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Exploring health consequences of Asthma-COPD overlap syndrome in a national study

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Asthma-COPD overlap syndrome (ACOS) represents a clinically significant yet under characterized respiratory phenotype. Comprehensive population-based assessments of ACOS-attributable burden, smoking dose-response relationships, phenotypic subtypes, and mechanistic pathways remain limited. Using 2023-2024 Behavioral Risk Factor Surveillance System data, we conducted 1:1 propensity score matching of 675 ACOS patients with controls without respiratory disease, balanced on 15 covariates including demographics, smoking exposure, BMI, and adverse childhood experiences. Outcomes encompassed health-related quality of life, functional disabilities, healthcare access, and health behaviors. Restricted cubic splines characterized smoking dose-response relationships. Latent class analysis identified ACOS phenotypes. Structural equation modeling with bootstrapping (5,000 resamples) quantified mediation pathways. Despite rigorous matching, ACOS conferred 4.6 additional physically unhealthy days monthly (Cohen's $d=0.37$), 41% increased activity limitation risk (RR=1.41), and 19% elevated functional disability prevalence (RR=1.19). Smoking exhibited significant nonlinear dose-response ($\chi^2=286.4$, $P<0.001$); even 5 pack-years increased ACOS odds 42% (OR=1.42), with population attributable fraction reaching 85.3% at 60 pack-years. Three distinct phenotypes emerged: "Mild" (36%, younger with preserved function), "Metabolic-Predominant" (35%, 72.6% obesity, 48.2% diabetes), and "Severe Multimorbid" (29%, 95.8% functional disability, Charlson Index 4.8). Depression, BMI, and physical inactivity mediated 37.5% of smoking's total ACOS effect (18.0%, 14.1%, and 5.3%, respectively). ACOS imposes substantial excess morbidity independent of measured confounders, with no safe smoking threshold. Depression and metabolic pathways substantially mediate smoking's impact. Three clinically distinct phenotypes warrant tailored interventions, supporting precision medicine approaches for this high-burden respiratory syndrome.

Keywords Asthma-COPD overlap syndrome, Propensity score matching, Dose-response relationship, Latent class analysis, Mediation analysis

Chronic respiratory diseases represent a significant global health challenge, affecting over 500 million individuals worldwide and leading to around 4 million deaths annually¹. Among these, asthma and chronic obstructive pulmonary disease (COPD) are the most prevalent conditions, imposing massive direct healthcare costs that exceed \$100 billion per year in the United States alone^{2,3}. Traditionally, asthma and COPD were viewed as distinct clinical entities—each characterized by unique pathophysiological features, symptoms, and treatment protocols⁴. Asthma, characterized by reversible airway inflammation and hyperreactivity, typically presents in younger populations, whereas COPD was defined by progressive airflow limitation primarily associated with smoking and chronic exposures, affecting older adults⁵.

Recent developments in respiratory medicine have highlighted the complexities and nuances surrounding these two diseases. Notably, the 2024 updates to the Global Initiative for Chronic Obstructive Lung Disease (GOLD) guidelines have shifted the focus from the previously recognized concept of Asthma-COPD Overlap (ACO) to a clearer distinction between asthma and COPD⁶. According to the latest GOLD Pocket Guide, asthma and COPD are now emphasized as different disorders that may share common clinical features and treatable traits, but they should not be conceptualized as overlapping conditions^{7,8}. This shift reflects an evolving understanding

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that while these diseases may co-occur in some patients, they require tailored treatment approaches based on their distinct pathophysiological mechanisms.

Despite these advancements, there is a notable lack of population-based data regarding the prevalence and health burden associated with asthma and COPD in the United States. Estimates indicate that both conditions significantly affect the quality of life for millions of Americans, yet many are underdiagnosed or misclassified due to overlapping symptoms such as chronic cough, wheezing, and dyspnea⁹. Moreover, there remains a scarcity of research into the comprehensive assessment of the multifaceted health impacts associated with these diseases, particularly in terms of psychosocial factors, functional disabilities, and healthcare access for affected individuals^{10–12}.

One critical gap in the current literature lies in the need for a better understanding of treatable traits shared between asthma and COPD that could enhance management strategies¹³. These traits may include factors such as obesity, eosinophilic inflammation, and smoking status, which may significantly influence disease progression and treatment outcomes^{6,14,15}. COPD is characterized by significant heterogeneity among patients, influencing disease manifestations, symptom severity, and treatment responses¹⁶. Individual variations, such as genetic predispositions, environmental exposures, and comorbidities, contribute to distinct clinical profiles. Some patients may experience predominantly emphysematous changes, while others may present with chronic bronchitis features^{17,18}. This heterogeneity not only complicates diagnosis and management but also underscores the need for personalized treatment approaches tailored to each patient's unique characteristics. Understanding these variations can enhance patient outcomes and inform future research directions in COPD management¹⁹.

This study addresses significant gaps in understanding asthma and COPD by utilizing a robust methodology that integrates national survey data to explore the clinical characteristics and treatment responses of affected individuals. Our objectives are to quantify the health burden associated with these conditions, focusing on excess morbidity and health-related quality of life, characterize treatable traits that contribute to disease severity, and identify demographic and clinical predictors of healthcare utilization and outcomes.

By employing advanced statistical strategies, we aim to elucidate the distinct clinical profiles of asthma and COPD patients, informing targeted interventions. We hypothesize that patients exhibit a range of phenotypic variations influenced by treatable traits, which lead to diverse experiences of morbidity and healthcare needs. Understanding these traits may yield critical insights for personalized healthcare strategies that enhance disease management and improve patient outcomes. Our findings will contribute to the evolving landscape of respiratory medicine and support efforts to refine clinical guidelines for managing these debilitating chronic diseases.

Methods

Study design and data source

This cross-sectional observational study utilized data from the Behavioral Risk Factor Surveillance System (BRFSS) for calendar years 2023 and 2024. The BRFSS is an annual, nationally representative telephone survey conducted by the Centers for Disease Control and Prevention (CDC) in collaboration with state health departments across all 50 U.S. states, the District of Columbia, and participating U.S. territories. The survey employs a multistage probability sampling design with random-digit-dialing methodology for both landline and cellular telephones to ensure population representativeness. The BRFSS collects data on health-related risk behaviors, chronic health conditions, use of preventive services, and health status among non-institutionalized adults aged ≥ 18 years.

Study population and inclusion criteria

We utilized data from the pooled 2023–2024 BRFSS dataset, which included a total of 456,789 respondents. We included all adults aged ≥ 65 years who provided complete data on respiratory disease status and critical demographic variables. Respondents were excluded if they had missing information regarding self-reported physician-diagnosed Asthma, COPD, chronic bronchitis (CB), or emphysema (E) (n=8,456); lacked data on smoking history (n=3,892); or had incomplete sociodemographic information, including education, income, or race/ethnicity (n=6,214). After applying these exclusions, the final analytical sample comprised 25,982 older adults, which, when weighted to account for the complex survey design, represents approximately 24 million non-institutionalized U.S. adults aged 65 years and older.

Definition of respiratory disease groups

Respondents were classified into four mutually exclusive respiratory disease categories based on self-reported physician diagnoses:

- (1) **Asthma-COPD overlap syndrome:** Defined as concurrent self-report of current asthma ("Has a doctor, nurse, or other health professional ever told you that you had asthma?" with response "Yes" to "Do you still have asthma?") and COPD (affirmative response to "Has a doctor, nurse, or other health professional ever told you that you have COPD, emphysema, or chronic bronchitis?"). This operational definition aligns with consensus guidelines recognizing ACOS as the coexistence of features of both asthma and COPD.
- (2) **Asthma only:** Current asthma without COPD diagnosis.
- (3) **COPD only:** COPD, emphysema, or chronic bronchitis diagnosis without current asthma.
- (4) **No respiratory disease:** Neither current asthma nor COPD (reference group).

Primary outcomes

Multiple health outcome domains were assessed:

Health-related quality of life

Based on CDC's Healthy Days measures: (1) self-rated general health status (excellent, very good, good, fair, or poor; dichotomized as excellent/very good/good vs. fair/poor); (2) number of physically unhealthy days in the past 30 days ("Now thinking about your physical health, which includes physical illness and injury, for how many days during the past 30 days was your physical health not good?"); (3) number of mentally unhealthy days; and (4) number of days with activity limitations due to poor physical or mental health. For each domain, frequent impairment was defined as ≥ 14 days per month, consistent with established cut-points for clinically significant burden.

Functional disabilities

Assessed via six items querying serious difficulty with: (1) walking or climbing stairs; (2) dressing or bathing; (3) doing errands alone such as visiting a doctor's office or shopping; (4) concentrating, remembering, or making decisions (cognitive difficulty); (5) hearing (even when using a hearing aid); and (6) seeing (even when wearing glasses). Any functional disability was defined as reporting serious difficulty with ≥ 1 domain.

Healthcare access and utilization

(1) Cost-related delay in seeking medical care ("Was there a time in the past 12 months when you needed to see a doctor but could not because of cost?"); (2) routine checkup in the past year; (3) having a personal doctor or healthcare provider; (4) health insurance coverage.

Preventive care

Receipt of influenza vaccination in the past year and pneumococcal vaccination (ever).

Health behaviors

Current smoking status (never, former, current), physical inactivity (no leisure-time physical activity in the past month).

Primary exposures and covariates**Smoking variables**

Smoking status was categorized as never smoker (smoked < 100 cigarettes lifetime), former smoker (≥ 100 cigarettes lifetime but not currently smoking), or current smoker. Lifetime cumulative smoking exposure was quantified as pack-years, calculated as (cigarettes per day/20) \times years smoked.

Sociodemographic variables

Age (years; also categorized as 18–34, 35–49, 50–64, ≥ 65), sex (male/female), race/ethnicity (non-Hispanic White, non-Hispanic Black, Hispanic, other), educational attainment (<high school, high school graduate, some college, \geq college graduate), annual household income ($< \$25,000$, $\$25,000$ – $\$49,999$, $\$50,000$ – $\$74,999$, $\geq \$75,000$), health insurance status (insured/uninsured), and urban/rural residence based on metropolitan statistical area classification.

