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J-shaped associations of pan-immune-inflammation value and systemic inflammation response index with stroke among American adults with hypertension: evidence from NHANES 1999–2020

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Introduction: Stroke, a leading cause of death and disability worldwide, is primarily ischemic and linked to hypertension. Hypertension, characterized by systemic chronic inflammation, significantly increases stroke risk. This study explores the association of novel systemic inflammatory markers (SII, PIV, SIRI) with stroke prevalence in hypertensive U.S. adults using NHANES data.

Methods: We analyzed data from hypertensive participants in the NHANES 1999–2020 survey, excluding those under 20, pregnant, or with missing data, resulting in 18,360 subjects. Systemic inflammatory markers (SII, PIV, SIRI) were calculated from blood counts. Hypertension and stroke status were determined by self-report and clinical measurements. Covariates included sociodemographic, lifestyle, and medical history factors. Weighted statistical analyses and multivariate logistic regression models were used to explore associations, with adjustments for various covariates. Ethical approval was obtained from the NCHS Ethics Review Board.

Results: In a cohort of 18,360 hypertensive individuals (mean age 56.652 years), 7.25% had a stroke. Stroke patients were older, had lower PIR, and were more likely to be female, single, less educated, smokers, non-drinkers, physically inactive, and have diabetes and CHD. Multivariate logistic regression showed that SII was not significantly associated with stroke. However, PIV and SIRI were positively associated with stroke prevalence. Each unit increase in lnPIV increased stroke odds by 14% (OR = 1.140, p = 0.0022), and lnSIRI by 20.6% (OR = 1.206, p = 0.0144). RCS analyses confirmed J-shaped associations for lnPIV and lnSIRI with stroke. Stratified analyses identified gender and smoking as significant effect modifiers. Smoking was significantly associated with elevated PIV, SIRI, and SII levels, especially in current smokers.

Conclusion: Elevated PIV and SIRI levels significantly increase stroke prevalence in hypertensive individuals, notably among males and smokers. A predictive model with PIV, SIRI, and sociodemographic factors offers strong clinical utility.

KEYWORDS

systemic inflammation, stroke, NHANES, complete blood count, hypertension, smoking

1 Introduction

Stroke is an acute cerebrovascular disease in which abnormal blood supply to the brain results in impaired brain function or death (1). Strokes are mainly divided into ischemic and hemorrhagic strokes, with ischemic strokes accounting for most cases (2). Stroke is the leading cause of death and severe disability worldwide. The recent Global Burden of Disease Study 2019 report suggests that in 2019, there were 12.2 million stroke incident cases, 101 million stroke prevalent cases, 143 million disability-adjusted life years attributable to stroke, and 6.55 million stroke deaths (the second leading cause of death) worldwide (3). The absolute number of incident and prevalent cases of stroke has increased significantly over the last three decades, although age-standardized incidence, prevalence, and mortality rates have declined (but remain elevated among those under 70 years of age) (3). Stroke not only dramatically reduces patients' quality of life, but also places a heavy economic burden on global public health. A survey by the American Heart Association put the annual direct and indirect economic burden of stroke at \$45.5 billion in 2014-2015 (4). Stroke prevention and early intervention by identifying modifiable risk factors such as promoting healthy lifestyles and addressing comorbidities are key to reducing the burden of disease (5, 6).

Hypertension is one of the most common non-communicable diseases globally and represents the most significant modifiable component of all-cause morbidity and mortality (7). Hypertension is strongly associated with the development of several cardiovascular diseases (CVD), including stroke. Hypertension is one of the major modifiable risk factors for stroke, which is significantly more prevalent in hypertensive populations and is associated with considerable disability and mortality (8-10). An important hallmark of hypertension is a state of systemic chronic inflammation, a mechanism thought to be involved in the pathogenesis of stroke in hypertensive patients (11-13). Inflammatory cells including T cells and inflammatory cytokines have an important causal role in hypertension-mediated target organ damage (14). Therefore, as a shared pathophysiologic event of hypertension and stroke, the exploration of relevant systemic inflammatory markers may be of great significance for stroke prediction, screening, and prevention in hypertensive individuals.

Multiple immune cell and inflammatory markers including multiple interleukins, and tumor necrosis factor-alpha have been shown to be elevated in hypertensive patients (15). Recently, some novel markers of systemic inflammation derived from complete blood count (CBC) have been proposed. Systemic inflammatory markers are a group of biomarkers that reflect the body's response to inflammation. A variety of conditions, including cardiovascular diseases, autoimmune diseases, and infections, can be assessed using these markers to assess inflammation, monitor disease progression, and predict outcomes. These novel systemic inflammatory markers, including systemic immune-inflammation index (SII), pan-immuneinflammation value (PIV), and systemic inflammation response index (SIRI), account for the role of CBC-derived inflammatory cells more comprehensively than the traditional neutrophil-to-lymphocyte ratio (NLR), monocyte-to-lymphocyte ratio (MLR), and platelet-tolymphocyte ratio (PLR), and have been first proposed to be strongly correlated with prognosis in multiple cancers (16–18). Several clinical studies have shown that these systemic inflammatory markers are strongly associated with the development and clinical outcome of hypertension (19–22). However, there is still a lack of research on the association of these systemic inflammatory markers with stroke occurrence in hypertensive populations. Only two retrospective studies have suggested that SII and SIRI are associated with the risk of stroke in hypertensive individuals, and both were relatively small cohorts from Asia (23, 24). There is a dearth of large sample and representative studies from other countries/regions. Furthermore, whether PIV is also a predictor of stroke in hypertensive populations remains unknown.

In this study, we aimed to explore the association of SII, PIV, and SIRI with stroke prevalence in hypertensive subjects using a U.S. nationally representative population-based cross-sectional survey, the National Health and Nutrition Examination Survey (NHANES). These findings may help to reveal the predictive value of these novel systemic inflammatory markers for stroke in hypertensive adults in the U.S. and provide a rationale for stroke prevention.

2 Methods

2.1 Study design and population

NHANES is a national survey conducted by the National Center for Health Statistics (NCHS) to assess the health and nutritional status of noninstitutionalized children and adults in the U.S. NCHS is part of the Centers for Disease Control (CDC) and is responsible for providing vital health epidemiologic data for the nation. NHANES is a nationally representative cross-sectional survey with multi-stage cluster probability sampling that includes questionnaires from comprehensive household interviews and physical examination data administered by professionals (25, 26).

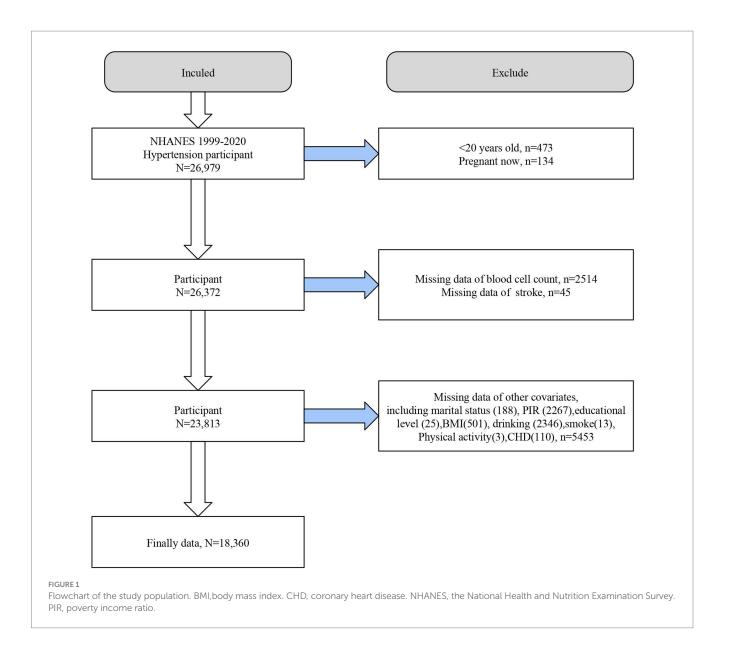
We first included all hypertensive participants in NHANES 1999–2020 (n = 26,979) and excluded individuals <20 years of age (n = 473), pregnant individuals (n = 134), those with missing blood count data (n = 2,514), missing stroke diagnostic information (n = 45), and missing covariates (n = 5,453). Ultimately, 18,360 individuals with hypertension were included in further analyses (Figure 1).

