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Article

Cognitive Functioning in Survivors of Toxic Oil Syndrome: A Case-Control Study Four Decades After the Epidemic

Running Title: Cognitive Functioning After Toxic Oil Syndrome

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Abstract: Toxic Oil Syndrome (TOS) was a major food-borne epidemic that occurred in Spain in May 1981, caused by the ingestion of rapeseed oil adulterated with aniline. Although its somatic sequelae have been extensively documented, the long-term cognitive consequences remain poorly understood more than four decades after exposure. In this case-control study, 50 individuals with clinically confirmed TOS were compared to 50 healthy controls matched for age, sex, and education. All participants completed a comprehensive neuropsychological assessment alongside validated questionnaires assessing fatigue, anxiety, depression, and health-related quality of life. Multivariate regression models adjusted for demographic and vascular risk factors revealed significantly poorer performance among TOS patients in attention, executive function, processing speed, and global cognition. However, these differences were no longer significant after further adjustment for mood symptoms, fatigue, and central nervous system-acting medication use. Structural equation modeling indicated that affective symptoms—particularly fatigue—substantially mediated the relationship between TOS and cognitive function. The cognitive pattern observed is reminiscent of disorders characterized by subcortical involvement and disrupted neural connectivity, such as multiple sclerosis and vascular cognitive impairment. Although postmortem studies have not identified overt white matter lesions in early TOS cases, the current findings suggest possible long-term alterations in neural networks, potentially involving both white and gray matter substrates. These results underscore the need to consider mood and fatigue symptoms when evaluating cognition in TOS survivors and highlight the potential for diffuse, enduring neurobiological consequences of the original toxic exposure.

Keywords: toxic oil syndrome; cognitive impairment; case-control study; long-term effects; fatigue; depression; anxiety; neurotoxicity

1. Introduction

Toxic oil syndrome (TOS) emerged as one of the most devastating public health crises in contemporary Spanish History, first identified in May 1981 during a massive outbreak of food poisoning. The source was traced to rapeseed oil denatured with aniline—intended for industrial use but fraudulently marketed as olive oil for human consumption [1,2]. The epidemic affected over 20,000 individuals and resulted in more than 300 deaths within the first year alone [2,3]. Survivors experienced a wide range of chronic, multisystemic sequelae, the most prominent being muscle atrophy, typically secondary to an eosinophilic inflammatory myopathy [3–6]. Other long-term complications included myalgias, muscle cramps [7] severe weight loss, pulmonary hypertension, scleroderma-like syndromes, joint contractures, Sjögren's syndrome, alopecia, pruritus, and chronic hepatitis [3].

Neurological involvement has been consistently observed in this population and can be divided into two major domains. The first involves the peripheral nervous system, with symptoms such as numbness, paresthesia, and hypoesthesia attributed to inflammatory neuropathy and perineural fibrosis [4,5,7]. The second includes central nervous system manifestations—insomnia, chronic headache, and memory disturbances—suggesting a potential central neurotoxic process [7,8].

Cognitive impairment has been one of the most functionally impactful sequelae in TOS survivors and was evident even in the early stages of the disease. Impairments have been documented in short- and long-term memory, particularly affecting semantic and episodic memory in both verbal and non-verbal formats [8]. Patients frequently display reduced attention, poor concentration, mental fatigue, and psychomotor slowing, with slower processing speeds and delayed reaction times [7,8]. While some of these deficits may appear mild, their cumulative effect can significantly impair daily functioning, supporting the hypothesis of persistent central nervous system involvement [7,8].

Despite the magnitude of the epidemic and the chronicity of its effects, the long-term cognitive trajectory of TOS survivors has been poorly characterized. Indeed, no studies have evaluated cognitive function in this population more than four decades after the exposure.

In this context, the present study aims to assess whether individuals affected by TOS continue to exhibit measurable cognitive deficits 43 years after the outbreak. While earlier studies have identified neurocognitive alterations in this population [7,8], it remains unclear whether these impairments persist, progress, or resolve over time when compared to demographically matched individuals from the general population.

2. Methods

2.1. Standard Protocol Approvals, Registrations, and Patient Consents.

The ethical standards committees approved all procedures on human experimentation at the University Hospital "12 de Octubre," Spain (CEIC codes: 17/035 and 23/616). We obtained written (signed) informed consent from all participants.

2.2. Study Design and Setting

Between April and June 2024, all TOS case participants and healthy controls were recruited from the province of Madrid, one of the regions most severely affected during the 1981 epidemic.

We designed the study as a case-control study, with the exposed group consisting of individuals diagnosed with toxic oil syndrome and the unexposed group comprising healthy controls. All interviews, cognitive assessments, and study procedures were conducted at the 12 de Octubre University Hospital in Madrid, Spain.

2.3. Participants

TOS cases were defined using the same diagnostic criteria applied in previous studies [9]. Eligible participants included individuals who had experienced either the acute or chronic phase of the disease. The acute phase was characterized by alveolar-interstitial pulmonary infiltrates and/or pleural effusion in the presence of absolute eosinophilia (>500 cells/mm³). The chronic phase was defined by the presence of myalgia and eosinophilia and/or one or more of the following clinical features clearly attributable to TOS: scleroderma-like skin changes, peripheral neuropathy, pulmonary hypertension, or hepatopathy.