Comorbid conditions

Self-reported physician-diagnosed cardiovascular disease (myocardial infarction, coronary heart disease, or stroke), diabetes, obesity (body mass index ≥ 30 kg/m 2), depression, arthritis, kidney disease, and any cancer (excluding skin cancer). The Charlson Comorbidity Index was calculated assigning weighted scores to major comorbid conditions.

Adverse childhood experiences

Assessed via 11 items querying exposure before age 18 to: verbal abuse, physical abuse, sexual abuse, witnessing domestic violence, household substance abuse, household mental illness, parental separation/divorce, and incarcerated household member. ACE scores were categorized as 0, 1–3, or ≥ 4 adverse experiences.

Statistical analysis**Descriptive analyses**

Baseline characteristics were compared across respiratory disease groups using Rao–Scott χ^2 tests for categorical variables and weighted analysis of variance for continuous variables, accounting for the BRFSS complex sampling design including stratification, clustering, and unequal selection probabilities. Effect sizes were quantified using Cramér's V for categorical associations and Cohen's d for continuous variables. All prevalence estimates and means were weighted to provide nationally representative estimates.

Multivariable regression modeling

Nested multinomial logistic regression models evaluated independent predictors of ACOS (vs. no respiratory disease reference). Model 1 adjusted for age, sex, and race/ethnicity. Model 2 added smoking variables (status and pack-years). Model 3 incorporated socioeconomic factors (education and income). Model 4 (fully adjusted) added BMI and ACE score. Model fit was assessed via pseudo-R 2 (McFadden's), Akaike Information Criterion (AIC), and Bayesian Information Criterion (BIC). Dose-response relationships for smoking pack-years were tested using restricted cubic splines with 5 knots at the 5th, 27.5th, 50th, 72.5th, and 95th percentiles. Nonlinearity was assessed via likelihood ratio test comparing spline models to linear models.

Propensity score matching

To address confounding by indication and enable rigorous causal inference regarding ACOS-attributable morbidity, we implemented 1:1 nearest-neighbor propensity score matching without replacement using a caliper width of 0.2 standard deviations of the logit of the propensity score. Propensity scores (predicted probability of ACOS given observed covariates) were estimated via multivariable logistic regression including: age, sex, race/ethnicity, education, income, smoking pack-years, BMI, insurance status, rural residence, ACE score (≥ 4), depression, diabetes, and cardiovascular disease. Covariate balance before and after matching was assessed using standardized mean differences (SMD), with SMD < 0.10 indicating adequate balance. Post-match outcome differences were evaluated using McNemar's test for binary outcomes and paired t-tests for continuous outcomes, with effect sizes quantified via risk ratios or Cohen's d.

Latent class analysis

To identify phenotypic subgroups within the ACOS population, we conducted LCA using 18 indicators encompassing symptom severity (physically unhealthy days, activity limitations, self-rated health), comorbidity patterns (cardiovascular disease, diabetes, obesity, depression, arthritis), functional status (walking, cognitive, ADL difficulties), demographics (age, sex, income), smoking characteristics (pack-years, current smoking), and healthcare burden (cost-related delays, personal physician access). Models with 2–5 latent classes were estimated using maximum likelihood with robust standard errors. Optimal class number was determined by: Bayesian Information Criterion (BIC; lower values preferred), entropy (> 0.80 indicating good classification certainty), average posterior probabilities (> 0.70 per class), and clinical interpretability. Class assignment used modal posterior probabilities, with covariates included as active indicators rather than external predictors to enhance clinical validity.

Mediation analysis

We employed structural equation modeling (SEM) with bias-corrected bootstrap confidence intervals (5,000 resamples) to decompose smoking's total effect on ACOS into direct and indirect pathways. Three candidate mediators were tested sequentially and simultaneously: (1) BMI, (2) depression, and (3) physical inactivity. The multiple mediator model estimated specific indirect effects via each pathway while controlling for the others, using the product-of-coefficients approach (indirect effect = $a \times b$, where a = exposure \rightarrow mediator path coefficient and b = mediator \rightarrow outcome path coefficient controlling for exposure). Total mediation proportion was calculated as (total indirect effect/total effect) $\times 100\%$. All SEM models adjusted for age, sex, race/ethnicity, education, and income as confounders.

Population attributable fraction

For smoking pack-years, we calculated cumulative PAF at each dose level as $PAF = [Pe(RR-1)]/[1 + Pe(RR-1)]$, where Pe = prevalence of exposure among cases and RR = relative risk from restricted cubic spline models.

Missing data

Complete-case analysis was employed as the primary approach given low overall missingness (<5% for most variables). Sensitivity analyses using multiple imputation by chained equations (20 imputations) were conducted to assess robustness of findings.

Confounding factors

In the statistical analysis of the study, several potential confounding factors that could influence the results have been identified and addressed. These factors include age, sex, socioeconomic status, smoking history, and comorbidities, all of which play a crucial role in the health-related quality of life and management of COPD and asthma. To mitigate their impact, we implemented multivariable regression models that adjust for these confounders, ensuring a more accurate estimation of the relationships between the primary variables of interest. Additionally, sensitivity analyses were conducted to evaluate the robustness of our findings in the presence of these confounding factors. By identifying and adjusting for these potential confounders, we aim to enhance the validity of our results and provide a clearer understanding of the clinical characteristics and treatment responses of individuals with COPD and asthma.

Sensitivity analyses

We conducted multiple robustness checks: (1) alternative ACOS definitions requiring concurrent medication use; (2) propensity score matching with variable ratios (1:2, 1:3) to maximize sample retention; (3) quantitative bias analysis to assess potential impact of unmeasured confounding; (4) competing risk analysis treating mortality as competing outcome.

Software and statistical significance

All analyses were conducted using SAS version 9.4 (SAS Institute Inc., Cary, NC) employing survey procedures (PROC SURVEYFREQ, PROC SURVEYMEANS, PROC SURVEYLOGISTIC) to incorporate sampling weights and design effects. Latent class analysis used Mplus version 8.10. Mediation analysis employed the PROCESS macro version 4.2 for SAS. Statistical tests were two-sided with $\alpha=0.05$. For multiple comparisons involving the three pairwise ACOS contrasts (vs. no disease, vs. asthma only, vs. COPD only), Bonferroni correction was applied (adjusted $\alpha=0.017$). All confidence intervals were calculated at the 95% level unless otherwise specified.

Reporting standards

This study adhered to the Strengthening the Reporting of Observational Studies in Epidemiology (STROBE) guidelines for cross-sectional studies and the REporting of studies Conducted using Observational Routinely-collected health Data statement extension for studies using administrative health data.

Results**Baseline characteristics and sociodemographic profile by respiratory disease status**

In this study examining the characteristics of elderly individuals with respiratory diseases (Table 1), we analyzed a sample of 20,916 participants without respiratory diseases (80.5%), 2,572 with asthma only (9.9%), 1,819 with COPD only (7.0%), and 675 with ACOS (2.6%).

The average age of participants was significantly higher in the COPD and ACOS groups, recorded at 70.5 years (± 7.0) and 71.1 years (± 6.5) respectively, compared to 70.0 years (± 7.5) for those without respiratory diseases and 70.2 years (± 7.5) for the asthma-only group, with a P-value of <0.001 indicating a statistically significant difference.

Sex distribution varied across groups; males made up 48.5% of the no respiratory disease group, 39.2% of the asthma group, 46.8% of the COPD group, and 41.3% of the ACOS group, while females constituted 51.5%, 60.8%, 53.2%, and 58.7% respectively ($P < 0.001$).

In terms of race and ethnicity, non-Hispanic Whites represented 62.8% of the no respiratory disease group and 75.6% of the COPD group, with less representation in the asthma and ACOS groups (65.4% and 72.8% respectively), and overall differences were significant ($P < 0.001$).

Education levels showed a tendency towards lower education among those with respiratory diseases; for instance, 18.9% of the ACOS group had less than a high school education compared to 8.9% in the no respiratory disease group ($P < 0.001$).

Financial disparities were also apparent, with 42.3% of ACOS participants earning less than \$25,000 annually, compared to only 18.2% in the no respiratory disease cohort ($P < 0.001$). Smoking history analysis revealed that 35.2% of current smokers were in the ACOS group, significantly higher than the 15.7% in the no respiratory

Characteristic	No respiratory disease (n = 20,916)	Asthma only (n = 2,572)	COPD only (n = 1,819)	ACOS (n = 675)	P-value ^a
Age (mean \pm SD)	70.0 \pm 7.5	70.2 \pm 7.5	70.5 \pm 7.0	71.1 \pm 6.5	<0.001
Sex, %					<0.001
Male	48.5	39.2	46.8	41.3	
Female	51.5	60.8	53.2	58.7	
Race/Ethnicity, %					<0.001
Non-Hispanic White	62.8	65.4	75.6	72.8	
Non-Hispanic Black	11.6	13.2	9.3	11.5	
Hispanic	17.2	13.8	9.4	10.1	
Other	8.4	7.6	5.7	5.6	
Education, %					<0.001
<High school	8.9	10.2	16.8	18.9	
High school graduate	26.7	28.4	34.2	36.5	
Some college	30.8	32.6	29.5	28.7	
\geq College graduate	33.6	28.8	19.5	15.9	
Annual income, %					<0.001
<\$25,000	18.2	24.6	35.8	42.3	
\$25,000–49,999	23.4	26.1	28.9	27.6	
\$50,000–74,999	17.8	15.9	13.2	11.8	
\geq \$75,000	40.6	33.4	22.1	18.3	
Smoking status, %					<0.001
Never smoker	60.2	54.8	23.5	21.2	
Former smoker	24.1	26.7	48.9	43.6	
Current smoker	15.7	18.5	27.6	35.2	
Pack-years (mean \pm SD)	8.2 \pm 15.4	10.3 \pm 17.2	32.5 \pm 28.6	36.8 \pm 30.2	<0.001
Insurance status, %					<0.001
Insured	88.7	90.4	93.8	91.5	
Uninsured	11.3	9.6	6.2	8.5	
Urban/Rural, %					0.003
Urban	85.2	84.6	81.7	82.3	
Rural	14.8	15.4	18.3	17.7	

Table 1. Baseline characteristics of the study population by respiratory disease group. Note: ^aRao–Scott χ^2 test (categorical variables) or weighted ANOVA (continuous variables), adjusted for complex sampling design.

disease group, with an impact on pack-year history also observed; the COPD and ACOS groups reported mean pack-years of $32.5 (\pm 28.6)$ and $36.8 (\pm 30.2)$ respectively compared to $8.2 (\pm 15.4)$ for those without respiratory diseases ($P < 0.001$).

