalized because of limitations. First, this is a cross sectional study that attempted to identify relationships among self-reported symptoms. It is commonly known that the intensities of FMS symptoms change frequently. Therefore, FMS patients might shift from one sub-cluster to another based on the varying intensities of symptoms, day after day. Although this structural equation analysis precludes an interpretation of causality or directionality among symptoms, our FMS symptoms model was examined based on hypothesized relationships of symptoms from existing literature. Other plausible linkages among these variables may need to be investigated, because their indirect and reciprocal relationships were not tested. Therefore, a longitudinal study is warranted to explore causal relationships between the variables being investigated in this study. Additional investigations to include physiological measurements such as physical performance assessments, cognitive function tests, autonomic dysfunction measures, and biological profiling such as genetic and pro-inflammatory markers from a larger number of subjects, will be useful.

Conclusions

Our study suggested two heterogeneous categories of patients with FMS based on their symptom experiences. These symptoms sub-cluster models provide relevant information that can be useful for clinical diagnosis and management of FMS. A longitudinal study capturing the variability in symptom experience of FMS subjects and comparing the symptom sub-cluster models reported in our cross-sectional study will be informative.

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Significance and Innovation

- This study investigated the symptoms cluster experienced by FMS patients using the score from FMS 2010 diagnostic tool with sophisticated statistical analysis.
- The result of this study is a first step to help clinicians classify and provide personalized interventions for FMS patients.
- Our study suggested 2 FMS sub-clusters and demonstrated the differences between the 2 sub-clusters

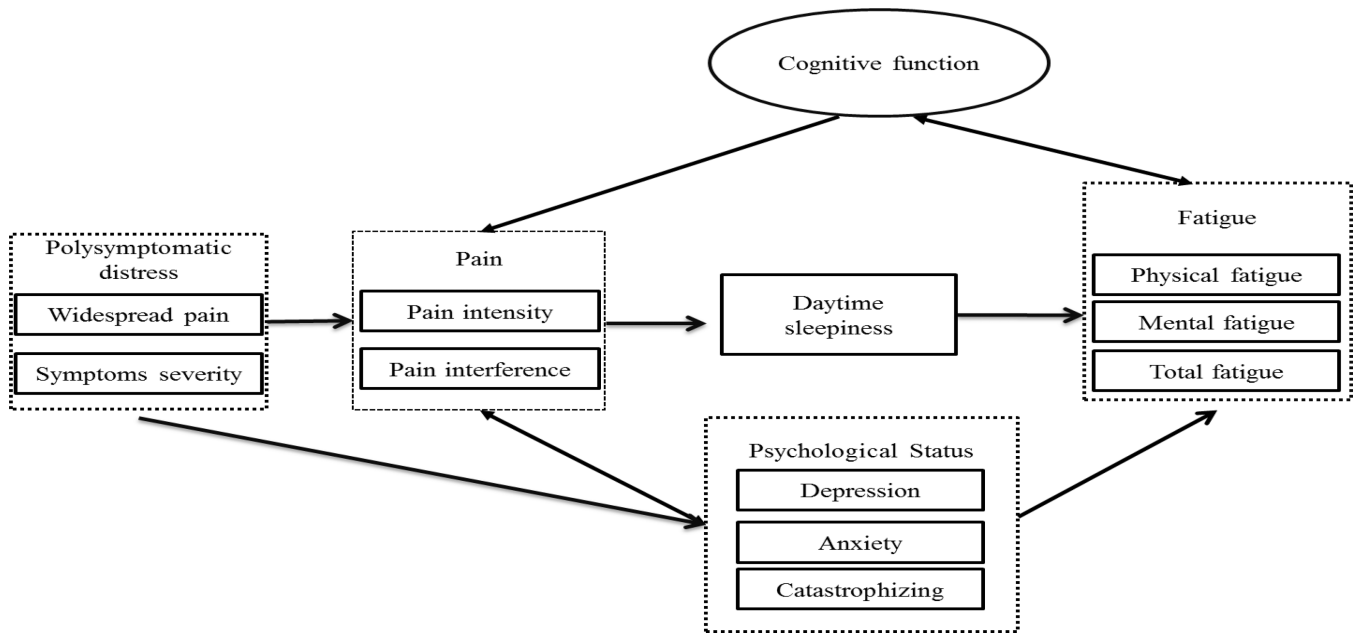


Figure 1. Hypothesized FMS symptoms cluster model. Hypothesized structural model on the relationship between polysymptomatic distress, pain, psychological status, cognitive function, daytime sleepiness, and fatigue.