We recruited TOS patients from the monographic clinical unit dedicated to toxic oil syndrome at the 12 de Octubre University Hospital in Madrid—currently, the only specialized unit in Spain exclusively devoted to the long-term management of these patients.

Patients were contacted consecutively until the target sample size of 50 participants was achieved. The referent (unexposed) group was recruited from friends and acquaintances residing in the same geographic area. The exposed group included 50 adults who had been exposed to toxic oil 43 years earlier and had developed clinically confirmed TOS. These individuals were frequency-matched to 50 unexposed referents by age (± 5 years), sex, and educational level.

A post hoc power analysis indicated that the sample size ($n = 50$ per group) achieved 87.85% power to detect a standardized mean difference of 0.4 (two-tailed $\alpha = 0.05$), observed in the Global Cognitive Score, the study's primary cognitive outcome.

Patients were excluded if they had a diagnosis of neurodegenerative disorders (e.g., Alzheimer's disease, Parkinson's disease) with worse cognitive performance, renal disease, cerebrovascular accidents, chronic alcoholism, or any traumatic injury involving the brain, spinal cord, or peripheral nervous system. Referent (control) participants were required to meet the same exclusion criteria as the exposed group.

2.4. Measurements

2.4.1. Demographic and Clinical Data

Personal data—including age, sex, educational attainment, medical history, and current treatments—were collected using a standardized questionnaire. Educational level was initially recorded in four categories: “incomplete primary,” “primary,” “secondary or higher,” and “university.” For analytical purposes, these were subsequently dichotomized as “primary or below” (including those with completed primary and those with no formal or incomplete primary schooling) versus “secondary or higher education”. Particular attention was paid to the documentation of medications with potential cognitive effects, specifically central nervous system (CNS)-acting drugs, including anxiolytics, stimulants, antipsychotics, antidepressants, antihistamines, and antiepileptic medications.

2.4.2. Fatigue Measurement

Fatigue was assessed using the Fatigue Impact Scale for Daily Use (D-FIS), a brief eight-item self-report questionnaire specifically designed to evaluate the perceived impact of fatigue on daily functioning [10,11]. Each item is rated on a 5-point scale ranging from 0 (“no problem”) to 4 (“extreme problem”), with higher total scores indicating greater fatigue-related interference [10,11].

2.4.3. Health-Related Quality of Life Assessment

Health-related quality of life was measured using the EuroQol instrument, a well-established and validated generic tool designed to assess perceived health status in healthy individuals and patients with a wide range of medical conditions [12]. The EuroQol consists of two components [12]. The first, known as the EQ-5D descriptive system, includes five items that assess current health problems across five dimensions: mobility, self-care, usual activities, pain/discomfort, and

anxiety/depression. Each dimension is rated using three ordinal response levels: (1) no problems, (2) moderate problems, and (3) severe problems. These responses yield a health profile that can describe up to 243 unique health states. EQ-5D index values are calculated according to standardized European algorithms [12], producing a single utility score where 1 indicates full health, 0 represents death, and negative values (with a minimum of -0.109) reflect health states perceived as worse than death. The second component of the EuroQol is the EQ visual analog scale (EQ VAS), a vertical scale ranging from 0 (worst imaginable health state) to 100 (best imaginable health state), on which respondents provide a subjective rating of their overall health status.

2.4.4. Depressive Symptoms

Depressive symptoms were assessed using the Beck Depression Inventory-II (BDI-II) [13], a validated self-report instrument designed to measure the severity of depression in adults. It consists of 21 items, each reflecting a symptom or attitude associated with depression (e.g., sadness, pessimism, fatigue, changes in sleep or appetite). Respondents rate each item on a 4-point scale ranging from 0 (no symptom) to 3 (severe symptom) based on their experiences over the previous two weeks. Total scores range from 0 to 63, with higher scores indicating greater symptom severity [13].

2.4.5. Anxiety Symptoms

Anxiety symptoms were assessed using the Beck Anxiety Inventory (BAI) [14], a widely used self-report questionnaire designed to evaluate the severity of common anxiety symptoms. Respondents rate how much they have been bothered by each symptom over the past week on a 4-point scale, from 0 ("not at all") to 3 ("severely—I could barely stand it"). The total score ranges from 0 to 63, with higher scores indicating more severe anxiety [14].

2.4.6. Cognitive Performance

Cognitive functioning was assessed using NeuroTrax™ digital testing [15–17]. This digital platform enables standardized, comprehensive cognitive testing in clinical settings and has demonstrated validity and reliability across a range of populations [15,16,18,19]. Due to time constraints, testing was limited to specific cognitive domains: memory (verbal and non-verbal), attention (Go-NoGo and Stroop Interference tasks), information processing speed, executive function (Go-NoGo, Stroop Interference, and Catch Game). All test instructions were delivered in Spanish, the participants' primary language [20,21].

2.5. Statistical Analyses

All statistical analyses and figure generation were performed using Python (version 3.12.2) and R (version 4.4.2). The following Python packages were used: *pandas* (v2.2.3) for data handling, *TableOne* (v0.9.1) for descriptive statistics, *statsmodels* (v0.14.4) for regression modeling, and *semopy* (v2.3.11) for structural equation modeling.