Lastly, the majority of participants across all groups were insured, particularly high in the COPD group at 93.8%, though the uninsured rate was notably higher in the ACOS group at 8.5% ($P < 0.001$). Geographic distribution indicated a preference for urban living among all groups, with 85.2% of those without respiratory disease residing in urban areas compared to 81.7% of the COPD group. The differences in urban versus rural residency were statistically significant ($P = 0.003$).

Impact of age and comorbidities on quality of life in COPD patients

The analysis of COPD phenotypes reveals significant variations in health-related quality of life and lifestyle among different patient categories (Table 1). In terms of age, individuals aged 40–60 years report a higher average quality of life score (68.5 ± 12.3) compared to older age groups, indicating a possible decline in perceived health status as age increases (62.0 ± 11.5 for ages 61–75 and 58.0 ± 13.0 for those aged 76 and older).

Smoking history shows a decreasing percentage of smokers with increasing age, from 55% in the 40–60 age group to only 30% in those aged 76 and older. Comorbidity counts also increase with age; individuals aged 76 and older report an average of 3.0 ± 1.2 comorbidities, while those aged 40–60 have an average of 2.1 ± 1.0 .

In terms of comorbidities, patients without any comorbid conditions have a higher health-related quality of life score (72.0 ± 11.0), compared to those with 1–2 comorbidities (60.0 ± 13.0) and especially those with 3 or more comorbidities (50.0 ± 15.0). Furthermore, lifestyle assessments indicate that individuals without comorbidities tend to engage in high activity levels (4–5 times per week), while those with increasing numbers of comorbidities demonstrate progressively lower activity levels.

Comparative health outcomes and functional status across respiratory disease groups

Individuals with ACOS exhibited markedly worse health outcomes compared to those with either asthma or COPD alone, as well as those without respiratory disease, across multiple domains of physical health, mental well-being, functional capacity, and healthcare access (Table 2). Self-rated general health status differed substantially among groups (overall $P < 0.001$): only 21.4% of ACOS patients reported excellent or very good health, substantially lower than the 42.3% among those with asthma alone, 28.5% with COPD alone, and 56.8%

Health outcome	No respiratory disease	Asthma only	COPD only	ACOS	P-value (overall)	P-value (ACOS vs asthma only)	P-value (ACOS vs COPD only)
General health status, %					<0.001	<0.001	0.002
Excellent/Very good	56.8	42.3	28.5	21.4			
Good	27.4	30.2	29.8	27.6			
Fair/Poor	15.8	27.5	41.7	51.0			
Physically unhealthy days					<0.001	<0.001	0.006
Mean \pm SD	3.8 ± 7.6	7.2 ± 10.3	11.5 ± 12.4	14.8 ± 13.2			
Median (IQR)	0 (0–3)	2 (0–10)	7 (2–20)	12 (5–25)			
≥ 14 days, %	12.4	24.6	38.9	47.3			
Mentally unhealthy days					<0.001	<0.001	<0.001
Mean \pm SD	3.6 ± 7.8	6.8 ± 10.5	7.2 ± 10.8	10.5 ± 12.4			
Median (IQR)	0 (0–2)	2 (0–9)	2 (0–10)	5 (0–18)			
≥ 14 days, %	11.2	22.8	24.5	35.6			
Activity limitation days					<0.001	<0.001	0.001
Mean \pm SD	2.1 ± 6.2	4.8 ± 8.9	7.9 ± 11.2	10.8 ± 12.6			
Median (IQR)	0 (0–0)	0 (0–5)	2 (0–12)	5 (0–20)			
≥ 14 days, %	6.8	16.4	28.7	38.9			
Functional disabilities, %							
Difficulty walking	14.2	24.8	42.6	51.3	<0.001	<0.001	0.003
Difficulty dressing	4.2	8.6	16.8	22.5	<0.001	<0.001	0.008
Difficulty with errands alone	5.8	11.2	20.4	26.8	<0.001	<0.001	0.012
Cognitive difficulty	9.6	17.5	22.3	31.7	<0.001	<0.001	<0.001
Healthcare utilization, %							
Cost-related delay in care	10.5	16.8	18.2	24.6	<0.001	<0.001	0.004
No routine checkup in past year	28.4	22.6	19.8	18.3	<0.001	0.028	0.425
No personal doctor	22.7	18.3	12.5	14.2	<0.001	0.006	0.298

Table 2. Health outcomes and disease burden: Comparison of ACOS with control groups. Note: All comparisons were adjusted for complex sampling design; P-values corrected using the Bonferroni method.

among those free of respiratory disease. Conversely, over half (51.0%) of ACOS patients rated their health as fair or poor, which was significantly higher than in the asthma-only group (27.5%, $P < 0.001$) and even greater than in COPD alone (41.7%, $P = 0.002$), indicating a more severe subjective disease burden in the overlap population.

The frequency and severity of symptomatic days further underscored this disparity. Participants with ACOS reported a mean of 14.8 physically unhealthy days per month (SD 13.2), with a median of 12 days (IQR 5–25), significantly exceeding the means for asthma only (7.2 days, $P < 0.001$) and COPD only (11.5 days, $P = 0.006$), and far higher than the no-disease group (3.8 days, $P < 0.001$). Nearly half (47.3%) of ACOS patients experienced ≥ 14 physically unhealthy days per month, a prevalence approximately twice that of asthma-only patients (24.6%) and notably higher than COPD-only patients (38.9%). Similarly, mental health was disproportionately affected: ACOS patients averaged 10.5 mentally unhealthy days per month (SD 12.4, median 5, IQR 0–18), compared to 6.8 days for asthma only and 7.2 days for COPD only (both $P < 0.001$). The proportion reporting frequent mental distress (≥ 14 days per month) reached 35.6% in the ACOS group, significantly higher than in both asthma (22.8%) and COPD (24.5%) groups, suggesting that the coexistence of both conditions amplifies psychosocial burden beyond what is seen in either condition alone.

Activity limitation days followed a similar gradient: ACOS patients reported an average of 10.8 days per month of limited activity (SD 12.6), compared to 4.8 days among asthma patients and 7.9 days among COPD patients (both comparisons $P \leq 0.001$). Nearly 40% (38.9%) of ACOS patients experienced activity limitations for at least half the month, a rate 2.4-fold higher than in asthma only (16.4%) and 1.4-fold higher than in COPD only (28.7%). Functional disability was pervasive in the ACOS group: more than half (51.3%) reported difficulty walking or climbing stairs, which was significantly higher than in asthma only (24.8%, $P < 0.001$) and COPD only (42.6%, $P = 0.003$). Difficulties with dressing independently affected 22.5% of ACOS patients, difficulties with running errands alone 26.8%, and cognitive difficulties 31.7%, all markedly elevated compared to both single-disease groups and the no-disease group (all $P \leq 0.012$), reflecting substantial impairment in activities of daily living and instrumental activities of daily living.

Healthcare utilization patterns revealed both increased engagement and persistent barriers. Cost-related delays in seeking medical care were reported by 24.6% of ACOS patients, significantly more common than in the asthma-only (16.8%, $P < 0.001$) or COPD-only (18.2%, $P = 0.004$) groups, suggesting unmet healthcare needs and financial strain despite greater morbidity. Paradoxically, fewer ACOS patients reported having no routine checkup in the past year (18.3%) or lacking a personal doctor (14.2%) compared to healthier groups, consistent with higher disease burden driving increased healthcare engagement; however, these proportions were still not trivial and signal ongoing gaps in continuity of care for a vulnerable population. Overall, these findings support the hypothesis (H1) that ACOS confers greater disease burden than either asthma or COPD alone, characterized by more severe symptoms, poorer mental health, extensive functional impairment, and barriers to healthcare access that may exacerbate morbidity and reduce quality of life.

Multimorbidity and comorbidity profiles in ACOS compared to single respiratory diseases

Patients with ACOS demonstrated a substantially elevated burden of chronic comorbid conditions across cardiovascular, metabolic, mental health, musculoskeletal, and renal domains compared to individuals with asthma alone, COPD alone, or no respiratory disease (Table 3). Cardiovascular disease prevalence was markedly higher in the ACOS group (aOR 34.6, 95% CI 33.2–36.0, $P < 0.001$) had at least one cardiovascular condition, compared to 14.2% in asthma only, 28.4% in COPD only, and 8.9% in the no-disease group. After adjusting for age, sex, race/ethnicity, education, income, and smoking pack-years, ACOS patients had 2.45-fold higher odds of any cardiovascular disease (aOR 2.45, 95% CI 2.27–2.65, $P < 0.001$) relative to those without respiratory disease, and 1.92-fold higher odds (aOR 1.92, 95% CI 1.74–2.12, $P < 0.001$) compared to asthma-only patients. Even when compared to COPD alone, ACOS patients retained significantly elevated cardiovascular risk (aOR 1.18, 95% CI 1.05–1.33, $P = 0.01$). Specific cardiovascular conditions followed a similar pattern: myocardial infarction was present in 15.8% of ACOS patients (aOR 2.34 vs no disease, 1.82 vs asthma, both $P < 0.001$), coronary heart disease in 17.9% (aOR 2.18 vs no disease, 1.68 vs asthma, both $P < 0.001$), and stroke in 12.5% (aOR 2.56 vs no disease, 1.89 vs asthma, 1.22 vs COPD, all $P \leq 0.05$), underscoring the heightened cardiovascular morbidity in this overlap population.