2.2 Calculation of SII, PIV, and SIRI

Beckman Coulter DxH 800 instrument at NHANES Mobile Examination Center was used to determine neutrophil, lymphocyte, monocyte, and platelet counts (expressed as×103 cells/µl). SII=(platelet count×neutrophil count)/lymphocyte count; PIV=(platelet count×neutrophil count×monocyte count)/lymphocyte count; SIRI=(neutrophil count×monocyte count)/lymphocyte count. Since SII, PIV, and SIRI were skewed in the study population, we performed a natural logarithmic (ln) transformation of their values.

2.3 Hypertension diagnosis

Hypertensive status was determined by a self-reported history of hypertension, a blood pressure test that indicated hypertension (≥140/90 mmHg) or taking anti-hypertensive medications (20). Blood pressure testing is performed by professional technicians at the



Mobile Examination Center (MEC) using mercury sphygmomanometers or electronic blood pressure measuring devices (NHANES 2017–2020) for three consecutive systolic and diastolic blood pressure measurements and in accordance with relevant international guidelines (27).

2.4 Stroke assessment

Participants' stroke status was determined by self-reported medical history in NHANES. Participants who answered "yes" to the question "Has a doctor or other health professional ever told you that you had a stroke?" were considered stroke survivors.

2.5 Covariates

Participants' sociodemographic characteristics, lifestyle, and medical comorbidity information were collected. Sociodemographic variables include age, gender, race/ethnicity, education level, family income-poverty ratio (PIR), and marital status, which were obtained through self-reporting in the NHANES demographic file. Lifestyle variables include body mass index (BMI), smoking, alcohol consumption, and physical activity. BMI = body weight (kg) divided by height (m) squared. Smoking status was categorized into current (active) smokers and inactive smokers (namely never smokers and former smokers) and was determined based on whether participants self-reported a lifetime number of cigarettes of at least 100 and whether they were currently smoking. Drinking history was categorized as never drinkers, former drinkers, light, moderate and heavy drinkers based on previous studies (28). Physical activity intensity was categorized as never, moderate, and vigorous based on self-reports from participants in the Global Physical Activity Questionnaire (29). The history of medical comorbidities includes diabetes and coronary heart disease (CHD). Diabetes status was determined by self-report, glucose/glucose tolerance testing, and history of relevant medication use (30). CHD history was obtained by self-report.

2.6 Statistical analysis

We weighted the statistical analyses according to the weighting method recommended by the NHANES Analytic Guidelines (WTMEC2YR) to ensure that the study sample was nationally representative. Data were processed and analyzed using R version 4.2.3 and EmpowerStats software. A two-sided p value of <0.05 was considered statistically significant. We performed baseline analyses according to stroke status in the hypertensive population (stroke and non-stroke groups), continuous variables were analyzed for between-group differences using weighted t tests and expressed as mean and standard error, and categorical variables were analyzed using weighted chi-square tests and expressed as number (percentage). Multivariate logistic regression models with varying degrees of adjustment were used to explore the association of SII, PIV, and SIRI with stroke in hypertensive individuals. Model 1 did not adjust for any covariates; Model 2 was a partial adjusted model, adjusting for age, gender, race, PIR, education level, and marital status; and Model 3 was a fully adjusted model, adjusting for all covariates. Restricted cubic spline (RCS) analysis was used to discuss potential nonlinear correlations. The curvefitting term was defined by the RCS function in the rms package, and the degrees of freedom (or knots) were determined according to the magnitude of the nonlinearity p value. The correlation matrix analysis was applied in this study to explore the relationship between SII, PIV, and SIRI. Stratified analyses were performed to explore whether these associations remained consistent across subgroups and to identify potential effect modifiers. The univariate analysis was conducted to assess the relationship between potential effect modifier and elevated levels of SII, PIV, and SIRI.

2.7 Ethics statement

All NHANES datasets included in this study were reviewed and approved by the NCHS Ethics Review Board, and all subjects provided written informed consent. This study analyzed pre-existing public datasets and all participants were de-identified and anonymized, hence ethical review approval from local institutions was waived.

3 Results

3.1 Baseline analysis

Baseline analyses according to stroke status in the hypertensive population were presented in Table 1. A total of 18,360 hypertensive individuals were included (mean age 56.652 years), with a stroke prevalence of 7.25%. Compared with participants without stroke, the stroke population was older, had a lower PIR, and was more likely to be female, single, \leq high school educated, quitters/current smokers, never drinkers/abstainers, physically inactive, and to have diabetes and CHD. Lymphocyte counts did not differ between groups, whereas significant differences existed in monocyte, neutrophil, and platelet counts. Interestingly, we found that lnSII was not different between groups (p = 0.077), whereas lnPIV and lnSIRI were significantly higher in stroke patients (p < 0.0001 for both).

TABLE 1 Survey-weighted baseline characteristics of individuals with hypertension from NHANES 1999–2020.

	Total	No-stroke	Stroke	<i>p</i> value
N	18,360	17,028	1,332	
Representing individuals	71,361,366	67,325,864	4,035,502	
Age	56.652 ± 0.214	56.085 ± 0.217	66.118±0.456	<0.0001
PIR	3.003 ± 0.031	3.041 ± 0.031	2.368 ± 0.066	< 0.0001
BMI, kg/m ²	30.837 ± 0.080	30.864 ± 0.082	30.396 ± 0.272	0.09
Lymphocyte, 1000 cells/µl	2.153±0.017	2.157 ± 0.017	2.084 ± 0.046	0.144
Monocyte, 1000 cells/μl	0.589 ± 0.003	0.587 ± 0.003	0.618 ± 0.008	<0.001
Segmented neutrophils, 1000 cells/µl	4.418 ± 0.022	4.410 ± 0.023	4.549 ± 0.060	0.025
Platelet, 1000 cells/μL	253.957 ± 1.014	254.493 ± 0.987	245.010 ± 3.430	0.004
SII	587.718 ± 4.276	585.218 ± 4.300	629.437 ± 17.283	0.012
Ln (SII)	6.228 ± 0.007	6.226 ± 0.007	6.264 ± 0.020	0.077
PIV	353.577 ± 3.543	351.382 ± 3.580	390.204±11.182	< 0.001
Ln (PIVI)	5.642 ± 0.009	5.638 ± 0.010	5.717 ± 0.026	0.003
SIRI	1.379 ± 0.012	1.366 ± 0.012	1.587 ± 0.043	<0.0001
Ln (SIRI)	0.143 ± 0.009	0.136 ± 0.009	0.261 ± 0.024	<0.0001
Sex				< 0.001
Male	9,193 (49.149)	8,538 (49.499)	655 (43.312)	
Female	9,167 (50.851)	8,490 (50.501)	677 (56.688)	
Race				0.135
Mexican American	2,472 (5.237)	2,341 (5.321)	131 (3.851)	
Non-Hispanic Black	4,583 (12.330)	4,212 (12.212)	371 (14.294)	
Non-Hispanic White	8,623 (72.657)	7,942 (72.648)	681 (72.797)	
Other Hispanic	1,312 (4.354)	1,235 (4.401)	77 (3.581)	
Other race	1,370 (5.422)	1,298 (5.418)	72 (5.478)	
Marital status				<0.0001
Non-single	10,880 (64.741)	10,180 (65.171)	700 (57.568)	
Single	7,480 (35.259)	6,848 (34.829)	632 (42.432)	
Education				<0.0001
<high school<="" td=""><td>2,336 (6.690)</td><td>2,132 (6.468)</td><td>204 (10.388)</td><td></td></high>	2,336 (6.690)	2,132 (6.468)	204 (10.388)	
High school	7,282 (39.070)	6,674 (38.563)	608 (47.525)	
>High school	8,742 (54.240)	8,222 (54.969)	520 (42.088)	
Smoke				< 0.001
Never	9,165 (49.619)	8,631 (50.070)	534 (42.100)	
Former	5,784 (31.860)	5,268 (31.598)	516 (36.233)	
Now	3,411 (18.521)	3,129 (18.332)	282 (21.666)	

(Continued)

TABLE 1 (Continued)

	Total	No-stroke	Stroke	<i>p</i> value
Drinking				<0.0001
Never	2,760 (12.472)	2,535 (12.116)	225 (18.406)	
Former	3,827 (18.719)	3,400 (18.049)	427 (29.886)	
Mild	6,569 (37.871)	6,153 (38.195)	416 (32.466)	
Moderate	2,397 (14.818)	2,265 (15.155)	132 (9.184)	
Heavy	2,807 (16.121)	2,675 (16.484)	132 (10.057)	
Physical activity				<0.0001
No	10,290 (49.599)	9,426 (48.849)	864 (62.228)	
Moderate	4,561 (28.163)	4,267 (28.347)	294 (25.159)	
Vigorous	3,509 (22.226)	3,336 (22.804)	173 (12.613)	
Diabetes				<0.0001
No	12,933 (76.732)	12,152 (77.644)	781 (61.522)	
Yes	5,427 (23.268)	4,876 (22.356)	551 (38.478)	
Coronary heart disease				<0.0001
No	16,917 (92.774)	15,845 (93.651)	1,072 (78.149)	
Yes	1,443 (7.226)	1,183 (6.349)	260 (21.851)	

PIR, Poverty income ratio; BMI, body mass index; Ln, natural logarithmic; SII, systemic immune-inflammation index; PIV, pan-immune-inflammation value; SIRI, systemic inflammation response index. The bold values of "N" means Unweighted number; bold values of "Representing individuals" means Weighted number.