In this study, cognitive scores were normalized using the mean and standard deviation of the control group. The same approach was applied to depression, anxiety, and fatigue scales. Z-scores were calculated for each participant in both groups. As in prior studies using NeuroTrax™, normalized scores from individual test measures (e.g., accuracy, response time) were then averaged to generate domain-specific index scores, which were subsequently averaged to produce a global cognitive score [15–17].

First, a descriptive analysis of the study population was conducted. Group homogeneity was assessed using appropriate parametric or non-parametric tests based on variable distribution. Subsequently, univariate analyses were performed to compare the two groups on the BDI-II, BAI, D-FIS, and EQ-5D index values, as well as on each NeuroTrax™ cognitive domain index and global cognitive score derived from the battery.

A series of multivariate models were constructed to assess the impact of different variables on cognitive performance. Initially, well-established confounders from previous literature were included [22–25] followed by the incorporation of intermediate clinical variables such as hypertension and diabetes. In a final model, affective variables (BDI-II and BAI) and fatigue (D-FIS) were added. A structural equation modeling approach was applied to estimate both the direct and indirect effects of these variables on cognitive outcomes.

3. Results

Data from 50 patients with TOS) and 50 matched healthy controls were analyzed (Table 1). No statistically significant differences were observed between the two groups in terms of sex distribution, age, or educational level. However, TOS patients had a significantly higher prevalence of arterial hypertension (66% vs. 18%; $p < 0.001$) and diabetes mellitus (24% vs. 6%; $p = 0.025$), as well as a greater use of CNS-acting medications (58% vs. 20%; $p < 0.001$).

Table 1. Demographic, clinical, and neuropsychological characteristics of study participants.

Variable	Overall (N = 100)	Control (N = 50)	Patient (N = 50)	P value
Sex, n (%)				0.284 ^a
Male	32 (32.0)	19 (38.0)	13 (26.0)	
Female	68 (68.0)	31 (62.0)	37 (74.0)	
Age, mean (standard deviation)	59.3 (8.0)	58.7 (8.2)	59.9 (7.8)	0.449 ^b
Education, n (%)				0.546 ^a
Illiterate or primary studies	44 (44.0)	20 (40.0)	24 (48.0)	
Secondary or higher	56 (56.0)	30 (60.0)	26 (52.0)	
Central nervous system-acting medications, N (%)	39 (39.0)	10 (20.0)	29 (58.0)	<0.001 ^a
Arterial hypertension, N (%)	42 (42.0)	9 (18.0)	33 (66.0)	<0.001 ^a
Diabetes mellitus, N (%)	15 (15.0)	3 (6.0)	12 (24.0)	0.025 ^a
EQ-5D index, median [Q1, Q3]	0.7 [0.3, 0.9]	0.9 [0.7,1.0]	0.4 [0.1, 0.8]	<0.001 ^c
Fatigue Impact Scale for Daily Use, mean (standard deviation)	12.4 (10.4)	5.2 (6.3)	19.7 (8.5)	<0.001 ^b
Beck Anxiety Inventory, median [Q1, Q3]	14.0 [3.0, 26.0]	4.0 [1.2,10.8]	22.5 [14.2, 30.8]	<0.001 ^c
Beck Depression Inventory, median [Q1, Q3]	13.0 [4.8, 21.0]	5.5 [2.0,13.8]	19.0 [12.2, 27.8]	<0.001 ^c
Global Cognitive Score, mean (standard deviation)	-0.3 (0.9)	-0.1 (0.8)	-0.5 (0.9)	0.010 ^b
Cognitive domains				
Memory, median [Q1, Q3]	0.2 [-0.6, 0.5]	0.3 [-0.4, 0.6]	-0.1 [-0.9, 0.4]	0.070 ^c
Executive function, mean (standard deviation)	-0.2 (0.9)	-0.0 (0.8)	-0.4 (1.0)	0.036 ^b
Attention, median [Q1, Q3]	-0.0 [-0.5, 0.4]	0.1 [-0.3, 0.5]	-0.2 [-1.3, 0.3]	0.024 ^c
Information processing speed, mean (standard deviation)	-0.3 (1.0)	-0.0 (0.8)	-0.6 (1.0)	0.002 ^b

^a Chi-square test; ^b Student t test; ^c Mann-Whitney U test.

Regarding affective symptoms (Figure 1), patients reported more symptoms on all scales: BDI-II (median 19.0 [IQR 12.2–27.8] vs. 5.5 [IQR 2.0–13.8]; $p < 0.001$; Figure 1A), BAI (median 22.5, interquartile range [IQR] 14.2–30.8 vs. 4.0 [IQR 1.2–10.8]; $p < 0.001$; Figure 1B), and D-FIS (mean 19.7 ± 8.5 vs. 5.2 ± 6.3 ; $p < 0.001$; Figure 1C). Their health-related quality of life, as assessed by the EQ-5D index, was significantly poorer (median 0.4 [IQR 0.1–0.8] vs. 0.9 [IQR 0.7–1.0]; $p < 0.001$; Figure 1D). A summary of the demographic and clinical characteristics is provided in Table 1.