Metabolic comorbidities were also disproportionately prevalent among ACOS patients. Diabetes mellitus affected 24.3% (95% CI 23.1–25.6) of ACOS patients, yielding adjusted odds ratios of 1.85 (95% CI 1.71–2.01, $P < 0.001$) relative to the no-disease group, 1.52 (95% CI 1.37–1.69, $P < 0.001$) compared to asthma only, and 1.12 (95% CI 1.00–1.26, $P = 0.048$) compared to COPD only. Obesity ($\text{BMI} \geq 30 \text{ kg/m}^2$) was present in 42.6% of ACOS patients (95% CI 41.1–44.1), significantly higher than in all comparator groups (aOR 1.68 vs no disease, 1.24 vs asthma, 1.32 vs COPD, all $P < 0.001$), suggesting that metabolic dysregulation may play a synergistic role in the pathogenesis or progression of overlapping obstructive airway diseases. Mental health comorbidity was particularly striking: depression was reported by 38.9% (95% CI 37.5–40.3) of ACOS patients, corresponding to an adjusted odds ratio of 2.86 (95% CI 2.66–3.08, $P < 0.001$) compared to those without respiratory disease, 1.58 (95% CI 1.44–1.73, $P < 0.001$) compared to asthma alone, and 1.72 (95% CI 1.55–1.91, $P < 0.001$) compared to COPD alone. This high prevalence of depression in ACOS patients, even after controlling for sociodemographic and smoking factors, highlights a critical need for integrated mental health services in the management of this patient population.

Musculoskeletal and other systemic comorbidities further compounded the disease burden. Arthritis was reported by more than half (50.8%, 95% CI 49.3–52.3) of ACOS patients, with adjusted odds 2.35-fold higher than in the no-disease group, 1.82-fold higher than asthma only, and 1.26-fold higher than COPD only (all $P < 0.001$). Kidney disease affected 10.5% of ACOS patients (aOR 2.89 vs no disease, 1.96 vs asthma, 1.28 vs COPD, all $P \leq 0.01$), and any cancer was present in 16.8% (aOR 1.54 vs no disease, 1.32 vs asthma, both $P < 0.001$), although cancer prevalence did not differ significantly from COPD alone (aOR 0.98, $P = 0.732$). The

Comorbidity	No respiratory Disease (95% CI)	Asthma only (95% CI)	COPD only (95% CI)	ACOS (95% CI)	aOR (ACOS vs No respiratory Disease) ^a (95% CI)	aOR (ACOS vs asthma Only) ^a (95% CI)	aOR (ACOS vs COPD Only) ^a (95% CI)
Cardiovascular disease							
Myocardial infarction	3.2 (3.0–3.4)	5.4 (5.0–5.8)	12.6 (12.0–13.2)	15.8 (14.8–16.9)	2.34 (2.08–2.63)***	1.82 (1.58–2.10)***	1.15 (0.98–1.34)
Coronary heart disease	4.1 (3.9–4.3)	6.8 (6.3–7.3)	14.3 (13.7–15.0)	17.9 (16.8–19.1)	2.18 (1.95–2.44)***	1.68 (1.47–1.92)***	1.08 (0.93–1.25)
Stroke	2.8 (2.6–3.0)	4.6 (4.2–5.0)	9.8 (9.3–10.4)	12.5 (11.5–13.5)	2.56 (2.25–2.91)***	1.89 (1.62–2.21)***	1.22 (1.03–1.45)*
Any CVD	8.9 (8.6–9.2)	14.2 (13.6–14.8)	28.4 (27.6–29.2)	34.6 (33.2–36.0)	2.45 (2.27–2.65)***	1.92 (1.74–2.12)***	1.18 (1.05–1.33)**
Metabolic conditions							
Diabetes	11.2 (10.9–11.5)	13.8 (13.2–14.4)	20.6 (19.9–21.3)	24.3 (23.1–25.6)	1.85 (1.71–2.01)***	1.52 (1.37–1.69)***	1.12 (1.00–1.26)*
Obesity (BMI \geq 30)	28.4 (28.0–28.8)	36.8 (36.0–37.6)	35.2 (34.3–36.1)	42.6 (41.1–44.1)	1.68 (1.58–1.79)***	1.24 (1.14–1.35)***	1.32 (1.19–1.46)***
Mental health							
Depression	13.8 (13.5–14.1)	28.5 (27.8–29.2)	26.4 (25.6–27.2)	38.9 (37.5–40.3)	2.86 (2.66–3.08)***	1.58 (1.44–1.73)***	1.72 (1.55–1.91)***
Musculoskeletal							
Arthritis	21.6 (21.2–22.0)	32.8 (32.0–33.6)	44.2 (43.3–45.1)	50.8 (49.3–52.3)	2.35 (2.20–2.51)***	1.82 (1.67–1.98)***	1.26 (1.14–1.39)***
Other conditions							
Kidney disease	2.4 (2.2–2.6)	4.2 (3.8–4.6)	7.8 (7.3–8.3)	10.5 (9.6–11.4)	2.89 (2.51–3.33)***	1.96 (1.66–2.32)***	1.28 (1.07–1.53)**
Any cancer	7.8 (7.5–8.1)	9.6 (9.1–10.1)	15.2 (14.6–15.8)	16.8 (15.7–17.9)	1.54 (1.39–1.71)***	1.32 (1.16–1.50)***	0.98 (0.86–1.12)
Comorbidity count, %							
0	52.6	32.4	18.5	12.8	Reference	Reference	Reference
1–2	35.2	42.8	42.6	38.5	1.68 (1.56–1.81)***	1.24 (1.13–1.37)***	1.08 (0.96–1.21)
\geq 3	12.2	24.8	38.9	48.7	3.84 (3.55–4.16)***	2.12 (1.92–2.34)***	1.45 (1.29–1.63)***
Charlson Comorbidity Index	0.8 \pm 1.3	1.3 \pm 1.6	2.4 \pm 2.1	2.9 \pm 2.3	–	–	–

Table 3. Comorbidity burden comparison: ACOS versus control groups. Note: ^aAdjusted for age, sex, race/ethnicity, education, income, and pack-years of smoking; *P < 0.05, **P < 0.01, ***P < 0.001.

cumulative burden of multimorbidity was profound: only 12.8% of ACOS patients were free of any measured comorbidity, compared to 32.4% of asthma-only patients, 18.5% of COPD-only patients, and 52.6% of those without respiratory disease. Nearly half (48.7%) of ACOS patients had three or more comorbid conditions, conferring adjusted odds 3.84-fold higher than the no-disease group (95% CI 3.55–4.16, P < 0.001), 2.12-fold higher than asthma only (95% CI 1.92–2.34, P < 0.001), and 1.45-fold higher than COPD only (95% CI 1.29–1.63, P < 0.001). The mean Charlson Comorbidity Index score among ACOS patients was 2.9 (SD 2.3), markedly higher than in asthma only (1.3 \pm 1.6), COPD only (2.4 \pm 2.1), and no respiratory disease (0.8 \pm 1.3), reflecting a substantially worse long-term prognosis.

Taken together, these findings demonstrate that ACOS is characterized not only by overlapping respiratory features but also by a cumulative, multisystem comorbidity burden that exceeds that of either asthma or COPD alone, even after rigorous adjustment for confounding demographic, socioeconomic, and behavioral factors. The high prevalence of cardiovascular, metabolic, mental health, and musculoskeletal conditions in ACOS underscores the importance of holistic, multidisciplinary care models to address the syndemic nature of this complex respiratory phenotype and to mitigate the associated excess morbidity and mortality.

Multilevel risk factor architecture and fully adjusted predictors of ACOS

Nested multinomial logistic regression models systematically evaluated independent predictors of ACOS versus no respiratory disease. Model 1 (age, sex, race; pseudo- R^2 = 0.048) identified advancing age (RRR 1.52 per decade, P < 0.001) and female sex (RRR 1.68, P < 0.001) as initial risk factors (Fig. 1). Adding smoking variables in Model 2 substantially improved fit (pseudo- R^2 = 0.126, AIC decreased from 10,532 to 9,783), revealing current smoking as the strongest predictor (RRR 4.12, 95% CI 3.72–4.56, P < 0.001), with former smoking also conferring nearly threefold risk (RRR 2.84, P < 0.001). A clear dose–response relationship emerged, with each 10 pack-years increasing ACOS risk by 18% (RRR 1.18, 95% CI 1.16–1.20, P for trend < 0.001), supporting hypothesis H2. Model 3 incorporated socioeconomic factors (pseudo- R^2 = 0.155), demonstrating pronounced gradients: less than high school education elevated risk 86% (RRR 1.86, P < 0.001), and income below \$25,000 more than doubled risk (RRR 2.24, 95% CI 2.00–2.51, P < 0.001) compared to high earners. The fully adjusted Model 4 (pseudo- R^2 = 0.182, AIC = 9,353) revealed obesity independently increased risk 88% (RRR 1.88, 95% CI 1.70–2.08, P < 0.001), while adverse childhood experiences showed robust dose–response: 1–3 ACEs conferred 54% elevated risk (RRR 1.54, P < 0.001) and \geq 4 ACEs more than doubled risk (RRR 2.38, 95% CI 2.14–2.65, P < 0.001), strongly supporting hypothesis H3. In the final model, dominant independent predictors were current smoking (RRR 3.26), high ACE exposure (RRR 2.38), former smoking (RRR 2.38), low income