3.2 Association of SII, PIV, and SIRI with stroke prevalence in hypertensive populations

In multivariate logistic regression analyses, we found that SII was not significantly associated with the odds of stroke among hypertensive patients in all adjusted models. In fully adjusted model 3, we found that PIV was significantly and positively associated with the prevalence of stroke in the hypertensive population. Each unit increase in lnPIV was associated with a 14% increase in the odds of stroke (odds ratio [OR] = 1.140, 95% CI = 1.048–1.241, p = 0.0022). Higher PIV was significantly associated with an increased prevalence of stroke (p for trend = 0.0056), and hypertensive patients with PIV at Q4 had significantly increased odds of stroke compared to Q1 (OR and 95% CI = 1.235 (1.048, 1.456), p = 0.012). Similarly, SIRI was positively associated with stroke prevalence after adjusting for all confounders (OR = 1.206, p = 0.0144). As SIRI increased, the odds of stroke increased significantly (p for trend = 0.0244). Interestingly, SIRI was significantly inversely associated with stroke prevalence only at Q2 (OR=0.767, p = 0.0456) (Table 2).

3.3 RCS, segmented regression and correlation matrix analysis

Consistently, RCS analyses showed that lnSII was not significantly associated with stroke prevalence in the hypertensive population (Figure 2A). Both lnPIV (p for nonlinear = 0.047) (Figure 2B) and lnSIRI

(p for nonlinear = 0.0029) (Figure 2C) were nonlinearly associated with the odds of stroke and showed J-shaped associations. Segmented regression analysis indicated that lnPIV was significantly and positively associated with the prevalence of stroke in hypertensive populations at >5.1 (OR = 1.325, p = 0.0002) and not significantly associated before the inflection point, although p for interaction was not significant. Similarly, lnSIRI was positively associated with stroke likelihood at > -0.25 (OR = 1.497, p < 0.0001) and not significantly associated before the inflection point (p for interaction = 0.0415) (Table 3).

The correlation matrix among the three systemic inflammatory markers (SII, PIV, and SIRI) is as follows: There is a strong positive correlation between SII and PIV, with a correlation coefficient of 0.8634 (95% CI, 0.8596–0.8670, p = 0.0000), indicating a statistically significant correlation. Between SII and SIRI, there is a moderate positive correlation, with a correlation coefficient of 0.6968 (95% CI, 0.6893–0.7042, p = 0.0000), also statistically significant. Finally, PIV and SIRI show a strong positive correlation, with a correlation coefficient of 0.8376 (95% CI, 0.8332–0.8418, p = 0.0000), underscoring a statistically significant relationship (Supplementary Table S1 and Supplementary Figure S1).

3.4 Stratified analysis

We performed stratified analyses of the association of lnPIV and InSIRI with stroke prevalence in hypertensive populations. Significant associations between lnPIV and stroke prevalence were only present in people ≥60 years of age, men, non-Hispanic whites, non-singles, <high school diploma, PIR1-3, BMI <25 or ≥ 30 kg/m², no or vigorous physical activity, quitters, abstainers/light drinkers, and those without diabetes and CHD. However, only gender and smoking were identified as significant effect modifiers (Figure 3). Subgroup analyses in the association of lnSIRI with stroke prevalence yielded broadly similar results. Smoking was the only effect modifier (Figure 4). A univariate analysis of the association between smoking and elevated levels of PIV, SIRI, and SII in individuals with hypertension was performed. For SII, former smokers have a β value of 16.324 (p = 0.0524), and current smokers have a β value of 26.300 (p = 0.0044). For PIV, former smokers have a β value of 31.643 (p < 0.0001), and current smokers have a β value of 57.834 (p < 0.0001). For SIRI, former smokers have a β value of 0.160 (p < 0.0001), and current smokers have a β value of 0.183 (p < 0.0001) (Supplementary Table S2). These results indicate a significant association between smoking and elevated levels of these indices, particularly in current smokers.

4 Discussion

This study represents the pioneering exploration of the relationship between SII, PIV, and SIRI levels and the incidence of stroke within a hypertensive cohort. The key findings of this investigation are outlined as follows: (1) The natural logarithm of SII (lnSII) exhibited no significant correlation with baseline clinical characteristics, suggesting its role as a relatively stable inflammation marker. Conversely, lnPIV and lnSIRI levels were notably elevated in stroke patients. (2) Within the hypertensive population, elevated levels of PIV and SIRI were strongly linked to a heightened prevalence of stroke, with the exception of SIRI, which

TABLE 2 Survey-weighted logistic regression examining the association of SII, PIV, SIRI with the prevalence of stroke among individuals with hypertension from NHANES 1999–2020.