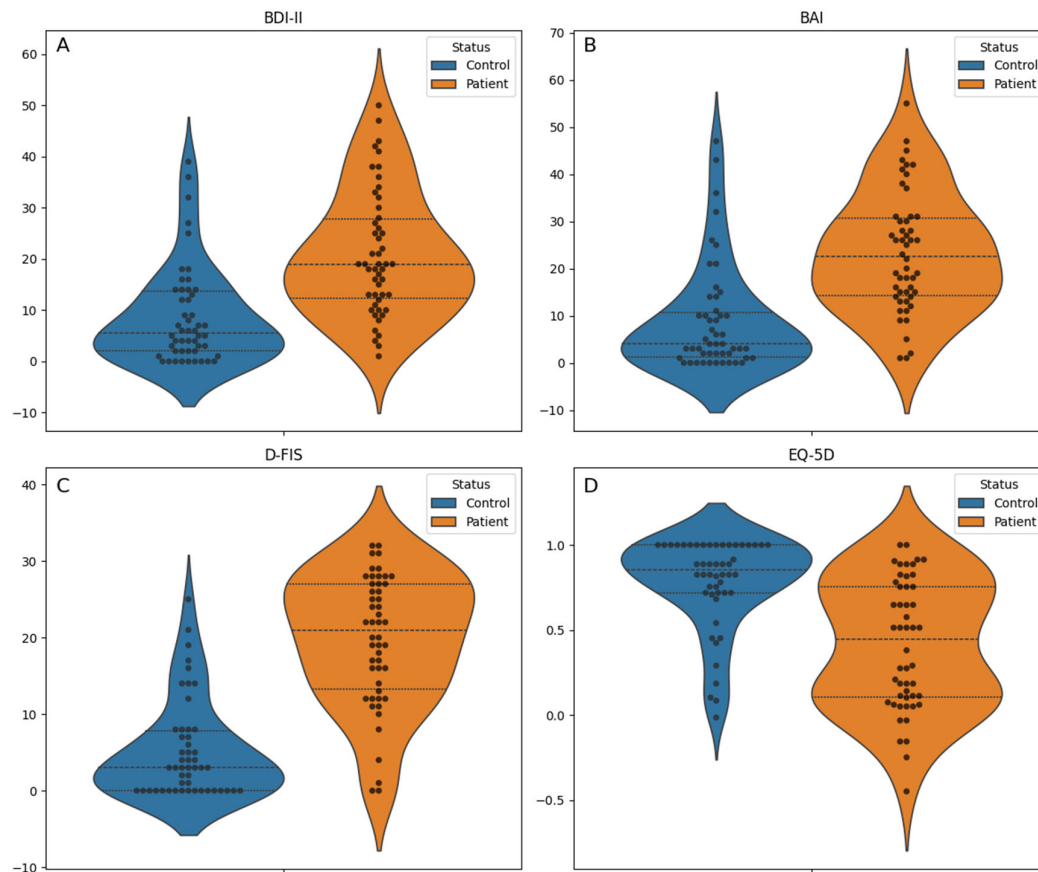


Figure 1. Group differences between patients and controls on the Beck Depression Inventory-II (BDI-II), Beck Anxiety Inventory (BAI), Fatigue Impact Scale for Daily Use (D-FIS), and EQ-5D. Data are visualized using swarm plots overlaid with medians and interquartile ranges (Q1–Q3). Higher scores on BDI-2, BAI, and D-FIS indicate greater symptom severity, while lower EQ-5D index values reflect poorer health-related quality of life.

Among the cognitive scores, no statistically significant differences were found for memory (median z-score: -0.1 [IQR -0.9 to 0.4] vs. 0.3 [IQR -0.4 to 0.6]; $p = 0.07$; Figure 2A) irrespective of whether immediate and delayed portions were combined or analyzed separately. However, TOS patients exhibited significantly poorer performance in executive function (mean z-score: -0.4 ± 1.0 vs. 0.0 ± 0.8 ; $p = 0.036$; Figure 2B), attention (median z-score: -0.2 [IQR -1.3 to 0.3] vs. 0.1 [IQR -0.3 to 0.5]; $p = 0.024$; Figure 2C), information processing speed (mean z-score: -0.6 ± 1.0 vs. 0.0 ± 0.8 ; $p = 0.002$; Figure 2D), and on the global cognitive score (mean z-score: -0.48 ± 0.18 vs. 0.05 ± 0.20 ; $p < 0.001$; Figure 2E), compared to controls. The distribution of these cognitive outcomes is illustrated in Figure 2 and Table 1.