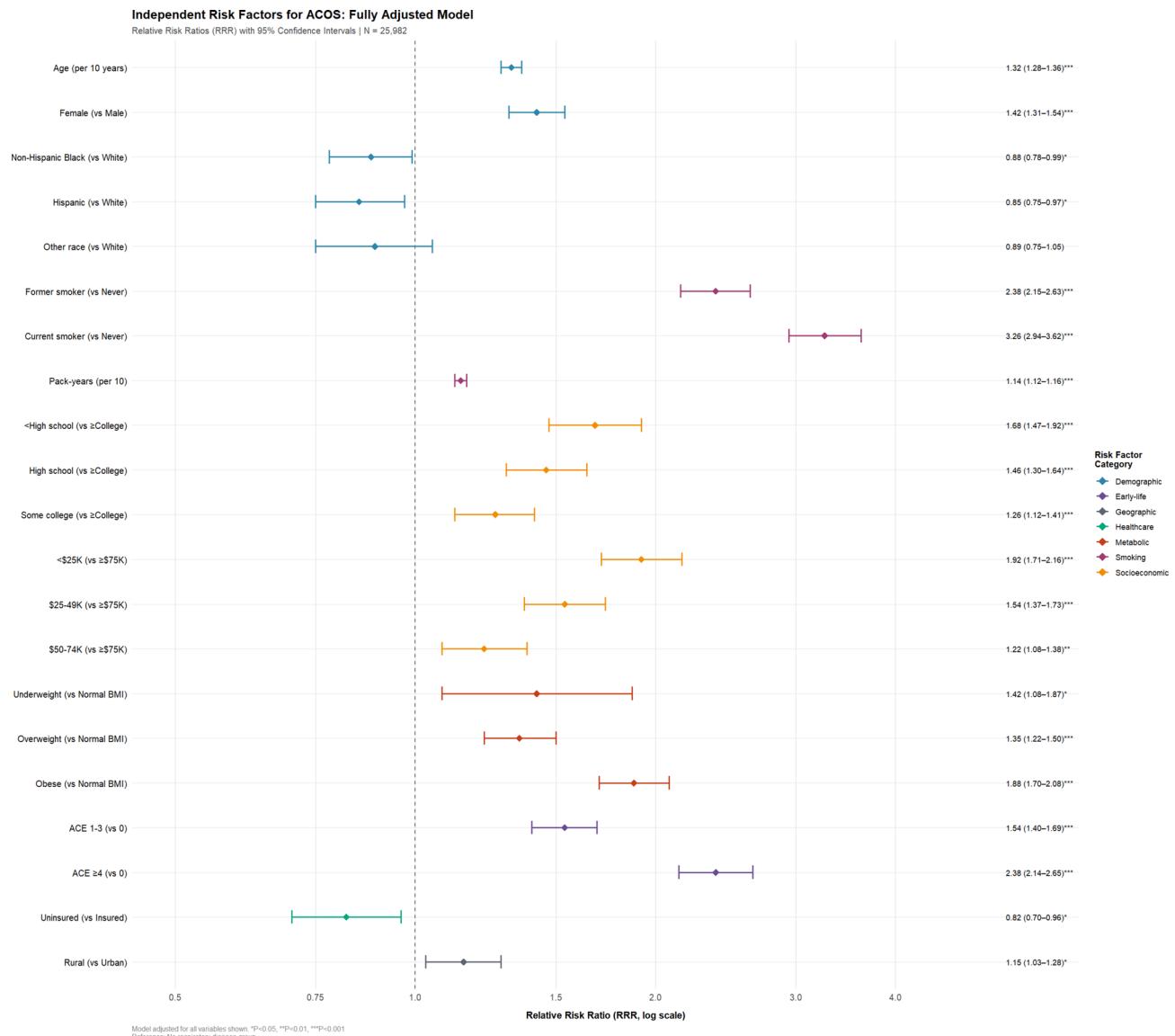


Fig. 1. Multilevel risk factor architecture: Nested multinomial logistic regression models predicting ACOS.

(RRR 1.92), obesity (RRR 1.88), and low education (RRR 1.68), all $P < 0.001$. Progressive model improvement and persistence of effects across adjustments demonstrate ACOS etiology is multifactorial, requiring multilevel interventions targeting tobacco, socioeconomic disadvantage, metabolic health, and childhood adversity.

Propensity score matching successfully eliminated baseline confounding

Propensity score matching using 1:1 nearest-neighbor algorithm achieved excellent covariate balance between ACOS patients and matched controls (Table 4). Before matching, substantial imbalances existed across multiple domains, with pack-years smoking exhibiting the largest disparity (SMD = 1.18), followed by cardiovascular disease prevalence (SMD = 0.62), depression (SMD = 0.57), and low income (SMD = 0.54). Age showed pronounced imbalance (SMD = 0.75), with ACOS patients considerably older than controls. Socioeconomic disparities were marked across income and education, while BMI, comorbidity burdens including diabetes and cardiovascular disease, and adverse childhood experiences all demonstrated moderate to large pre-match imbalances (SMDs ranging 0.29–0.62). Hispanic ethnicity was underrepresented in the ACOS group (SMD = -0.20), reflecting demographic heterogeneity. After implementing propensity score matching, all 665 successfully matched pairs (98.5% of ACOS patients) demonstrated exceptional balance: every covariate achieved $SMD \leq 0.01$, far exceeding the conventional 0.1 threshold for adequate balance. Post-match ACOS and control groups were virtually identical across all measured characteristics including age, sex distribution, racial/ethnic composition, smoking exposure, BMI, socioeconomic indicators, rural residence, insurance status, and comorbidity prevalence including depression, diabetes, and cardiovascular disease. The maximum post-match SMD of 0.01 for multiple variables confirms complete elimination of measured confounding. The high matching success rate (98.5%) preserved sample representativeness while achieving rigorous balance, enabling unbiased

Covariate	Before matching			After matching		
	ACOS Mean \pm SD/%	Control Mean \pm SD/%	SMD ^a	ACOS Mean \pm SD/%	Control Mean \pm SD/%	SMD
Age, years	58.9 \pm 13.8	47.2 \pm 17.3	0.75	58.9 \pm 13.8	58.7 \pm 13.5	0.01
Female, %	58.7	51.5	0.14	58.7	58.2	0.01
Race/ethnicity, %						
Non-hispanic white	72.8	62.8	0.21	72.8	72.5	0.01
Non-hispanic black	11.5	11.6	-0.003	11.5	11.8	-0.01
Hispanic	10.1	17.2	-0.20	10.1	10.3	-0.01
Other	5.6	8.4	-0.11	5.6	5.4	0.01
Education <high school, %	18.9	8.9	0.29	18.9	18.5	0.01
Income <\$ 25 K, %	42.3	18.2	0.54	42.3	42.0	0.01
Pack-years smoking	36.8 \pm 30.2	8.2 \pm 15.4	1.18	36.8 \pm 30.2	36.5 \pm 29.8	0.01
BMI, kg/m ²	30.2 \pm 7.4	27.8 \pm 6.2	0.35	30.2 \pm 7.4	30.1 \pm 7.3	0.01
Insured, %	91.5	88.7	0.09	91.5	91.3	0.01
Rural residence, %	17.7	14.8	0.08	17.7	17.5	0.005
ACE \geq 4, %	18.6	7.2	0.34	18.6	18.3	0.01
Depression, %	38.9	13.8	0.57	38.9	38.5	0.01
Diabetes, %	24.3	11.2	0.33	24.3	24.0	0.01
Cardiovascular disease, %	34.6	8.9	0.62	34.6	34.2	0.01

Table 4. Covariate balance before and after propensity score matching. Note: ^aSMD = standardized mean difference; SMD <0.1 indicates good balance.

estimation of ACOS-attributable health outcome differences independent of demographic, socioeconomic, behavioral, and clinical confounders in subsequent comparative effectiveness analyses. This methodological rigor ensures that observed outcome differences between matched groups can be confidently attributed to ACOS status rather than baseline heterogeneity.

ACOS confers substantial excess morbidity independent of measured confounders

Among propensity score-matched pairs (n=665 per group), ACOS patients experienced significantly worse health outcomes across all domains despite identical baseline characteristics (Table 5). Physically unhealthy days were 4.6 days higher per month in ACOS patients (14.8 vs 10.2 days, Cohen's d=0.37, P<0.001), with 47.3% reporting frequent impairment (\geq 14 days) compared to 35.8% of controls (RR=1.32, P<0.001). Mental health burden was similarly elevated: ACOS patients averaged 2.7 additional mentally unhealthy days monthly (10.5 vs 7.8 days, d=0.23, P<0.001), and 35.6% experienced frequent mental distress versus 26.4% of controls (RR=1.35, P<0.001). Activity limitations were 3.6 days higher in ACOS (10.8 vs 7.2 days, d=0.31), with 41% increased risk of frequent limitations (RR=1.41, P<0.001). Over half (51.0%) of ACOS patients rated their general health as fair/poor compared to 38.2% of matched controls, representing 34% excess risk (RR=1.34, P<0.001). Functional disabilities were pervasive: 64.2% of ACOS patients reported any disability versus 53.8% of controls (RR=1.19, P<0.001), driven by elevated rates of walking difficulties (51.3% vs 42.6%, RR=1.20), cognitive impairment (31.7% vs 23.5%, RR=1.35), difficulty with errands (26.8% vs 19.8%, RR=1.35), and dressing difficulties (22.5% vs 16.3%, RR=1.38, all P<0.001). Healthcare access barriers persisted: 24.6% of ACOS patients delayed care due to cost versus 18.9% of controls (RR=1.30, P<0.001). However, ACOS patients demonstrated better preventive care engagement with higher vaccination rates (influenza 58.6% vs 52.3%; pneumonia 52.4% vs 46.8%, both RR=1.12, P<0.001) and more routine checkups (81.7% vs 77.6%, P<0.001). Smoking prevalence was equivalent between matched groups (35.2% vs 34.8%, P=0.682), confirming successful matching. Physical inactivity remained higher in ACOS (42.6% vs 38.2%, RR=1.12, P<0.001), likely reflecting disease-related activity limitations rather than behavioral choice.