SII 1.000 (1.000, 1.000) (1.000, 1.000) (1.000, 1.000) (1.000)		Model 1	Model 2	Model 3
SII Log 1.136 (0.986, 1.309) 0.0785 1.275) 0.1533 1.078 (0.933, 1.244) 0.3102 SII quartile Q1 Ref. Ref. Ref. Q2 1.120 (0.880, 1.200 (0.926, 1.189 (0.907, 1.425) 0.3580 1.557) 0.1699 1.559) 0.2127 Q3 0.853 (0.666, 0.893 (0.685, 1.092) 0.2093 1.164) 0.4032 1.140) 0.3108 Q4 1.210 (0.978, 1.197 (0.946, 1.154 (0.905, 1.497) 0.0811 1.516) 0.1366 1.473) 0.2502 P for trend 0.3011 0.4612 0.6994 PIV 1.000 (1.000, 1.000 (1.000, 1.000 (1.000, 1.000) (1.000) (1.000, 1.001) 0.0014 1.000) 0.0043 1.000) 0.0147 PIV Log 1.196 (1.066, 1.181 (1.086, 1.140 (1.048, 1.342) 0.0027 1.283) 0.0001 1.241) 0.0022 PIV quartile Q1 Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 0.925 (0.779, 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 0.979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.004 <td>SII</td> <td>1.000 (1.000,</td> <td>1.000 (1.000,</td> <td>1.000 (1.000,</td>	SII	1.000 (1.000,	1.000 (1.000,	1.000 (1.000,
SII quartile		1.000) 0.0080	1.000) 0.0458	1.000) 0.1076
SII quartile Ref. Ref. Ref. Q2 1.120 (0.880, 1.200 (0.926, 1.189 (0.907, 1.425) 0.3580 1.557) 0.1699 1.559) 0.2127 0.3 0.853 (0.666, 0.893 (0.685, 1.140) 0.3108 0.868 (0.662, 1.092) 0.2093 1.164) 0.4032 1.140) 0.3108 Q4 1.210 (0.978, 1.197 (0.946, 1.154 (0.905, 1.497) 0.0811 1.516) 0.1366 1.473) 0.2502 1.6994 Pfor trend 0.3011 0.4612 0.6994 0.6994 PIV 1.000 (1.000, 1.000 (1.000, 1.000 (1.000, 1.000) (1.000) 0.0043 1.000) 0.0147 1.001 (0.0011 1.000) 0.0043 1.000) 0.0147 PIV Log 1.196 (1.066, 1.181 (1.086, 1.140 (1.048, 1.342) 0.0027 1.283) 0.0001 1.241) 0.0022 PIV quartile Q1 Q1 Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 0.925 (0.779, 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 1.023 (0.0758, 1.008 (0.851, 0.979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0004 0.0056 0.0066 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021	SII Log	1.136 (0.986,	1.108 (0.963,	1.078 (0.933,
Q1 Ref. Ref. Ref. Q2 1.120 (0.880, 1.200 (0.926, 1.189 (0.907, 1.559) 0.2127 1.425) 0.3580 1.557) 0.1699 1.559) 0.2127 Q3 0.853 (0.666, 0.893 (0.685, 1.092) 0.2093 1.164) 0.4032 1.140) 0.3108 Q4 1.210 (0.978, 1.197 (0.946, 1.154 (0.905, 1.497) 0.0811 1.516) 0.1366 1.473) 0.2502 P for trend 0.3011 0.4612 0.6994 PIV 1.000 (1.000, 1.000 (1.000, 1.000 (1.000, 1.000) (1.000, 1.001) 0.0011 1.000) 0.0043 1.000) 0.0147 PIV Log 1.196 (1.066, 1.181 (1.086, 1.140 (1.048, 1.342) 0.0027 1.283) 0.0001 1.241) 0.0022 PIV quartile Q1 Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 1.098) 0.3709 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 0.979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.648, 1.245) <0.0001		1.309) 0.0785	1.275) 0.1533	1.244) 0.3102
Q2	SII quartile			
1.425) 0.3580	Q1	Ref.	Ref.	Ref.
Q3	Q2	1.120 (0.880,	1.200 (0.926,	1.189 (0.907,
1.092) 0.2093		1.425) 0.3580	1.557) 0.1699	1.559) 0.2127
Q4 1.210 (0.978, 1.497) 0.0811 1.197 (0.946, 1.473) 0.2502 1.154 (0.905, 1.473) 0.2502 P for trend 0.3011 0.4612 0.6994 PIV 1.000 (1.000, 1.000 (1.000, 1.000) (1.000, 1.000) 0.0147 1.000) 0.0043 1.000) 0.0147 PIV Log 1.196 (1.066, 1.181 (1.086, 1.140 (1.048, 1.342) 0.0027 1.283) 0.0001 1.241) 0.0022 PIV quartile Q1 Ref. Ref. Ref. Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 0.925 (0.779, 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 0.979 (0.825, 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.632) <0.0001	Q3	0.853 (0.666,	0.893 (0.685,	0.868 (0.662,
P for trend 0.3011 0.4612 0.6994 PIV 1.000 (1.000, 1.000 (1.000, 1.000 (1.000, 1.000) 0.0043 PIV Log 1.196 (1.066, 1.181 (1.086, 1.140 (1.048, 1.342) 0.0027 PIV quartile Q1 Ref. Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 0.925 (0.779, 1.022) 0.0759 1.248) 0.8275 1.194) 0.9244 1.632) 0.0347 Piv trend 0.040 0.0004 0.0004 Pfor trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.295 (1.040, 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 Ref. Ref. Ref. Ref. Ref. Ref. Ref. Ref. Ref. 0.979 (0.825, 1.194) 0.9244 1.1610 0.0021 P for trend 0.0040 0.0004 0.0056 SIRI Comparison of the first		1.092) 0.2093	1.164) 0.4032	1.140) 0.3108
P for trend 0.3011 0.4612 0.6994 PIV 1.000 (1.000, 1.000 (1.000, 1.000 (1.000, 1.001) 0.0011 1.000) 0.0043 1.000) 0.0147 PIV Log 1.196 (1.066, 1.181 (1.086, 1.140 (1.048, 1.342) 0.0027 1.283) 0.0001 1.241) 0.0022 PIV quartile Q1 Ref. Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 0.925 (0.779, 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 0.979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603	Q4	1.210 (0.978,	1.197 (0.946,	1.154 (0.905,
PIV		1.497) 0.0811	1.516) 0.1366	1.473) 0.2502
1.001) 0.0011 1.000) 0.0043 1.000) 0.0147	P for trend	0.3011	0.4612	0.6994
PIV Log 1.196 (1.066, 1.181 (1.086, 1.140 (1.048, 1.342) 0.0027 1.283) 0.0001 1.241) 0.0022 PIV quartile Q1 Ref. Ref. Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 0.979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603	PIV	1.000 (1.000,	1.000 (1.000,	1.000 (1.000,
PIV quartile Q1 Ref. Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 0.979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603		1.001) 0.0011	1.000) 0.0043	1.000) 0.0147
PIV quartile Q1 Ref. Ref. Ref. Ref. Q2 0.797 (0.622, 0.932 (0.787, 0.925 (0.779, 1.022) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 0.979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603	PIV Log	1.196 (1.066,	1.181 (1.086,	1.140 (1.048,
Q1 Ref. Ref. Ref. Q2 0.797 (0.622, 1.022) 0.0759 0.932 (0.787, 1.092) (0.779, 1.092) (0.779, 1.002) 0.0759 1.105) 0.4204 1.098) 0.3709 Q3 0.973 (0.758, 1.008 (0.851, 1.0979 (0.825, 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001		1.342) 0.0027	1.283) 0.0001	1.241) 0.0022
Q2	PIV quartile			
1.022) 0.0759 1.105) 0.4204 1.098) 0.3709	Q1	Ref.	Ref.	Ref.
Q3	Q2	0.797 (0.622,	0.932 (0.787,	0.925 (0.779,
Q4 1.248) 0.8275 1.194) 0.9244 1.162) 0.8092 Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001		1.022) 0.0759	1.105) 0.4204	1.098) 0.3709
Q4 1.251 (1.018, 1.311 (1.115, 1.235 (1.048, 1.539) 0.0347 1.543) 0.0011 1.456) 0.0120 P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001	Q3	0.973 (0.758,	1.008 (0.851,	0.979 (0.825,
P for trend 0.0040 0.0004 0.00056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603		1.248) 0.8275	1.194) 0.9244	1.162) 0.8092
P for trend 0.0040 0.0004 0.0056 SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603	Q4	1.251 (1.018,	1.311 (1.115,	1.235 (1.048,
SIRI 1.183 (1.124, 1.111 (1.050, 1.096 (1.035, 1.245) <0.0001 1.174) 0.0003 1.161) 0.0021 SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) <0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603		1.539) 0.0347	1.543) 0.0011	1.456) 0.0120
1.245) <0.0001	P for trend	0.0040	0.0004	0.0056
SIRI Log 1.431 (1.255, 1.255 (1.087, 1.206 (1.040, 1.632) < 0.0001 1.449) 0.0023 1.400) 0.0144 SIRI quartile Q1 Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) < 0.0001 1.675) 0.0547 1.580) 0.1603	SIRI	1.183 (1.124,	1.111 (1.050,	1.096 (1.035,
1.632) <0.0001		1.245) < 0.0001	1.174) 0.0003	1.161) 0.0021
SIRI quartile Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) < 0.0001	SIRI Log	1.431 (1.255,	1.255 (1.087,	1.206 (1.040,
Q1 Ref. Ref. Ref. Ref. Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 1.011) 0.0626 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) <0.0001 1.675) 0.0547 1.580) 0.1603		1.632) < 0.0001	1.449) 0.0023	1.400) 0.0144
Q2 0.787 (0.622, 0.782 (0.604, 0.767 (0.593, 0.996) 0.0475 0.782 (0.604, 0.993) 0.0456 Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) < 0.0001	SIRI quartile			
Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) < 0.0001 1.675) 0.0547 1.580) 0.1603	Q1	Ref.	Ref.	Ref.
Q3 1.071 (0.847, 0.976 (0.751, 0.930 (0.716, 1.354) 0.5699 1.267) 0.8539 1.208) 0.5883 Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) < 0.0001 1.675) 0.0547 1.580) 0.1603	Q2	0.787 (0.622,	0.782 (0.604,	0.767 (0.593,
Q4 1.571 (1.271, 1.292 (0.997, 1.940) < 0.0001		0.996) 0.0475	1.011) 0.0626	0.993) 0.0456
Q4 1.571 (1.271, 1.292 (0.997, 1.211 (0.928, 1.940) < 0.0001 1.675) 0.0547 1.580) 0.1603	Q3	1.071 (0.847,	0.976 (0.751,	0.930 (0.716,
1.940) <0.0001 1.675) 0.0547 1.580) 0.1603		1.354) 0.5699	1.267) 0.8539	1.208) 0.5883
	Q4	1.571 (1.271,	1.292 (0.997,	1.211 (0.928,
P for trend <0.0001 0.0048 0.0244		1.940) < 0.0001	1.675) 0.0547	1.580) 0.1603
	P for trend	<0.0001	0.0048	0.0244

Model 1 did not adjust for any covariates.