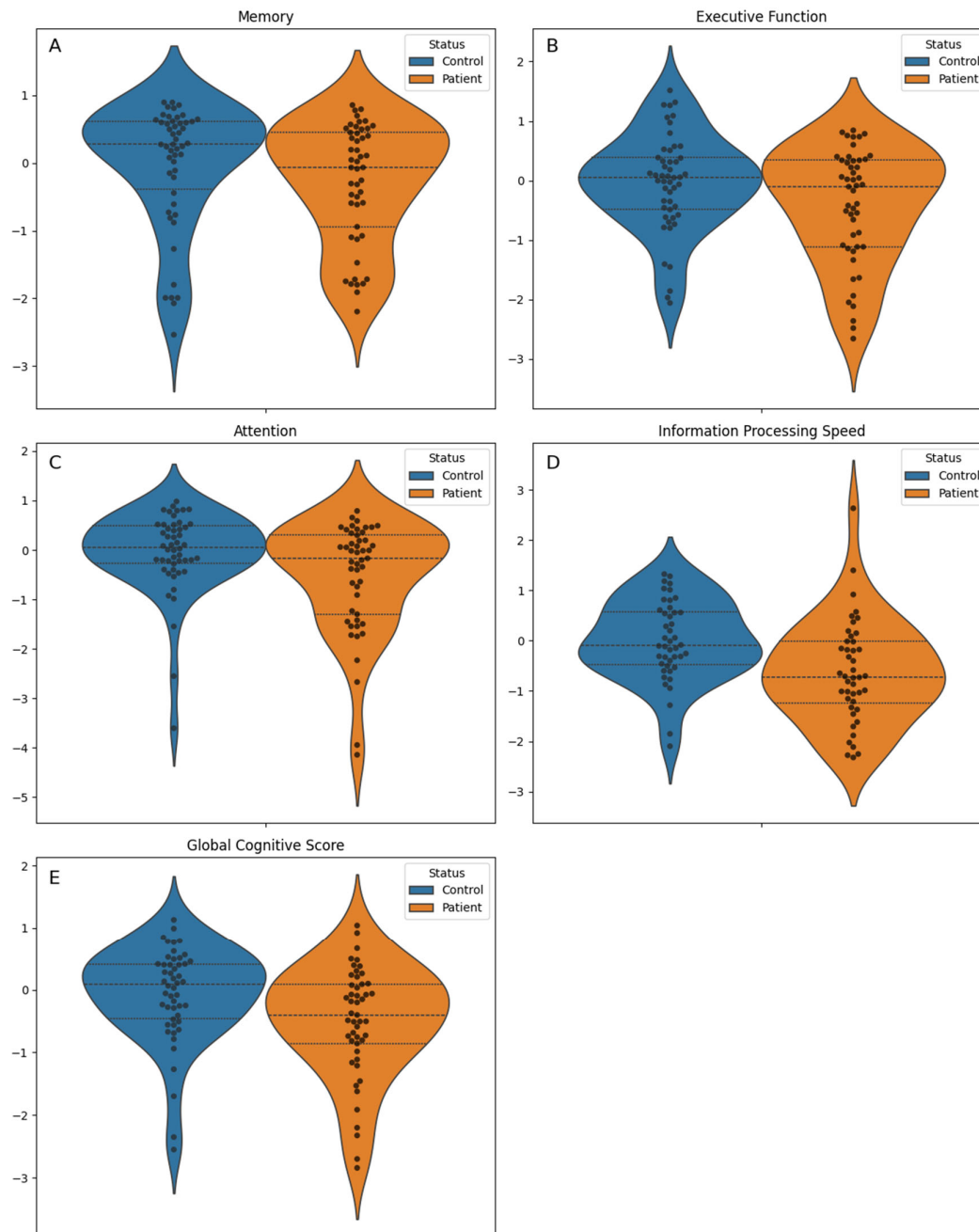


Figure 2. Comparison of cognitive performance between groups (toxic oil syndrome patients and controls). Individual test scores were transformed into z-scores using the mean and standard deviation of the control group as the reference. Domain index scores were calculated by averaging the contributing standardized test scores, and the global cognitive score was calculated as the average of all domain index scores (see Methods). Boxplots display the median and interquartile range (IQR; Q1 to Q3), with overlaid swarm plots to illustrate individual data points. Higher z-scores reflect better cognitive performance.

The models were adjusted for demographic variables (age, age squared, sex, educational level) and relevant medical history (arterial hypertension, diabetes mellitus), consistent with previous evidence linking these factors to cognitive outcomes [26–28]. The inclusion of arterial hypertension and diabetes mellitus was further justified by their higher prevalence in our sample, as well as in the broader population affected by TOS [29,30].

The multivariate models (Table 2) revealed significant associations between TOS diagnosis and reduced cognitive performance. Specifically, TOS was associated with poorer memory scores ($\beta = -0.307$, $p = 0.050$) and information processing speed ($\beta = -0.606$, $p = 0.002$) as well as lower global cognitive scores ($\beta = -0.382$, $p = 0.006$).

Table 2. Multivariable Linear Regression Models Examining the Association of Toxic Oil Syndrome Diagnosis with Cognitive Performance, Adjusted for Demographic and Clinical Covariates.