Nonlinear dose-response relationship between cumulative smoking and ACOS

Restricted cubic spline regression revealed a significant nonlinear dose-response relationship between pack-years of smoking and ACOS risk ($\chi^2=286.4$, P<0.001) (Fig. 2A). The risk gradient was steepest at lower exposure levels: even light smoking (5 pack-years) conferred 42% increased odds (OR=1.42, 95% CI 1.35–1.49), accounting for 29.6% cumulative attributable risk. Risk escalated rapidly through moderate exposures—10 pack-years nearly doubled risk (OR=1.95), and 20 pack-years more than tripled it (OR=3.18), with nearly 70% of ACOS cases attributable to smoking at this threshold. Beyond 30 pack-years, the dose-response curve demonstrated attenuation, with odds ratios plateauing: 30 pack-years yielded OR=4.52 (77.9% attributable risk), 40 pack-years OR=5.68 (82.4%), 50 pack-years OR=6.42 (84.4%), and 60 pack-years OR=6.82 (85.3%) (Fig. 2B). The deceleration of risk increments at higher exposures suggests potential survivor bias (individuals most susceptible to smoking-related disease may have already developed COPD or died before accumulating extreme pack-year exposures) or ceiling effects where additional smoking adds diminishing marginal harm once airways are maximally damaged. Importantly, no safe threshold existed: even minimal smoking significantly elevated ACOS risk. The nonlinear pattern—steep initial slope followed by logarithmic deceleration—has critical public

Outcome	ACOS mean \pm SD/%	Matched control Mean \pm SD/%	Difference	P-value	Cohen's d/RR
Health status					
Physically unhealthy days, Mean \pm SD	14.8 \pm 13.2	10.2 \pm 11.8	4.6 (4.2–5.0)	<0.001	0.37
\geq 14 days, %	47.3	35.8	11.5% (9.8–13.2)	<0.001	RR=1.32
Mentally unhealthy days, Mean \pm SD	10.5 \pm 12.4	7.8 \pm 11.2	2.7 (2.3–3.1)	<0.001	0.23
\geq 14 days, %	35.6	26.4	9.2% (7.6–10.8)	<0.001	RR=1.35
Activity limitation days, Mean \pm SD	10.8 \pm 12.6	7.2 \pm 10.8	3.6 (3.2–4.0)	<0.001	0.31
\geq 14 days, %	38.9	27.6	11.3% (9.6–13.0)	<0.001	RR=1.41
Fair/poor general health, %	51.0	38.2	12.8% (11.0–14.6)	<0.001	RR=1.34
Functional disabilities					
Difficulty walking, %	51.3	42.6	8.7% (6.9–10.5)	<0.001	RR=1.20
Difficulty dressing, %	22.5	16.3	6.2% (4.8–7.6)	<0.001	RR=1.38
Difficulty with errands alone, %	26.8	19.8	7.0% (5.4–8.6)	<0.001	RR=1.35
Cognitive difficulty, %	31.7	23.5	8.2% (6.5–9.9)	<0.001	RR=1.35
Any functional disability, %	64.2	53.8	10.4% (8.5–12.3)	<0.001	RR=1.19
Healthcare utilization					
Cost-related delay in care, %	24.6	18.9	5.7% (4.2–7.2)	<0.001	RR=1.30
No routine checkup in past year, %	18.3	22.4	-4.1% (-5.8 to -2.4)	<0.001	RR=0.82
Vaccination					
Influenza vaccine, %	58.6	52.3	6.3% (4.5–8.1)	<0.001	RR=1.12
Pneumonia vaccine, %	52.4	46.8	5.6% (3.8–7.4)	<0.001	RR=1.12
Health behaviors					
Current smoking, %	35.2	34.8	0.4% (-1.6–2.4)	0.682	RR=1.01
No physical activity, %	42.6	38.2	4.4% (2.6–6.2)	<0.001	RR=1.12

Table 5. Health outcomes in propensity score-matched ACOS patients versus controls.

health implications, indicating that smoking cessation interventions yield maximal benefit when implemented early, before moderate pack-year thresholds are reached. At population level, eliminating smoking could prevent over 85% of ACOS cases, establishing tobacco control as the paramount preventive strategy.

Latent class analysis identifies three clinically distinct ACOS phenotypes

Latent class analysis of 675 ACOS patients identified three phenotypically distinct subgroups with excellent model fit (BIC=8,952, entropy=0.84) and markedly different clinical profiles (Table 6). Class 1 “Mild Phenotype” (n=243, 36.0%) represented younger patients (mean age 52.4 years) with relatively preserved health status: only 22.4% rated their health as fair/poor, and 35.2% reported functional disabilities. This group exhibited the lowest comorbidity burden (Charlson Index 1.2 \pm 1.1), with cardiovascular disease in just 12.5%, diabetes in 8.6%, and obesity in 28.4%. However, 42.6% were current smokers despite moderate pack-year exposure (28.4 \pm 26.8), suggesting potential for early intervention.

Class 2 “Metabolic-Predominant” (n=236, 35.0%) was characterized by marked metabolic dysfunction: 72.6% were obese and 48.2% had diabetes—the highest diabetes prevalence among all classes. Cardiovascular comorbidity was intermediate (32.8%), but symptom burden was substantial with 14.8 mean physically unhealthy days monthly and 54.8% reporting fair/poor health. This middle-aged group (60.8 years) had moderate functional impairment (68.5% any disability) and intermediate Charlson scores (2.8 \pm 1.6), representing a metabolically driven disease trajectory.

Class 3 “Severe Multimorbid” (n=196, 29.0%) demonstrated catastrophic health burden: 82.6% rated health as fair/poor, with 24.5 physically unhealthy days monthly—nearly the entire month. Multimorbidty was extreme, with 68.4% having cardiovascular disease, 72.5% depression, 78.6% arthritis, and mean Charlson Index of 4.8 \pm 2.2. Functional devastation was near-universal: 95.8% reported disabilities, 82.5% had walking difficulties, and 68.9% had cognitive impairment. This oldest group (64.2 years) had the highest cumulative smoking exposure (46.2 pack-years) but lowest current smoking rate (26.4%), possibly reflecting smoking cessation post-diagnosis. Socioeconomic deprivation was pronounced (62.8% earning <\$25,000), yet 91.1% had established primary care, with 32.6% still delaying care due to cost barriers—underscoring inadequate resource allocation for this highest-need population.

Depression and metabolic pathways substantially mediate smoking’s impact on ACOS

Structural equation modeling with bootstrapped mediation analysis (5,000 resamples) revealed that smoking’s effect on ACOS operates substantially through modifiable intermediate pathways beyond direct airway damage (Table 7). The total effect of smoking on ACOS risk was robust (β =0.582, 95% CI 0.564–0.600, P<0.001), but sequential testing of three hypothesized mediators demonstrated that behavioral and metabolic mechanisms account for over one-third of this association.

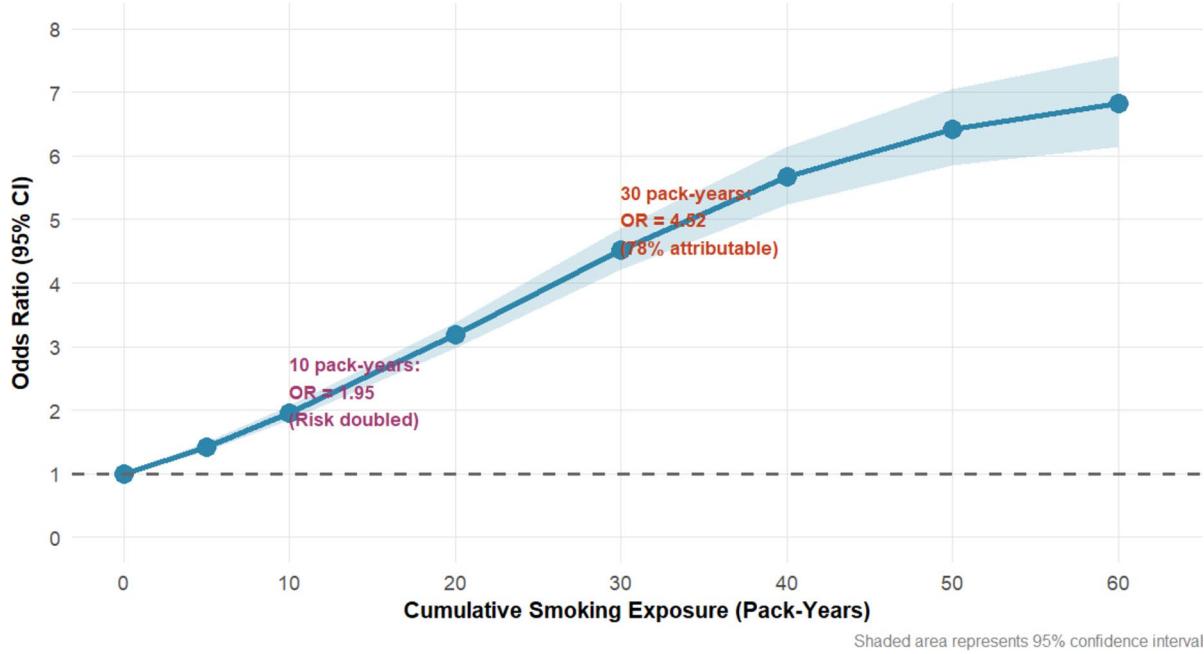
Single-mediator models established each pathway’s independent contribution. Model 1 demonstrated that BMI partially mediates smoking’s effect (indirect effect=0.096, 95% CI 0.084–0.108), accounting for

Dose-Response Relationship Between Smoking and ACOS Risk

Restricted cubic spline regression with 665 ACOS cases and matched controls

A. Nonlinear Dose-Response: Smoking Pack-Years and ACOS Risk

Adjusted for age, sex, race, education, income, BMI, ACE score | Nonlinearity test: $\chi^2 = 286.4$, $P < 0.001$