Model 2 adjusted age, gender, race, PIR, education level, and marital status.

Model 3 adjusted for all covariates.

demonstrated a significant inverse correlation with stroke prevalence solely at the second quartile (Q2). (3) Sex differences and smoking status emerged as significant effect modifiers in the relationship between lnPIV and stroke, while smoking status was the sole effect modifier in the association between lnSIRI and stroke within the hypertensive cohort.

The crucial role of chronic inflammation in stroke is wellestablished, and hypertension is known to related with inflammation and immunity, implying that inflammation is a common background for both diseases (31-33). In the hypertensive population, there are distinct characteristics related to inflammation and the risk of stroke. Hypertension induces various alterations in the body's blood vessels. These changes include adaptive remodeling, hypertrophy, and stiffness of vascular walls, as well as a decrease in vessel diameter, leading to increased vascular resistance and circulatory disruptions (34, 35). The brain is particularly vulnerable to circulatory changes associated with hypertension, impacting key mechanisms that regulate cerebral blood flow and disrupt brain energy balance, ultimately affecting the function of neurovascular units (35, 36). Systemic and neuroinflammation contribute to hypertension pathophysiology by inducing reactive oxygen species (ROS) production and cell damage, triggering the release of damage-associated molecular patterns (DAMPs) that activate Toll-like receptor 4 (TLR4) (35, 37). Activation of endothelial TLR4, via MyD88 protein, stimulates transcription factors AP-1 and NF-kB, exacerbating inflammation in vascular tissues (35, 38, 39). Patients with hypertension often exhibit elevated levels of inflammatory markers, including C-reactive protein and IL-6, which have been linked to a higher risk of stroke (35, 40, 41). Immune cells in the brain, including microglia/macrophages, are impacted by hypertension, leading to their activation and expression of pro-inflammatory molecules like IL-1b, IL-6, and TNF-a. Perivascular macrophages (PVMs) near arterioles and venules play a crucial role in neurovascular unit dysfunction, with studies showing their involvement in blood-brain barrier (BBB) disruption (42, 43). CD36 receptor expression in microglia is elevated in hypertensive conditions, contributing to BBB lesions and astrocyte activation post-stroke, exacerbating inflammatory processes (44, 45). Although no studies have directly compared systemic inflammatory markers (PIV, SIRI, SII) between hypertensive and non-hypertensive patients, a study of non-dipper hypertensive patients found their SII levels to be higher than those of dipper hypertensive patients (46). Similarly, another study found a U-shaped relationship between the SII levels and the risk of hypertension in US adults, suggesting a complex role for SII in hypertension (47).

A newly proposed inflammatory biomarker, known as the Pan-Immune-Inflammation Value (PIV), incorporates neutrophil, monocyte, platelet, and lymphocyte counts to provide a more comprehensive assessment of the systemic immune inflammatory response (17, 48). The Systemic Inflammation Response Index (SIRI), comprising platelets and three subtypes of white blood cells, has been identified as a significant correlate of cardiovascular disease (CVD). Specifically, SIRI is recognized as a superior marker of chronic inflammation and demonstrates strong prognostic predictive value for patients with acute strokes and tumors (24, 49, 50). In our study, there strong correlation among PIV, SIRI, and (Supplementary Table S1 and Supplementary Figure S1). They are all similar systemic inflammation indicators derived from whole blood cells, but the composition of inflammatory cells in their respective calculation formulas differs, which may lead to some being linear and others nonlinear association between stroke prevalence. The immune and inflammatory response is a common process in the clinical manifestations of cardiac and cerebral acute ischemia following atherothrombosis (48, 51, 52). Following cerebral ischemic injury, damage-associated molecular patterns (DAMPs) released by necrotic

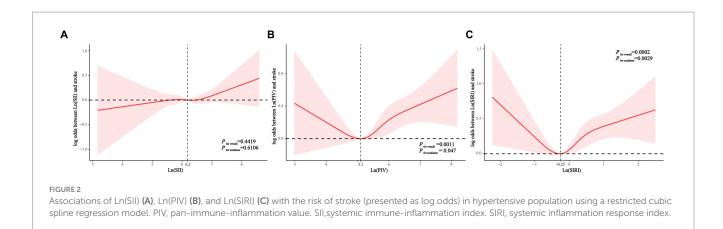


TABLE 3 Segmented regression analysis of InPIV and InSIRI with the prevalence of stroke in hypertensive populations.

	Ln (SIRI) < = - 0.25	Ln (SIRI) > -0.25	P-interaction
Ln	0.855 (0.523, 1.399)	1.497 (1.268, 1.768)	0.0415
(SIRI)	0.5343	<0.0001	
	Ln (PIV) <=5.1	Ln (PIV)>5.1	
Ln	1.036 (0.671, 1.599)	1.325 (1.145, 1.533)	0.2914
(PIV)	0.8729	0.0002	

cells trigger the activation of resident immune cells within the central nervous system, including microglia and astrocytes. This activation then leads to the recruitment of peripheral immune cells to initiate adaptive immune responses (48, 53). In our study, a robust correlation was observed between PIV, SIRI, and the prevalence of stroke among former smokers when compared to non-smokers and current smokers. Cigarette smoking elevates the risk of cardiovascular events through two primary mechanisms: an atherosclerotic effect and a prothrombotic effect. Smoking compromises vasodilation and triggers inflammation, ultimately culminating in the development of atherosclerosis (54, 55). In the recent research, Darragh Duffy from the Pasteur Institute in France, specifically examines the impact of smoking on innate and adaptive immune responses, elucidating both the short-term and long-term regulatory effects of smoking on immune reactions based on the analysis of the Milieu Intérieur cohort (56). To further assess the impact of smoking on the immune system, the authors constructed a model to evaluate the level of smoking effects. The results revealed that current smoking influences both innate and adaptive immune responses, leading to heightened inductions of CXCL5, IL-2, and IL-13. Compared to non-smokers, former smokers did not exhibit a significant increase in CXCL5 secretion following innate immune stimulation; however, the secretion of IL-2 and IL-13 increased after adaptive immune stimulation. These findings suggest that smoking exerts a short-term impact on innate immunity but has long-lasting effects on adaptive immune responses (56). The result of our study might be explained by the fact that former smoker status changed the adaptive immune responses and led to the higher level of inflammation. This suggests that the avoidance of initial exposure to tobacco is crucial for enhancing long-term immunity, thereby aiding in the reduction of stroke risk within the hypertensive population.

In the present study, a robust correlation between PIV and the prevalence of stroke in hypertension population was observed in males compared to females. The overall incidence of stroke in men is estimated to be 33% higher than in women (57). This malepredominant trend is also evident in the perinatal, neonatal, and pediatric populations, with males exhibiting an elevated risk for both hemorrhagic and ischemic strokes (58-60). Innate factors such as host genetics and chromosomal sex are pivotal in influencing both the host immune system and the neuroimmune response to brain injury. Ischemic stroke disrupts intracellular communication among astrocytes, neurons, and resident immune cells within the central nervous system (CNS). Elevated cytokine and chemokine production coordinates the recruitment of peripheral immune cells and facilitates neuroinflammation (61). Female hormones, particularly estrogen, may exert a significant role in neuroprotection following ischemic events. Estrogen, produced within the brain, functions as a neuroprotective and anti-inflammatory agent post-stroke. As a sex steroid and neurosteroid hormone, estrogen's neuroprotective properties involve the modulation of the immunological response that arises following ischemic brain injury (62-64). Sex differences play a significant role in the occurrence and outcomes of ischemic stroke, with immune regulation based on sex being a key factor. Discrepancies in stroke incidence and prognosis may also be impacted by sex-specific risk factors, including the use of oral contraceptives and menopausal status (60, 65). Unfortunately, despite the recognition of gender disparities as a crucial risk factor in numerous studies, gender-specific stroke treatments have not yet been developed.

Our research investigated the occurrence of stroke in the hypertensive population of the United States, a demographic that has received limited attention in prior studies. This study offers important insights into the interplay between immune response, inflammation and stroke among individuals with hypertension. In our analysis, we found that the relationship between PIV and SIRI and stroke risk was non-linear, with former smokers and males experiencing an increased risk. These findings have significant implications for early identification of high-risk individuals and guiding recommendations for stroke prevention in hypertensive patients. By using these results, we can develop personalized prevention strategies, lifestyle interventions, and potential therapeutic targets to reduce inflammation in the body. Among the findings is an emphasis on the risk of stroke associated with smoking, reinforcing the importance of smoking cessation interventions and advocating for not initiating tobacco use.