Predictor Variable	Global Cognitive Score Coefficient (<i>p value</i>)	Memory Coefficient (<i>p value</i>)	Executive Function Coefficient (<i>p value</i>)	Attention Coefficient (<i>p value</i>)	Information Processing Speed Coefficient (<i>p value</i>)
Diabetes mellitus	−0.183 (0.294)	0.044 (0.824)	−0.318 (0.107)	−0.472 (0.050)	0.170 (0.523)
Educational level	0.437 (0.001)	0.657 (<0.001)	0.229 (0.119)	0.213 (0.231)	0.644 (0.001)
Age squared	−0.002 (0.019)	−0.003 (0.004)	−0.001 (0.271)	−0.003 (0.021)	−0.000 (0.910)
Age	0.176 (0.071)	0.284 (0.012)	0.057 (0.603)	0.254 (0.058)	−0.020 (0.904)
Arterial hypertension	0.220 (0.115)	0.262 (0.101)	0.218 (0.164)	0.199 (0.295)	0.215 (0.273)
Sex (female)	−0.356 (0.006)	0.025 (0.864)	−0.517 (<0.001)	−0.453 (0.010)	−0.439 (0.018)
Toxic oil syndrome diagnosis	−0.382 (0.006)	−0.307 (0.050)	−0.274 (0.074)	−0.369 (0.048)	−0.606 (0.002)

Coefficients and p-values from multivariable linear regression models assessing the association of demographic and clinical variables (diabetes mellitus, educational level, age squared, age, arterial hypertension, sex, toxic oil syndrome diagnosis) with five cognitive measures: global cognition, memory, executive function, attention, and information processing speed. All models were adjusted for the full set of listed covariates.

In a subsequent step, additional variables—including fatigue (D-FIS), depressive symptoms (BDI-2), anxiety (BAI), and the use of central nervous system (CNS)-acting medications—were incorporated to assess potential confounding effects.

Multicollinearity among the independent variables was evaluated using the variance inflation factor. No evidence of severe multicollinearity was found among the affective variables; however, a variance inflation factor of 6.77 between BAI and BDI-2 suggested moderate collinearity. Pearson correlation analyses revealed a strong positive correlation between fatigue and BAI ($\rho = 0.72$) and between fatigue and BDI-2 ($\rho = 0.72$). Additionally, Spearman correlation coefficients between CNS-acting medication use and the affective scales indicated moderate associations: $\rho = 0.36$ with fatigue, $\rho = 0.38$ with depression, and $\rho = 0.32$ with anxiety.

To minimize collinearity issues, we derived a composite variable from the BAI and BDI-2 scores using principal component analysis. This principal component accounted for 90.9% of the shared variance between the two scales, capturing the underlying affective dimension. This composite variable, together with the D-FIS, was incorporated into separate structural equation models for each cognitive domain score. The resulting models are depicted in Figure 3.

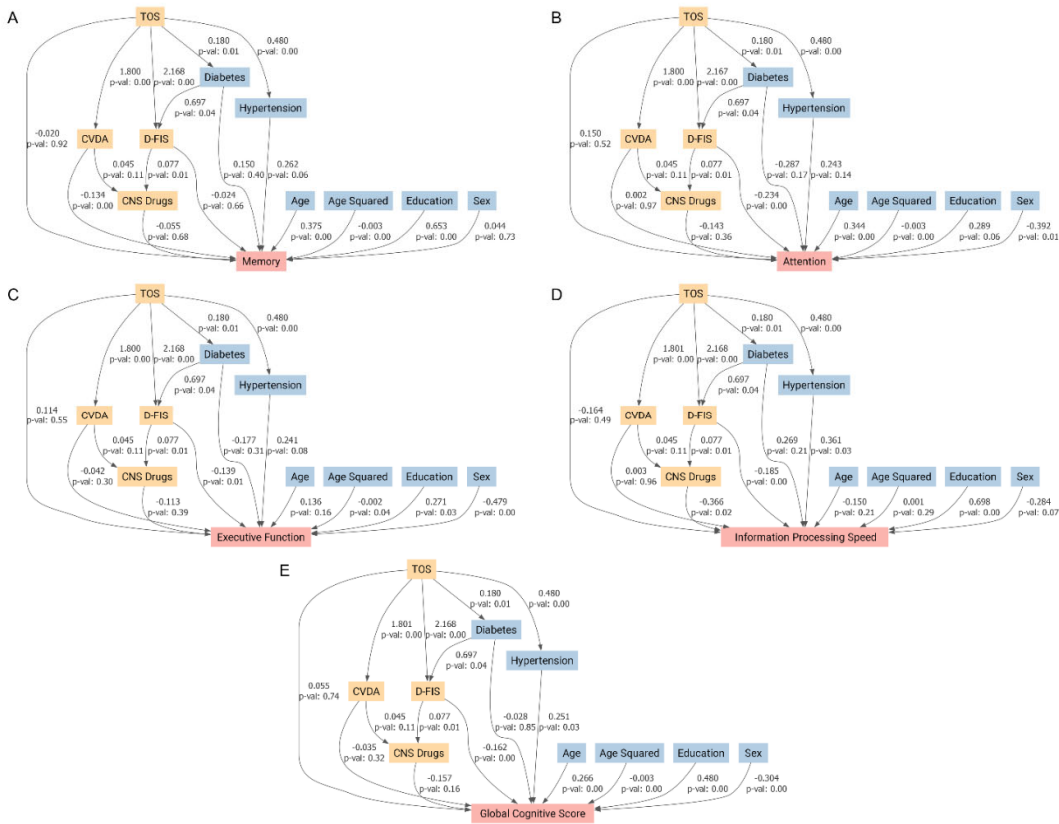


Figure 3. Structural equation modeling diagrams illustrating the relationships between toxic oil syndrome (TOS) and performance on five cognitive outcomes: A) memory, B) attention, C) executive function, D) information processing speed, and E) global cognitive score. The models include the following predictors: TOS diagnosis, sex (female), use of central nervous system-acting medications (CNS Drugs), composite variable for depression and anxiety (CVDA), and Fatigue Impact Scale for Daily Use (D-FIS). Direct and indirect effects are shown, with uncorrected p-values reported for each path.