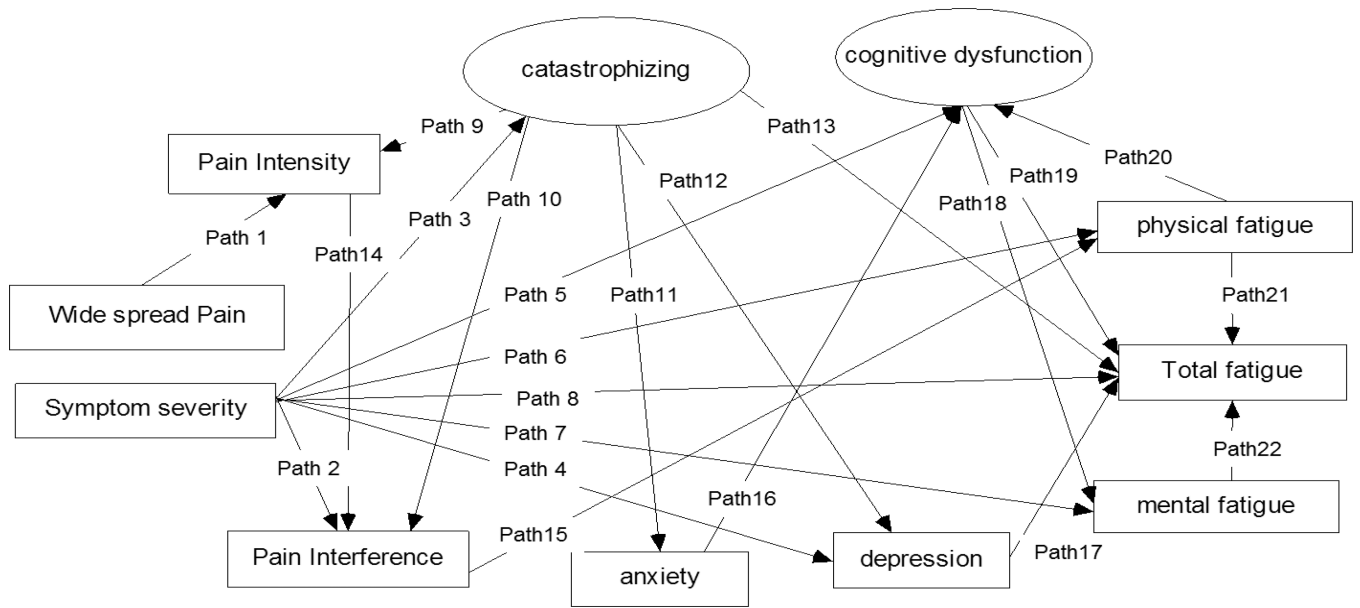


Figure 2.
Final FMS symptoms cluster model using structural equation modelling.