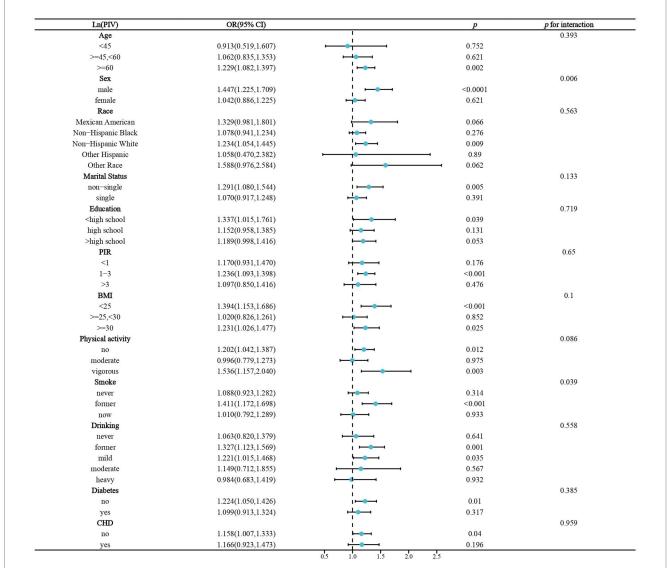


FIGURE 3
Stratified analysis of the association of Ln(PIV) and stroke prevalence in hypertensive populations. BMI, body mass index. CHD, coronary heart disease. OR, odds ratio. PIR, poverty income ratio.

Moreover, these insights can assist public health policymakers in designing more effective stroke prevention programs for hypertensives, including specific interventions for high-risk groups such as men and former smokers. Considering estrogen supplementation for postmenopausal women is also suggested as a potential preventive measure. Nevertheless, our study has several limitations that warrant consideration. Firstly, a few limitations of the NHANES self-reported diagnosis include nonresponse bias, measurement errors, subjectivity, a complex survey design, and diagnostic tests that are prone to errors. A thorough consideration is needed when interpreting results from NHANES-based studies due to the possibility of biases and inaccuracies being introduced (66, 67). Secondly, it is crucial to acknowledge that our study design was cross-sectional, precluding the establishment of causal relationships between PIV, SIRI levels, and the prevalence of stroke. In our study, it can only be said that the inflammation levels in individuals with hypertension combined with stroke are higher than those without stroke in the hypertensive population, but it cannot be concluded whether inflammation is caused by hypertension or stroke. Future prospective studies are essential to validate these associations. Thirdly, despite our meticulous adjustment for potential confounding variables in the regression analysis, the influence of residual factors, such as autoimmune diseases, on the outcomes cannot be entirely eliminated. Fourthly, the NHANES database relies on standardized questionnaires during home visits to collect participants' medical histories, limiting our ability to delve into the specific subtypes of hypertension and stroke. Consequently, the potential impact of hypertension phenotypes and stroke severity on the observed outcomes remains uncertain. Fifthly, the systemic inflammatory markers (SII, PIV, and SIRI) can potentially vary depending on the timing of blood collection. Our research has limitations regarding the specific details of blood collection timing in NHANES, but it generally involves collecting biospecimens as part of its comprehensive health assessment surveys. The specific timing of these collections could vary depending on the survey cycle and the

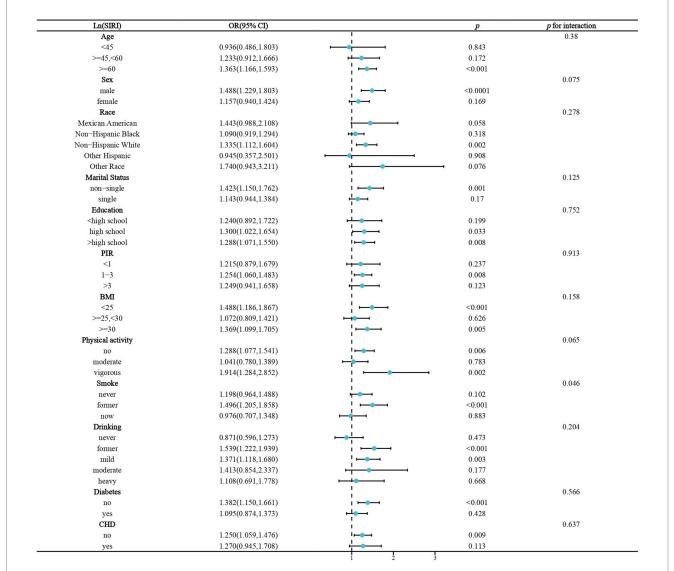


FIGURE 4

Stratified analysis of the association of Ln(SIRI) and stroke prevalence in hypertensive populations. BMI, body mass index. CHD, coronary heart disease. OR, odds ratio.PIR, poverty income ratio.

particular health or nutritional focus of the study at any given time (68). Notably, all participants' blood samples were collected under standardized conditions at the NHANES Mobile Examination Center (MEC). All blood samples were analyzed by professional technicians to ensure the consistency and reliability of the data (69). Furthermore, our study was confined to participants from the United States, potentially limiting the generalizability of our findings to populations with distinct risk profiles and health behaviors. Additionally, variations in treatment approaches across different generations could have influenced our results. Despite these constraints, it is imperative to recognize the need for further investigations to elucidate the roles of PIV and SIRI in the context of hypertension and stroke prevalence. Such endeavors will enhance our comprehension of the shared pathophysiological mechanisms underlying hypertension and stroke, paving the way for the development of more efficacious treatment modalities.

5 Conclusion

Elevated levels of PIV and SIRI in individuals with hypertension are associated with a notable rise in stroke prevalence, especially among males and former smokers. Furthermore, a predictive clinical model encompassing PIV, SIRI, age, gender, race, education, marital status, and poverty income ratio demonstrates strong prognostic utility for estimating stroke prevalence within the hypertensive population, offering valuable clinical insights.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary material, further inquiries can be directed to the corresponding author.

Ethics statement

The studies involving humans were approved by the National Center for Health Statistics Institutional Review Board. The studies were conducted in accordance with the local legislation and institutional requirements. The participants provided their written informed consent to participate in this study. Written informed consent was obtained from the individual(s) for the publication of any potentially identifiable images or data included in this article.

Author contributions

JC: Formal analysis, Investigation, Writing – original draft. CL: Formal analysis, Investigation, Writing – original draft. DT: Methodology, Writing – review & editing. YL: Formal analysis, Methodology, Writing – review & editing.