To explore the relationships among variables and construct the structural equation models, multiple linear regression analyses were performed using the composite depression/anxiety variable, D-FIS, and the use of CNS-acting medications as independent variables. The results of these models are shown in Table 3.

Table 3. Associations of Clinical and Demographic Variables with Affective Symptoms, Fatigue, and CNS-Acting Medication Use.

Predictor Variable	Depression/Anxiety Composite Coefficient	<i>p</i> value	Fatigue Impact Scale for Daily Use Coefficient	<i>p</i> value	Use of Central Nervous System-Acting Medications Coefficient	<i>p</i> value
Diabetes mellitus	0.590	0.187	4.561	0.042	0.102	0.457
Educational level	0.020	0.951	2.620	0.108	−0.067	0.502
Age	−0.003	0.900	0.020	0.843	0.002	0.742
Arterial hypertension	−0.003	0.994	1.345	0.445	0.024	0.824
Sex (female)	0.040	0.902	1.304	0.418	0.117	0.243
Toxic oil syndrome diagnosis	1.695	<0.001	13.062	<0.001	0.328	0.003

Coefficients and p-values from linear regression models evaluating the associations of clinical and demographic predictors (diabetes mellitus, educational level, age, arterial hypertension, sex, and toxic oil syndrome diagnosis) with three outcomes: composite depression/anxiety score, Fatigue Impact Scale for Daily Use, and central nervous system-acting medications.

In these models (Figure 3), the previously observed significant impact of a TOS diagnosis on cognitive performance disappeared. However, affective variables emerged as significant predictors across various cognitive domains. Fatigue showed a significant association with attention, executive function, information processing speed, and the global cognitive score (all $p < 0.01$), whereas the composite depression/anxiety variable was significantly associated with memory performance ($p < 0.01$). Notably, the use of CNS-acting medications did not have a direct significant effect on cognitive function in any of the models.

In the mediation analysis of these models (Table 4), mediated proportions near or exceeding 1 were observed, suggesting that the effects of these confounding variables largely account for the association between TOS and cognitive performance.

Table 4. Mediation Effects of Affective Variables, Fatigue, and CNS-Acting Medications on the Relationship Between TOS and Cognitive Performance.

	Global Cognitive Score	Memory	Attention	Executive Function	Information Processing Speed
Average Direct Effect	0.055	-0.02	0.15	0.114	-0.164
Average Causal Mediation Effects (composite variable of depression/anxiety)	-0.062	-0.241	0.003	-0.076	0.005
Average Causal Mediation Effects (Fatigue Impact Scale for Daily Use)	-0.351	-0.051	-0.508	-0.302	-0.402
Average Causal Mediation Effects (Central Nervous System-Acting Medications)	-0.019	-0.007	-0.018	-0.014	-0.045
Total Effect	-0.377	-0.318	-0.372	-0.278	-0.606
Mediated proportion	1.146	0.938	1.404	1.409	0.729

Average Direct Effect, Average Causal Mediation Effects, total effect, and mediated proportion of the association between toxic oil syndrome and cognitive performance, considering mediation through three variables: the composite depression/anxiety score, fatigue (Fatigue Impact Scale for Daily Use), and central nervous system-acting medications. Cognitive outcomes include global cognitive score, memory, attention, executive function, and information processing speed. Negative values indicate poorer performance.

The mediation effect of affective variables on information processing speed was comparatively lower (72.9%) than that of other cognitive subscales, with fatigue accounting for most of the effect (66.3% of the total). Notably, the direct influence of TOS was not statistically significant in this model.

4. Discussion

After adjusting for demographics and clinical variables, TOS survivors demonstrated significantly poorer cognitive performance than matched controls. However, these group differences were no longer significant when affective symptoms (depression and anxiety), fatigue, and the use of CNS-acting medications were included in the models.

These findings suggest that persistent cognitive impairment in TOS may be more strongly influenced by long-term psychological and functional sequelae—such as fatigue and mood symptoms—than by overt structural damage detectable through conventional neuroimaging. However, these sequelae may themselves reflect subtle or diffuse alterations in brain connectivity, potentially linked to the original toxic exposure. Consistent with prior studies on the neuropsychiatric aftermath of TOS[29], patients exhibited significantly higher levels of depression, anxiety, and fatigue. Importantly, structural equation modeling showed that affective symptoms, particularly fatigue, mediated the majority of the association between TOS and cognitive outcomes.