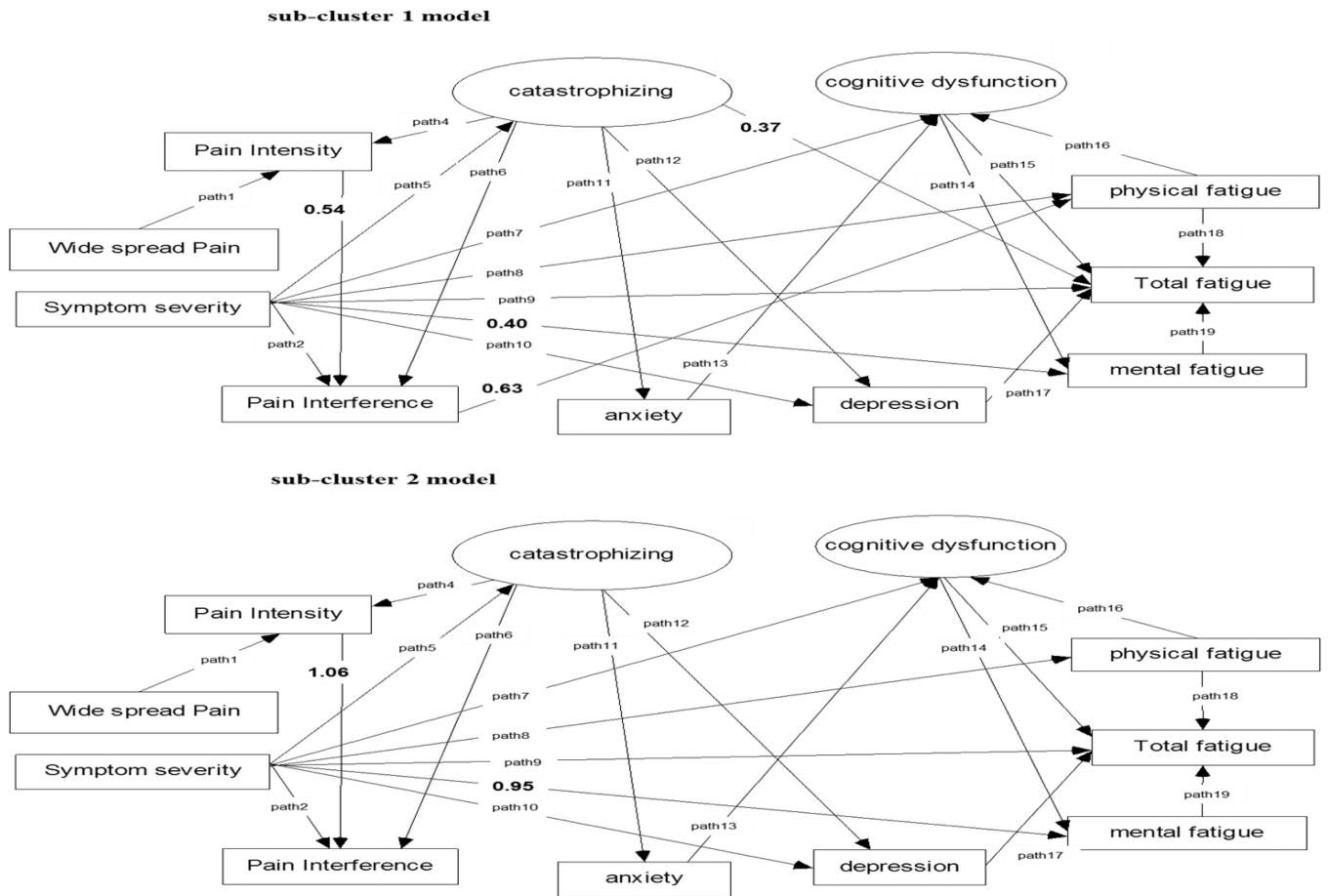


Figure 3. FMS symptoms sub-cluster models. Error terms not shown. In symptom sub-cluster model 2, the path from pain interference to physical fatigue and the path from catastrophizing to total fatigue were not significant.

Table 1

Demographic and clinical characteristics of sample

Characteristics	n	%	Range	Mean (SD)
Gender				
Male	12	10		
Female	108	90		
Age				
21–30	12	10	21.00–82.00	46.30 (11.00)
31–40	26	21.7		
41–50	35	29.2		
51–60	38	31.7		
>61	9	7.5		
Race				
Caucasian/White	71	59.2		
African American	35	29.2		
Hispanic/Asian	4	3.3		
Others	4	3.3		
Missing data	6	5.0		
Education				
Less than 12 grade	6	5.0		
12 grade	20	16.7		
Trade School	4	3.3		
College	49	40.8		
Graduate School	35	29.2		
Missing data	6	5.0		
Marital status				
Never Married	32	26.7		
First marriage	44	36.7		
Divorces	22	18.3		
Widowed	3	2.5		
Remarried	12	10.0		

Characteristics	n	%	Range	Mean (SD)	Cluster 1 Mean (SD) (N = 94)	Cluster 2 Mean (SD) (N = 26)	F univariate; P value
Missing data	7	5.8					
Employment							
Full time	35	29.2					
Part time	13	10.8					
Unemployed	21	17.5					
Disability	30	25.0					
Students	8	6.7					
Retired	6	5.0					
Missing data	7	5.8					
Pain Threshold							
# of tender points	120		0–18.0	14.28 (4.54)			
Pain threshold (kg)	120		0–8.13	2.89 (1.56)			
Gender							
		Female		85 (90.4%)	23 (88.5%)		
		Male		9 (9.6%)	3 (11.5%)		F = 1.11; p = 0.34
Age				46.30 (11.3)	46.30 (10.2)		
		Caucasian		49 (52.1%)	22 (84.6%)		
		African American		33 (35.1%)	2 (7.7%)		
		Other		12 (11.8%)	2 (7.7%)		
BMI				31.14 (8.15)	31.39 (8.8)		F = 0.05; p = 0.82
Polysymptomatic distress		WPI		14.15 (2.5)	5.77 (2.4)		F = 238.4; p = 0.00**
		SSS-fatigue		2.45 (0.7)	1.96 (0.8)		F = 8.87; p = 0.00**
		SSS- unrefreshed waking		2.46 (0.8)	2.04 (1.0)		F = 5.09; p = 0.03*
		SSS-cognitive symptoms		1.83 (0.9)	1.35 (1.0)		F = 5.71; p = 0.02*
		SSS-somatic symptoms		2.17 (0.6)	1.65 (0.6)		F = 16.47; p = 0.00**
Tender point		Number of tender points		14.77 (4.2)	12.50 (5.4)		F = 5.266; p = 0.02*
		Average pain threshold		2.69 (1.4)	3.60 (1.8)		F = 7.44; p = 0.01*
Pain		Pain Intensity		6.02 (1.7)	3.62 (2.1)		F = 35.37; p = 0.00**