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References

- 1. Campbell B, Khatri P. Stroke, Stroke. Lancet. (2020) 396:129-42. doi: 10.1016/S0140-6736(20)31179-X
- 2. Campbell B, De Silva DA, Macleod MR, Coutts SB, Schwamm LH, Davis SM, et al. Ischaemic stroke. *Nat Rev Dis Primers*. (2019) 5:70. doi: 10.1038/s41572-019-0118-8
- 3. GBD 2019 Stroke Collaborators. Global, regional, and national burden of stroke and its risk factors, 1990-2019: a systematic analysis for the global burden of disease study 2019. *Lancet Neurol.* (2021) 20:795–820. doi: 10.1016/S1474-4422(21)00252-0
- 4. Virani SS, Alonso A, Benjamin EJ, Bittencourt MS, Callaway CW, Carson AP, et al. Heart disease and stroke statistics-2020 update: a report from the American Heart Association. *Circulation*. (2020) 141:e139–596. doi: 10.1161/CIR.00000000000000757
- 5. Feske SK. Ischemic Stroke. Am J Med. (2021) 134:1457–64. doi: 10.1016/j.amjmed.2021.07.027
- 6. Boehme AK, Esenwa C, Elkind MS. Stroke risk factors, genetics, and prevention. Circ Res. (2017) 120:472–95. doi: 10.1161/CIRCRESAHA.116.308398
- 7. Oparil S, Acelajado MC, Bakris GL, Berlowitz DR, Cífková R, Dominiczak AF, et al. Hypertension. *Nat Rev Dis Primers*. (2018) 4:18014. doi: 10.1038/nrdp.2018.14
- 8. Buonacera A, Stancanelli B, Malatino L. Stroke and hypertension: an appraisal from pathophysiology to clinical practice. *Curr Vasc Pharmacol.* (2019) 17:72–84. doi: 10.217 4/1570161115666171116151051
- 9. Dubow J, Fink ME. Impact of hypertension on stroke. Curr Atheroscler Rep. (2011) 13:298–305. doi: 10.1007/s11883-011-0187-y
- 10. Cipolla MJ, Liebeskind DS, Chan SL. The importance of comorbidities in ischemic stroke: impact of hypertension on the cerebral circulation. *J Cereb Blood Flow Metab.* (2018) 38:2129–49. doi: 10.1177/0271678X18800589
- 11. Low A, Mak E, Rowe JB, Markus HS, O'Brien JT. Inflammation and cerebral small vessel disease: a systematic review. *Ageing Res Rev.* (2019) 53:100916. doi: 10.1016/j. arr.2019.100916
- 12. Madhur MS, Elijovich F, Alexander MR, Pitzer A, Ishimwe J, van Beusecum JP, et al. Hypertension: do inflammation and immunity hold the key to solving this epidemic. *Circ Res.* (2021) 128:908–33. doi: 10.1161/CIRCRESAHA.121.318052
- 13. Sylvester MA, Brooks HL. Sex-specific mechanisms in inflammation and hypertension. *Curr Hypertens Rep.* (2019) 21:53. doi: 10.1007/s11906-019-0959-2
- 14. Deussen A, Kopaliani I. Targeting inflammation in hypertension. Curr Opin Nephrol Hypertens. (2023) 32:111–7. doi: 10.1097/MNH.0000000000000862
- 15. Xiao L, Harrison DG. Inflammation in hypertension. Can J Cardiol. (2020) 36:635–47. doi: 10.1016/j.cjca.2020.01.013

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fneur.2024.1417863/full#supplementary-material

SUPPLEMENTARY FIGURE S1

The correlation matrix analysis among the three systemic inflammatory markers (SII, PIV, and SIRI).

- 16. Hu B, Yang XR, Xu Y, Sun YF, Sun C, Guo W, et al. Systemic immune-inflammation index predicts prognosis of patients after curative resection for hepatocellular carcinoma. *Clin Cancer Res.* (2014) 20:6212–22. doi: 10.1158/1078-0432.CCR-14-0442
- 17. Fucà G, Guarini V, Antoniotti C, Morano F, Moretto R, Corallo S, et al. The panimmune-inflammation value is a new prognostic biomarker in metastatic colorectal cancer: results from a pooled-analysis of the Valentino and TRIBE first-line trials. *Br J Cancer*. (2020) 123:403–9. doi: 10.1038/s41416-020-0894-7
- 18. Qi Q, Zhuang L, Shen Y, Geng Y, Yu S, Chen H, et al. A novel systemic inflammation response index (SIRI) for predicting the survival of patients with pancreatic cancer after chemotherapy. *Cancer*. (2016) 122:2158–67. doi: 10.1002/cncr.30057
- 19. Jin N, Huang L, Hong J, Zhao X, Hu J, Wang S, et al. The association between systemic inflammation markers and the prevalence of hypertension. *BMC Cardiovasc Disord*. (2023) 23:615. doi: 10.1186/s12872-023-03661-6
- 20. Cao Y, Li P, Zhang Y, Qiu M, Li J, Ma S, et al. Association of systemic immune inflammatory index with all-cause and cause-specific mortality in hypertensive individuals: results from NHANES. *Front Immunol.* (2023) 14:1087345. doi: 10.3389/fimmu.2023.1087345
- 21. Zhao S, Dong S, Qin Y, Wang Y, Zhang B, Liu A. Inflammation index SIRI is associated with increased all-cause and cardiovascular mortality among patients with hypertension. *Front Cardiovasc Med.* (2022) 9:1066219. doi: 10.3389/fcvm.2022.1066219
- 22. Wu B, Zhang C, Lin S, Zhang Y, Ding S, Song W. The relationship between the pan-immune-inflammation value and long-term prognoses in patients with hypertension: National Health and nutrition examination study, 1999-2018. *Front Cardiovasc Med.* (2023) 10:1099427. doi: 10.3389/fcvm.2023.1099427
- 23. Aydin C, Alpsoy Ş, Akyüz A, Özkaramanli Gür D, Emlek N, Şahin A, et al. Could the systemic immune-inflammation index be a predictor to estimate cerebrovascular events in hypertensive patients. *Blood Press Monit.* (2022) 27:33–8. doi: 10.1097/MBP.000000000000560
- 24. Cai X, Song S, Hu J, Wang L, Shen D, Zhu Q, et al. Systemic inflammation response index as a predictor of stroke risk in elderly patients with hypertension: a cohort study. *J Inflamm Res.* (2023) 16:4821–32. doi: 10.2147/JIR.S433190
- 25. Chen TC, Clark J, Riddles MK, Mohadjer LK, Fakhouri T. National Health and nutrition examination survey, 2015-2018: sample design and estimation procedures. *Vital Health Stat.* (2020) 2:1–35.
- 26. Ahluwalia N, Dwyer J, Terry A, Moshfegh A, Johnson C. Update on NHANES dietary data: focus on collection, release, analytical considerations, and uses to inform public policy. *Adv Nutr.* (2016) 7:121–34. doi: 10.3945/an.115.009258

- 27. Chen Y, Li Y, Liu M, Xu W, Tong S, Liu K. Association between systemic immunity-inflammation index and hypertension in US adults from NHANES 1999-2018. *Sci Rep.* (2024) 14:5677. doi: 10.1038/s41598-024-56387-6
- 28. Jiang M, Tang X, Wang P, Yang L, Du R. Association between daily alcohol consumption and serum alpha klotho levels among U.S. adults over 40 years old: a cross-sectional study. *BMC Public Health*. (2023) 23:1901. doi: 10.1186/s12889-023-16830-1
- $29.\ Yu\ L,$ Chen Y, Xu M, Li R, Zhang J, Zhu S, et al. Association of weight-adjusted-waist index with asthma prevalence and the age of first asthma onset in United States adults. Front Endocrinol. (2023) 14:1116621. doi: 10.3389/fendo.2023.1116621
- 30. Wang L, Li X, Wang Z, Bancks MP, Carnethon MR, Greenland P, et al. Trends in prevalence of diabetes and control of risk factors in diabetes among US adults, 1999-2018. *JAMA*. (2021) 326:1–13. doi: 10.1001/jama.2021.9883
- 31. Ungprasert P, Cheungpasitporn W, Wijarnpreecha K, Ahuja W, Ratanasrimetha P, Thongprayoon C. Risk of ischemic stroke in patients with polymyositis and dermatomyositis: a systematic review and meta-analysis. *Rheumatol Int.* (2015) 35:905–9. doi: 10.1007/s00296-014-3166-0
- 32. Ungprasert P, Sanguankeo A, Upala S. Risk of ischemic stroke in patients with systemic sclerosis: a systematic review and meta-analysis. *Mod Rheumatol.* (2016) 26:128–31. doi: 10.3109/14397595.2015.1056931
- 33. Agita A, Alsagaff MT. Inflammation, immunity, and hypertension. *Acta Med Indones*. (2017) 49:158–65.
- 34. Lee RM, Dickhout JG, Sandow SL. Vascular structural and functional changes: their association with causality in hypertension: models, remodeling and relevance. *Hypertens Res.* (2017) 40:311–23. doi: 10.1038/hr.2016.145
- 35. Przykaza Ł. Understanding the connection between common stroke comorbidities, their associated inflammation, and the course of the cerebral ischemia/reperfusion cascade. *Front Immunol.* (2021) 12:782569. doi: 10.3389/fimmu.2021.782569
- 36. Meissner A. Hypertension and the brain: a risk factor for more than heart disease. *Cerebrovasc Dis.* (2016) 42:255–62. doi: 10.1159/000446082
- 37. McCarthy CG, Goulopoulou S, Wenceslau CF, Spitler K, Matsumoto T, Webb RC. Toll-like receptors and damage-associated molecular patterns: novel links between inflammation and hypertension. *Am J Physiol Heart Circ Physiol.* (2014) 306:H184–96. doi: 10.1152/ajpheart.00328.2013
- 38. Bomfim GF, Santos RAD, Oliveira MA, Giachini FR, Akamine EH, Tostes RC, et al. Toll-like receptor 4 contributes to blood pressure regulation and vascular contraction in spontaneously hypertensive rats. *Clin Sci.* (2012) 122:535–43. doi: 10.1042/CS20110523
- 39. Yang Y, Lv J, Jiang S, Ma Z, Wang D, Hu W, et al. The emerging role of toll-like receptor 4 in myocardial inflammation. *Cell Death Dis.* (2016) 7:e2234. doi: 10.1038/cddis 2016 140
- 40. Sesso HD, Buring JE, Rifai N, Blake GJ, Gaziano JM, Ridker PM. C-reactive protein and the risk of developing hypertension. *JAMA*. (2003) 290:2945–51. doi: 10.1001/jama.290.22.2945
- 41. Cirillo P, Golino P, Calabrò P, Cali G, Ragni M, De Rosa S, et al. C-reactive protein induces tissue factor expression and promotes smooth muscle and endothelial cell proliferation. *Cardiovasc Res.* (2005) 68:47–55. doi: 10.1016/j. cardiores.2005.05.010
- 42. Shi P, Diez-Freire C, Jun JY, Qi Y, Katovich MJ, Li Q, et al. Brain microglial cytokines in neurogenic hypertension. *Hypertension*. (2010) 56:297–303. doi: 10.1161/HYPERTENSIONAHA.110.150409
- 43. Santisteban MM, Ahn SJ, Lane D, Faraco G, Garcia-Bonilla L, Racchumi G, et al. Endothelium-macrophage crosstalk mediates blood-brain barrier dysfunction in hypertension. *Hypertension*. (2020) 76:795–807. doi: 10.1161/HYPERTENSIONAHA. 120.15581
- 44. Bao Y, Qin L, Kim E, Bhosle S, Guo H, Febbraio M, et al. CD36 is involved in astrocyte activation and astroglial scar formation. *J Cereb Blood Flow Metab.* (2012) 32:1567–77. doi: 10.1038/jcbfm.2012.52
- 45. Ueno M, Nakagawa T, Nagai Y, Nishi N, Kusaka T, Kanenishi K, et al. The expression of CD36 in vessels with blood-brain barrier impairment in a stroke-prone hypertensive model. *Neuropathol Appl Neurobiol.* (2011) 37:727–37. doi: 10.1111/j.1365-2990.2011.01172.x
- $46.\ Akyüz\ A,$ Işık F. Systemic immune-inflammation index: a novel predictor for non-dipper hypertension. Cureus. (2022) $14{:}e28176.$ doi: 10.7759/cureus.28176
- 47. Jayedi A, Rahimi K, Bautista LE, Nazarzadeh M, Zargar MS, Shab-Bidar S. Inflammation markers and risk of developing hypertension: a meta-analysis of cohort studies. *Heart*. (2019) 105:686–92. doi: 10.1136/heartjnl-2018-314216