The observed cognitive profile—notably deficits in processing speed, executive function, and attention—closely resembles that seen in neurological conditions with subcortical involvement, including multiple sclerosis [31,32], CADASIL [33,34], vascular cognitive impairment [35], and mild

traumatic brain injury [36–38]. Such convergence may reflect common neurobiological mechanisms. Affective symptoms, especially depression and fatigue, have well-documented cognitive consequences across clinical populations [39,40], with depression identified as a risk factor for Alzheimer's disease [41,42] and vascular dementia [41].

In our study, the mediation pattern of processing speed was particularly notable. Although not statistically significant as a direct effect, a large proportion of its variance was explained by affective symptoms—mainly fatigue. This supports a mechanism involving network inefficiency and impaired brain connectivity, both of which have been implicated in mood disorders [43]. Recent research highlights differential cognitive effects of affective symptoms: fatigue predominantly impacts attention and processing speed, as shown by Martin et al. [44] in post-COVID syndrome patients, where reduced alertness was strongly correlated with elevated fatigue. In contrast, depression and anxiety are more consistently associated with episodic and working memory impairments. Kushwaha et al. [45] confirmed this across psychiatric disorders, and Delgado-Alonso et al. [46] demonstrated that COVID-19 patients with high affective burden exhibited both cognitive complaints and measurable deficits in multiple domains. These findings support the notion of distinct, though overlapping, neurobiological pathways mediating the cognitive effects of fatigue, depression, and anxiety.

Neuropathological studies in TOS have revealed central chromatolysis in anterior horn cells and brainstem nuclei—consistent with peripheral nervous system involvement [5,47]. However, alterations in central brain structures (e.g., locus coeruleus, midline raphe, basis pontis, medullary reticular formation, and cuneate nuclei) and the presence of non-necrotizing vasculitis suggest broader CNS involvement. In severe cases, prothrombotic states and hypoxia may have contributed to focal ischemic lesions and anoxic encephalopathy [4,5,47].

Cognitive dysfunction linked to small vessel disease typically manifests as slowed processing speed and executive deficits [35,48], generally attributed to white matter lesions and brain atrophy [28]. However, emerging evidence underscores the importance of gray matter involvement in structural and functional network disruption [49]. Cognitive deficits rarely occur without visible magnetic resonance imaging abnormalities, highlighting the crucial role of both white and gray matter integrity in maintaining cognitive performance [50].

In this context, cognitive dysfunction in TOS may reflect long-term disruption in brain connectivity—a feature shared with CADASIL [51], vascular dementia [52], multiple sclerosis [53], and traumatic brain injury (particularly through diffuse axonal injury) [38]. These conditions frequently involve deep gray matter structures, either through ischemia [48,50,54] or progressive atrophy [32,55].

Altered brain connectivity is also well recognized in depression involving cortical areas such as the dorsolateral and ventromedial prefrontal cortices [56,57], hippocampus [58], anterior cingulate cortex [59], and insula [43] as well as subcortical regions like the amygdala [60], putamen [61], and thalamus/pulvinar [59].

In our analysis, fatigue emerged as the strongest mediator of the relationship between TOS exposure and cognitive performance. Fatigue is highly prevalent in other subcortical disorders like multiple sclerosis, CADASIL, and traumatic brain injury, where it often overlaps with depression in a bidirectional manner [34,40]. Our findings align with prior research showing that fatigue detrimentally affects processing speed [32,37,40], executive function [39], and attention [32].

While the absence of significant group-level cognitive differences after adjusting for affective factors might suggest a psychological origin, it is more plausible that these symptoms reflect downstream consequences of chronic toxic exposure involving neurobiological pathways that disrupt large-scale brain networks. Although postmortem studies in early TOS cases did not show overt white matter pathology [4,5,47], the persistence of cognitive impairment decades later suggests a mechanism beyond focal structural lesions.

This study has limitations. Its cross-sectional design precludes causal inferences about the relationships among affective symptoms, fatigue, and cognition. Longitudinal data are needed to

determine the directionality of these associations. Second, the lack of neuroimaging limits our ability to directly link cognitive outcomes to underlying structural or functional brain changes. Future research incorporating multimodal imaging is needed. Third, multicollinearity among fatigue, anxiety, depression, and health-related quality of life may have biased estimates in domains like attention and processing speed, which are especially sensitive to central fatigue. Although we addressed this via separate models and principal component analysis, residual confounding remains possible. Fourth, some mediation models yielded proportions exceeding 100%, which may reflect suppressor effects, model misspecification, or unmeasured confounders. Finally, the modest sample size may have limited our ability to detect smaller cognitive differences.

In summary, TOS survivors continue to exhibit cognitive deficits more than four decades after exposure, particularly in attention, executive function, and processing speed. These impairments appear to be substantially mediated by chronic fatigue and mood symptoms. However, these mediators may themselves represent downstream consequences of the original toxic exposure, reflecting a complex interplay of neurotoxic, vascular, and neuropsychiatric mechanisms. Emerging evidence supports the hypothesis that long-term disruption in brain connectivity—rather than isolated structural lesions—may underlie enduring cognitive dysfunction in TOS, as observed in other subcortical pathologies. Future research incorporating neuroimaging, eye trackers, larger sample sizes, and longitudinal designs will be essential to disentangle these interconnected effects and clarify their underlying neurobiological substrates.

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