	Cluster 1 Mean (SD) (N = 94)	Cluster 2 Mean (SD) (N = 26)	F _{univariate} ; Pvalue
Pain Interference	6.48 (2.0)	4.05 (2.9)	F = 23.98; p = 0.00**
Physical fatigue	16.03 (3.36)	13.77 (3.09)	F = 9.54; p = 0.00**
Mental fatigue	15.25 (3.9)	13.19 (4.1)	F = 5.57; p = 0.02*
Total fatigue	76.98 (13.4)	64.84 (12.3)	F = 17.37; p = 0.00**
Day time sleepiness	9.28 (4.5)	8.55 (3.67)	F = 0.59; p = 0.45
Depression	8.39 (3.9)	5.96 (3.8)	F = 7.88; p = 0.01*
Anxiety	9.69 (4.5)	7.12 (4.2)	F = 6.85; p = 0.01*
Catastrophizing	8.12 (4.7)	4.82 (3.8)	F = 10.74; p = 0.00**
	4.72 (3.2)	2.55 (2.3)	F = 10.67; p = 0.00**
Helplessness	10.86 (6.3)	5.65 (5.4)	F = 14.85; p = 0.00**
Cognitive dysfunction	20.50 (4.6)	17.65 (4.3)	F = 8.00; p = 0.01*
Visual-perceptual ability	15.27 (5.1)	13.73 (4.1)	F = 1.97; p = 0.16
Verbal memory	22.82 (4.9)	21.66 (4.7)	F = 1.07; p = 0.30
Visual-spatial memory	18.94 (4.8)	17.31 (4.0)	F = 2.49; p = 0.12
Attention/concentration	22.77 (5.5)	21.10 (4.7)	F = 2.00; p = 0.16

* p < 0.05;

** p < 0.01;

SD = standard deviation; BMI = body mass index; WPI = widespread pain index; SSS = symptoms severity score.

Table 2

Discriminant function analysis for sub-cluster characteristics

		Saturation loading	Classification function coefficients	
			FM Sub-cluster 1	FM Sub-cluster 2
Widespread pain		0.979	2.217	0.815
Symptom severity	Fatigue	0.189	1.167	1.293
	Unrefreshed waking	0.143	2.078	1.674
	Cognitive symptoms	0.152	-1.092	-0.547
	Somatic symptoms	0.257	5.185	3.960
Constant			-24.990	-8.926

χ^2 (df) = 130.990 (5), $p < 0.001$; Eigenvalue of function 1 = 2.108; 100% variance

Standardized path coefficient, standard error, unstandardized path coefficient of significant paths in final FMS symptoms cluster model (N = 120)

Table 3

Path	Predictors	Dependent variables	β	SE	Unstandardized path coefficient
1	Widespread Pain index	Pain Intensity	.201	.036	5.550***
2	Symptom severity	Pain Interference	.258	.058	4.449***
3		Catastrophizing	.502	.169	2.978**
4		Depression	.470	.115	4.079***
5		Cognitive dysfunction	.311	.145	2.150*
6		Physical fatigue	.433	.128	3.381***
7		Mental fatigue	.546	.127	4.291***
8		Total fatigue	1.038	.270	3.848***
9	Catastrophizing	Pain Intensity	.164	.039	4.174***
10		Pain Interference	.146	.036	4.039***
11		Anxiety	.690	.088	7.861***
12		Depression	.422	.082	5.146***
13		Total fatigue	.311	.150	2.076*
14	Pain Intensity	Pain Interference	.665	.068	9.723***
15	Pain Interference	Physical fatigue	.485	.125	3.887***
16	Anxiety	Cognitive dysfunction	.268	.068	3.921***
17	Depression	Total fatigue	.337	.153	0.199*
18	Cognitive dysfunction	Mental fatigue	.589	.098	5.994***
19		Total fatigue	-.653	.222	-2.943**
20	Physical fatigue	Cognitive dysfunction	.303	.099	3.056**
21		Total fatigue	2.077	.171	12.174***
22	Mental fatigue	Total fatigue	1.619	.181	8.965***

* p < 0.05,

** p < 0.01,

β: Standardized path coefficient; WPI: Widespread Pain Index; SSS: Symptoms Severity Scores

;1.000) < p

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