- 48. Wang S, Zhang L, Qi H, Zhang FL, Fang Q, Qiu L. Pan-immune-inflammatory value predicts the 3 months outcome in acute ischemic stroke patients after intravenous thrombolysis. *Curr Neurovasc Res.* (2023) 20:464–71. doi: 10.217 4/0115672026276427231024045957
- 49. Huang L. Increased systemic immune-inflammation index predicts disease severity and functional outcome in acute ischemic stroke patients. *Neurologist*. (2023) 28:32–8. doi: 10.1097/NRL.000000000000464
- 50. Zhou Y, Zhang Y, Cui M, Zhang Y, Shang X. Prognostic value of the systemic inflammation response index in patients with acute ischemic stroke. *Brain Behav.* (2022) 12:e2619. doi: 10.1002/brb3.2619
- 51. Han W, Yi HJ, Shin DS, Kim BT. Pan-immune-inflammation value predict delayed cerebral ischemia in patients with aneurysmal subarachnoid hemorrhage. *J Clin Neurosci.* (2024) 121:47–52. doi: 10.1016/j.jocn.2024.02.003
- 52. Ministrini S, Carbone F, Montecucco F. Updating concepts on atherosclerotic inflammation: from pathophysiology to treatment. *Eur J Clin Investig.* (2021) 51:e13467. doi: 10.1111/eci.13467
- $53.\,DeLong\,JH,\,Ohashi\,SN,\,O'Connor\,KC,\,Sansing\,LH.\,Inflammatory responses after ischemic stroke. Semin Immunopathol. (2022) 44:625–48. doi: 10.1007/s00281-022-00943-7$
- 54. Anadani M, Turan TN, Yaghi S, Spiotta AM, Gory B, Sharma R, et al. Change in smoking behavior and outcome after ischemic stroke: post-hoc analysis of the SPS3 trial. Stroke. (2023) 54:921–7. doi: 10.1161/STROKEAHA.121.038202
- 55. Ambrose JA, Barua RS. The pathophysiology of cigarette smoking and cardiovascular disease: an update. *J Am Coll Cardiol.* (2004) 43:1731–7. doi: 10.1016/j. jacc.2003.12.047
- 56. Saint-André V, Charbit B, Biton A, Rouilly V, Possémé C, Bertrand A, et al. Smoking changes adaptive immunity with persistent effects. *Nature*. (2024) 626:827–35. doi: 10.1038/s41586-023-06968-8
- 57. Appelros P, Stegmayr B, Terént A. Sex differences in stroke epidemiology: a systematic review. *Stroke*. (2009) 40:1082–90. doi: 10.1161/STROKEAHA.108.540781
- 58. Armstrong-Wells J, Johnston SC, Wu YW, Sidney S, Fullerton HJ. Prevalence and predictors of perinatal hemorrhagic stroke: results from the kaiser pediatric stroke study. *Pediatrics.* (2009) 123:823–8. doi: 10.1542/peds.2008-0874
- 59. Golomb MR, Fullerton HJ, Nowak-Gottl U, Deveber G. Male predominance in childhood ischemic stroke: findings from the international pediatric stroke study. *Stroke*. (2009) 40:52–7. doi: 10.1161/STROKEAHA.108.521203
- 60. Koellhoffer EC, McCullough LD. The effects of estrogen in ischemic stroke. Transl Stroke Res. (2013) 4:390–401. doi: $10.1007/\mathrm{s}12975-012-0230-5$
- 61. Banerjee A, McCullough LD. Sex-specific immune responses in stroke. *Stroke*. (2022) 53:1449–59. doi: 10.1161/STROKEAHA.122.036945
- 62. Suzuki S, Gerhold LM, Böttner M, Rau SW, dela Cruz C, Yang E, et al. Estradiol enhances neurogenesis following ischemic stroke through estrogen receptors alpha and beta. *J Comp Neurol.* (2007) 500:1064–75. doi: 10.1002/cne.21240
- 63. Arevalo MA, Azcoitia I, Garcia-Segura LM. The neuroprotective actions of oestradiol and oestrogen receptors. *Nat Rev Neurosci.* (2015) 16:17–29. doi: 10.1038/nrn3856
- 64. Ritzel RM, Capozzi LA, McCullough LD. Sex, stroke, and inflammation: the potential for estrogen-mediated immunoprotection in stroke. *Horm Behav.* (2013) 63:238–53. doi: 10.1016/j.yhbeh.2012.04.007
- 65. Niu P, Li L, Zhang Y, Su Z, Wang B, Liu H, et al. Immune regulation based on sex differences in ischemic stroke pathology. *Front Immunol.* (2023) 14:1087815. doi: 10.3389/fimmu.2023.1087815
- 66. Forthofer RN. Investigation of nonresponse bias in NHANES II. Am J Epidemiol. (1983) 117:507–15. doi: 10.1093/oxfordjournals.aje.a113568
- 67. Gu X, Balasubramanian R. Study design for non-recurring, time-to-event outcomes in the presence of error-prone diagnostic tests or self-reports. $Stat\ Med.$ (2016) 35:3961–75. doi: 10.1002/sim.6962
- 68. Paulose-Ram R, Graber JE, Woodwell D, Ahluwalia N. The National Health and nutrition examination survey (NHANES), 2021-2022: adapting data collection in a COVID-19 environment. *Am J Public Health*. (2021) 111:2149–56. doi: 10.2105/AJPH.2021.306517
- 69. McQuillan GM, McLean JE, Chiappa M, Corporation H, Lukacs SL. National Health and nutrition examination survey biospecimen program: NHANES III (1988-1994) and NHANES 1999-2014. *Vital Health Stat.* (2015) 2:1–14.