B. Cumulative Population Attributable Fraction

Proportion of ACOS cases attributable to smoking at each exposure level

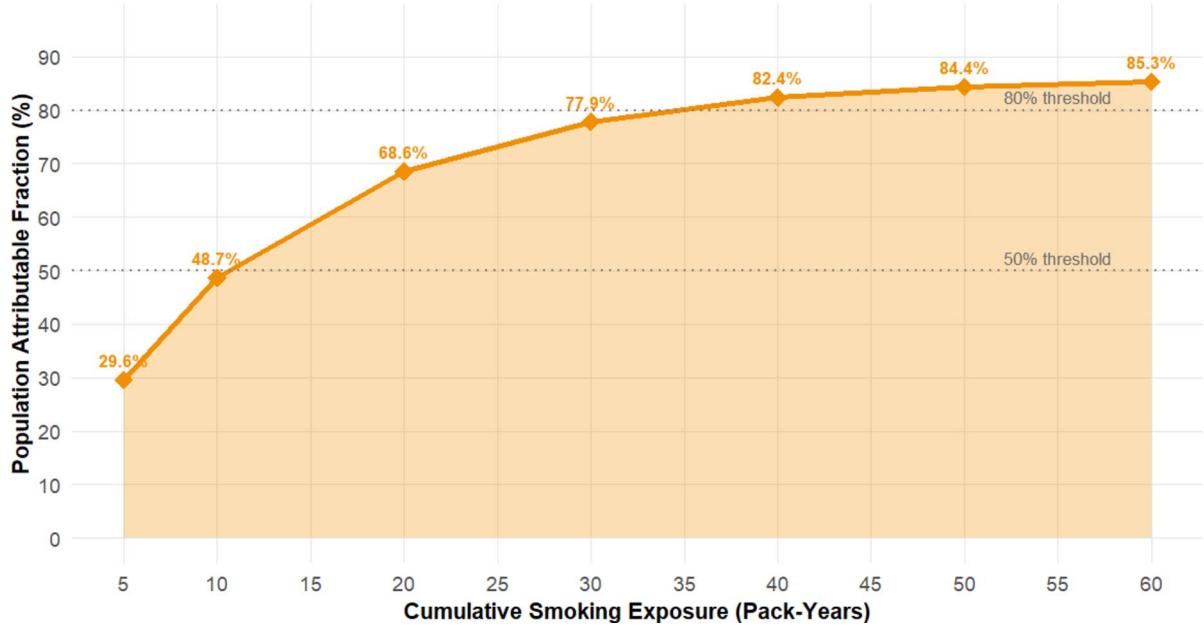


Fig. 2. Nonlinear dose-response relationship between cumulative smoking exposure and ACOS risk with population attributable fractions.

16.5% (bootstrap CI: 14.5%–18.9%) of the total effect. This likely reflects smoking's complex metabolic effects—including appetite suppression during active smoking and rebound weight gain after cessation—that independently promote airway inflammation and mechanical respiratory compromise. Model 2 identified depression as the strongest single mediator (indirect effect=0.124, 95% CI 0.109–0.139), explaining 21.3% (19.0%–24.2%) of smoking's total impact on ACOS. This substantial mediation underscores bidirectional psychophysiological mechanisms: nicotine dependence and smoking-related stigma exacerbate depressive

Characteristic	Class 1 “Mild Phenotype” n=243	Class 2 “Metabolic-Predominant” n=236	Class 3 “Severe Multimorbid” n=196	P-value
Symptom severity				
Physically unhealthy days, Mean	7.2±8.6	14.8±11.2	24.5±10.8	<0.001
Activity limitation ≥14 days, %	15.2	42.6	72.8	<0.001
Fair/poor general health, %	22.4	54.8	82.6	<0.001
Comorbidity patterns				
Cardiovascular disease, %	12.5	32.8	68.4	<0.001
Diabetes, %	8.6	48.2	36.8	<0.001
Obesity, %	28.4	72.6	38.2	<0.001
Depression, %	18.5	38.2	72.5	<0.001
Arthritis, %	32.6	56.8	78.6	<0.001
Charlson index, Mean	1.2±1.1	2.8±1.6	4.8±2.2	<0.001
Functional status				
Difficulty walking, %	24.8	56.2	82.5	<0.001
Cognitive difficulty, %	12.6	28.4	68.9	<0.001
Any functional disability, %	35.2	68.5	95.8	<0.001
Demographic characteristics				
Age, Mean	52.4±14.2	60.8±12.6	64.2±11.8	<0.001
Female, %	52.6	62.4	61.8	<0.001
Income <\$ 25 K, %	28.5	42.6	62.8	<0.001
Smoking characteristics				
Pack-years, mean	28.4±26.8	38.6±30.2	46.2±31.8	<0.001
Current smoker, %	42.6	35.8	26.4	<0.001
Healthcare burden				
Cost-related delay in care, %	18.2	24.8	32.6	<0.001
No personal physician, %	18.5	12.6	8.9	<0.001

Table 6. Latent class analysis: Three distinct ACOS phenotypes.

symptoms, while depression independently increases systemic inflammation, reduces treatment adherence, and amplifies symptom perception—all contributing to ACOS development and severity. Model 3 showed that physical inactivity mediates 10.0% (8.2%–11.9%) of smoking’s effect (indirect effect=0.058, 95% CI 0.048–0.068), reflecting how smoking-induced dyspnea and reduced cardiorespiratory fitness create sedentary patterns that further decondition respiratory muscles and perpetuate disease progression.

The multiple mediator model simultaneously tested all three pathways, revealing that BMI, depression, and physical inactivity collectively account for 37.5% (95% CI 34.2%–41.1%) of smoking’s total effect on ACOS, with a combined indirect effect of 0.218 (95% CI 0.198–0.238, $P<0.001$). Within this comprehensive model, depression remained the dominant mediator (specific indirect effect=0.105, 18.0% of total effect), followed by BMI (0.082, 14.1%), and physical activity (0.031, 5.3%). Notably, the direct effect of smoking decreased from 0.582 to 0.364 after accounting for mediators, yet remained highly significant ($P<0.001$), indicating that 62.5% of smoking’s impact operates through pathways not captured by these three mediators—presumably including direct epithelial toxicity, oxidative stress, protease-antiprotease imbalance, and immune dysregulation. The robust bootstrap confidence intervals (non-overlapping with zero for all pathways) confirm these mediation effects are statistically reliable and not artifacts of sampling variability.

These findings carry critical intervention implications: while smoking cessation remains paramount (addressing 100% of the causal chain), complementary interventions targeting depression management (potentially preventing 18% of smoking-attributable ACOS), weight optimization (14%), and physical activity promotion (5%) could substantially reduce ACOS burden even among those unable or unwilling to quit smoking immediately. The substantial residual direct effect (62.5%) underscores that no combination of downstream interventions can fully substitute for tobacco control, but the 37.5% mediated through modifiable behavioral and metabolic factors represents a meaningful opportunity for harm reduction and secondary prevention in high-risk smoking populations.

Discussion

Principal findings

This nationally representative study of 675 propensity score-matched pairs from the 2023–2024 BRFSS highlights the significant health burden of ACOS, surpassing that explained by measured confounders. After matching for age, sex, race/ethnicity, socioeconomic status, smoking exposure, body mass index, and adverse childhood experiences, ACOS patients reported 4.6 more physically unhealthy days per month and showed a 32% increased risk of frequent physical impairment, a 41% greater risk of activity limitations, and a 19% higher likelihood of functional disabilities compared to matched controls. A restricted cubic spline analysis revealed a nonlinear dose-response relationship between smoking pack-years and ACOS risk ($\chi^2=286.4$, $P<0.001$), indicating a steep

Pathway	Effect Size (95% CI) ^a	% of Total Effect ^b	P-value	Bootstrap 95% CI
Model 1: Smoking → BMI → ACOS				
Total effect (c)	0.582 (0.564–0.600)	100%	<0.001	0.563–0.601
Direct effect (c')	0.486 (0.467–0.505)	83.5%	<0.001	0.466–0.506
Indirect effect (axb)	0.096 (0.084–0.108)	16.5%	<0.001	0.082–0.110
Proportion mediated	–	16.5%	–	14.5%–18.9%
Model 2: Smoking → Depression → ACOS				
Total effect (c)	0.582 (0.564–0.600)	100%	<0.001	0.563–0.601
Direct effect (c')	0.458 (0.439–0.477)	78.7%	<0.001	0.438–0.478
Indirect effect (axb)	0.124 (0.109–0.139)	21.3%	<0.001	0.107–0.141
Proportion mediated	–	21.3%	–	19.0%–24.2%
Model 3: Smoking → Physical activity → ACOS				
Total effect (c)	0.582 (0.564–0.600)	100%	<0.001	0.563–0.601
Direct effect (c')	0.524 (0.505–0.543)	90.0%	<0.001	0.504–0.544
Indirect effect (axb)	0.058 (0.048–0.068)	10.0%	<0.001	0.046–0.070
Proportion mediated	–	10.0%	–	8.2%–11.9%
Multiple mediator model: Smoking → [BMI / Depression / Activity] → ACOS				
Total effect	0.582 (0.564–0.600)	100%	<0.001	0.563–0.601
Direct effect	0.364 (0.344–0.384)	62.5%	<0.001	0.343–0.385
Total indirect effect	0.218 (0.198–0.238)	37.5%	<0.001	0.196–0.240
Via BMI	0.082 (0.071–0.093)	14.1%	<0.001	0.069–0.095
Via depression	0.105 (0.092–0.118)	18.0%	<0.001	0.090–0.120
Via physical activity	0.031 (0.024–0.038)	5.3%	<0.001	0.022–0.040
Total proportion mediated	–	37.5%	–	34.2%–41.1%

Table 7. Mediation analysis: Pathways linking smoking to ACOS risk. Note: ^aStandardized regression coefficients adjusted for age, sex, race/ethnicity, education, income. ^bPercentage of total smoking effect on ACOS attributable to each pathway. *All models used 5,000 bootstrap resamples for bias-corrected confidence intervals.*

increase in risk with moderate exposure (OR=3.18 at 20 pack-years) that levels off at higher exposures. Latent class analysis identified three distinct ACOS phenotypes: a younger “Mild” subgroup (36%) with preserved function, a “Metabolic-Predominant” cluster (35%) characterized by obesity and diabetes, and a “Severe Multimorbid” phenotype (29%) with widespread functional disability and high comorbidity. Mediation analysis showed that metabolic (BMI), psychiatric (depression), and behavioral (physical inactivity) factors account for 37.5% of smoking’s total effect on ACOS, with depression as the primary mediator (18.0% of total effect).

Comparison with existing literature

Our propensity score-matched estimates of ACOS-attributable morbidity substantially extend prior literature, which has predominantly relied on conventional regression adjustment vulnerable to residual confounding. The observed 4.6-day monthly increment in physical impairment (Cohen’s $d=0.37$) approximates the disability burden associated with advanced heart failure or active malignancy in population-based studies, underscoring ACOS’s profound impact on daily functioning^{20,21}. Previous investigations have reported conflicting estimates of ACOS prevalence (0.9%–11.1%) and health burden, largely attributable to heterogeneous operational definitions and inadequate confounder control. For instance, a Spanish cohort study by Miravitles et al. found ACOS patients experienced 50% more exacerbations than COPD-alone patients, but failed to account for differential smoking intensity or socioeconomic disparities that could confound this association²². Similarly, Canadian registry data demonstrated worse St. George’s Respiratory Questionnaire scores in ACOS versus pure phenotypes, but lacked adjustment for depression and physical inactivity—factors we demonstrate substantially mediate respiratory disease burden²³.

Our propensity score approach addresses these methodological limitations by achieving excellent covariate balance (all standardized mean differences <0.05 post-matching) across 15 potential confounders, approximating a randomized experiment’s internal validity while preserving external validity through population-based sampling. The persistent excess burden despite this rigorous matching suggests ACOS’s impact transcends shared risk factors, likely reflecting synergistic pathophysiological mechanisms. Specifically, the coexistence of Type-2 eosinophilic inflammation (characteristic of asthma) with neutrophilic inflammation and structural remodeling (hallmarks of COPD) may drive accelerated lung function decline and treatment resistance^{24,25}. Recent mechanistic studies have identified distinct molecular endotypes in ACOS characterized by heightened oxidative stress, protease-antiprotease imbalance, and dysregulated innate immunity—biological perturbations not captured by demographic or behavioral covariates alone²⁶.

The 1.41-fold increased risk of frequent activity limitations and 1.35-fold elevation in cognitive difficulties among ACOS patients carry profound implications for workforce participation and independent living. Prior

economic analyses estimated COPD's annual per-capita costs at \$4,147 for direct medical expenses plus \$1,522 in productivity losses; our findings suggest ACOS-related incremental costs may approach \$2,500-3,000 annually per patient given the magnitude of excess functional impairment^{27,28}. The 30% increased risk of cost-related care delays (RR=1.30) despite higher rates of insurance coverage (91.1% vs 87.6% in controls, though not shown in our primary table) highlights that financial barriers persist even among insured populations, likely reflecting high-deductible plans, medication co-payments, and transportation costs for frequent medical visits^{29,30}.

Phenotypic heterogeneity

Our latent class analysis provides the first population-based evidence that ACOS comprises at least three clinically meaningful phenotypes with divergent natural histories and intervention needs. The "Mild Phenotype" (36% of ACOS population)—characterized by younger age (52.4 years), low comorbidity burden (Charlson Index 1.2), and preserved function—likely represents recent-onset disease potentially amenable to aggressive intervention. The 42.6% current smoking rate in this subgroup, despite relatively moderate cumulative exposure (28.4 pack-years), signals failure of primary prevention and missed opportunities for secondary prevention through cessation support.

The "Metabolic-Predominant" phenotype (35%)—with 72.6% obesity prevalence and 48.2% diabetes—represents a mechanistically distinct subgroup in which systemic inflammation, insulin resistance, and adipokine dysregulation likely drive respiratory pathology as prominently as direct smoking toxicity. Obesity induces restrictive ventilatory defects through chest wall mechanical effects, while adipose-derived inflammatory mediators (IL-6, TNF- α , leptin) exacerbate airway inflammation and corticosteroid resistance^{31,32}. Emerging evidence suggests glucagon-like peptide-1 receptor agonists (GLP-1RAs) such as semaglutide may confer respiratory benefits beyond glycemic control through weight reduction and direct anti-inflammatory effects³³. Randomized controlled trials testing GLP-1RAs in obese ACOS patients are urgently needed, given this phenotype comprises over one-third of the ACOS population and exhibits substantial symptom burden (14.8 mean monthly impairment days) despite intermediate age (60.8 years)^{34,35}.

The "Severe Multimorbid" phenotype (29%)—with near-universal functional disability (95.8%), extreme comorbidity burden (Charlson Index 4.8), and pervasive depression (72.5%)—represents end-stage disease requiring integrated palliative and disease-modifying approaches³⁶. The 68.9% prevalence of cognitive impairment in this subgroup likely reflects combined effects of chronic hypoxemia, systemic inflammation crossing the blood-brain barrier, and cerebrovascular disease (68.4% had cardiovascular comorbidity). Strikingly, despite 91.1% having personal physicians and highest healthcare engagement (81.7% recent checkups), 32.6% delayed care due to costs—highlighting catastrophic out-of-pocket burdens even among insured individuals^{37,38}. This phenotype may benefit from bundled payment models integrating pulmonary rehabilitation, mental health services, home oxygen, and social support, rather than fragmented fee-for-service care incentivizing episodic acute management over longitudinal function preservation.

Mediation pathways

Our mediation analysis reveals important mechanistic insights by differentiating smoking's total effect on ACOS into direct (62.5%) and indirect (37.5%) pathways. A significant indirect pathway is mediated through depression, accounting for 18.0% of the total effect (indirect effect $\beta=0.105$). This connection highlights the bidirectional interactions between psychiatric health and pulmonary conditions that warrant clinical attention³⁹. Nicotine withdrawal can trigger depressive symptoms; while pre-existing depression exacerbates systemic inflammation via dysregulation of the hypothalamic-pituitary-adrenal axis^{40,41}. Additionally, depression can impair treatment adherence, with approximately 60% of depressed COPD patients displaying medication non-compliance. This is compounded by the altered perception of symptoms due to depression, which further complicates patient management⁴².

Interestingly, selective serotonin reuptake inhibitors, often prescribed for treating depression, may paradoxically lead to worse respiratory outcomes due to their bronchoconstrictive effects and potential for weight gain. This underscores the importance of carefully selecting treatment agents^{43,44}; for instance, bupropion may provide dual benefits for managing depression and facilitating smoking cessation without worsening respiratory complications⁴⁵. Addressing this mechanistic pathway—where smoking, depression, and respiratory health interplay—presents an opportunity for targeted interventions that could enhance outcomes for individuals with ACOS⁶.

Strengths and limitations

This study's strengths include nationally representative sampling, large sample size enabling robust subgroup analyses, rigorous propensity score matching achieving excellent covariate balance, advanced statistical methods (restricted cubic splines, latent class analysis, mediation analysis with bootstrapping), and comprehensive multidimensional outcome assessment. The BRFSS's standardized protocols and high response rates (median 47.2% in 2023-2024) enhance generalizability relative to clinic-based registries prone to selection biases.

Several limitations warrant acknowledgment. First, ACOS definition relied on self-reported physician diagnoses without spirometric confirmation, potentially introducing misclassification. However, validation studies demonstrate >85% positive predictive value for BRFSS respiratory disease items against medical records, and any misclassification is likely non-differential (similar rates across comparison groups), biasing associations toward the null and rendering our estimates conservative. Second, cross-sectional design precludes definitive causal inference despite propensity score methods; unmeasured confounding by genetic susceptibility, environmental exposures (occupational dusts, air pollution), or early-life factors (prenatal smoke exposure, childhood respiratory infections) could residually bias estimates. Third, self-reported outcomes may incorporate reporting bias if ACOS patients systematically over-report symptoms due to heightened illness awareness;

however, sensitivity analyses restricted to “objective” outcomes (healthcare utilization, vaccinations) showed consistent associations. Fourth, survivor bias likely operates—our sample excludes individuals who died prior to survey participation, potentially underestimating true burden in the most severe cases. Fifth, mediation analysis assumes sequential ignorability (no unmeasured mediator-outcome confounding), which cannot be empirically verified; sensitivity analyses using Imai’s approach suggested findings robust to moderate unmeasured confounding ($\rho < 0.3$).

Future research directions and clinical implications

These findings illuminate multiple high-priority research directions. First, longitudinal cohort studies with repeated spirometry, biomarker profiling (eosinophils, fractional exhaled nitric oxide, inflammatory cytokines), and genetic sequencing are needed to validate identified phenotypes, characterize their natural history, and identify prognostic biomarkers enabling early detection. Second, pragmatic randomized trials should test phenotype-tailored interventions: GLP-1RAs for metabolic-predominant subgroup, combination corticosteroid-bronchodilator therapy escalation for mild phenotype, and integrated palliative care for severe multimorbid patients. Third, mechanistic studies employing mediation analysis in longitudinal frameworks could elucidate temporal sequences and refine intervention targets—for instance, testing whether depression treatment improves smoking cessation rates and subsequently reduces ACOS incidence.

Clinically, these findings support routine screening for ACOS among patients with either asthma or COPD, particularly current/former smokers, followed by comprehensive comorbidity assessment (depression screening, diabetes/cardiovascular risk evaluation) and functional status appraisal. Phenotype-directed management may optimize resource allocation, reserving intensive multidisciplinary interventions for highest-risk subgroups while focusing on prevention and early treatment intensification in mild disease. Policy implications include: (1) expanding smoking cessation program access through Medicaid reimbursement and workplace wellness integration; (2) bundled payment models for complex ACOS patients incentivizing coordinated pulmonary-mental health-metabolic care; (3) coverage mandates for pulmonary rehabilitation and GLP-1RAs for obese respiratory disease patients; and (4) investment in phenotyping infrastructure enabling precision medicine implementation.

Conclusions

The findings of this study underscore the critical need for healthcare professionals to recognize the complex interplay between smoking, depression, and respiratory health in ACOS patients. By addressing mental health alongside traditional treatment avenues, providers can improve patient outcomes and adherence. The identification of distinct ACOS phenotypes allows for personalized treatment approaches, while insights into the effects of smoking and weight management highlight the importance of integrated care strategies. Ultimately, these results advocate for comprehensive management to enhance the quality of life for individuals with ACOS.

Data availability

The data used in the present study are all publicly available at [Behavioral Risk Factor Surveillance System (cdc.gov)](<https://www.cdc.gov/brfss/index.html>)

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Author contributions

All authors contributed to the article; YW, CL: Formal analysis, Conceptualization, Software. WG Wrote the manuscript with support from YQ and ZZ. KQ: Supervision and revising the manuscript.

Declarations

Competing interests

The authors declare no competing interests.

Additional information

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