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Molecular transmission network analysis reveals the challenge of HIV-1 in ageing patients in China: elderly people play a crucial role in the transmission of subtypes and high pretreatment drug resistance in developed Eastern China, 2019–2023

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Abstract

Background The number and proportion of HIV/AIDS patients among older people are continuously and rapidly increasing in China. We conducted a detailed molecular epidemiological analysis of HIV-1 epidemic strains in a developed city in eastern China and found that elderly people play a crucial role in the transmission of subtypes and high pretreatment drug resistance (PDR).

Methods A total of 1048 samples were obtained from 1129 (92.8%) newly confirmed HIV-1-positive and treatment-naive patients between 2019 and 2023. The 1316 bp target fragment of the pol gene was amplified by reverse transcription polymerase chain reaction (RT–PCR) and nested PCR, and Maximum-likelihood (ML) phylogenetic trees and molecular transmission network were constructed to analyse the subtypes and transmission clusters. Molecular transmission network was visualized using Cytoscape with the distance threshold of 0.0075. PDR-associated mutations were determined according to the Stanford University HIV Drug Resistance Database.

Results A total of 933 pol sequences (89.0%, 933/1048) were successfully obtained, and twelve HIV-1 subtypes were detected. CRF07_BC was the predominant subtype, accounting for 48.1% (449/933) of sequences, followed by CRF01_AE (29.4%, 274/933). A total of 398 individuals (42.7%, 398/933) formed 89 clusters in the network. Multi-variable logistic regression analysis revealed that age, nationality, subtype, and PDR were the most significant factors associated with clustering in the transmission network. The prevalence of PDR was 14.6% (136/933). PDR associated

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with non-nucleoside reverse transcriptase inhibitors (10.0%, 93/933) was much more common than that associated with nucleoside reverse transcriptase inhibitors (1.8%, 17/933) and protease inhibitors (3.2%, 30/933) ($\chi^2 = 77.961$, $p < 0.001$). The most frequent NNRTI mutations were K103N/S/KN/NS (52.2%, 71/136), which led to the highest proportion of high-level resistance to nevirapine and efavirenz (52.2%).

Conclusions Our study revealed the important influence of elderly people on CRF07_BC transmission and the high prevalence of PDR. The clustering of drug-resistant cases was significant, which suggested the potential for localized widespread transmission of drug-resistant strains. HIV screening and the determination of PDR are recommended for older patients to improve early detection and reduce treatment failure and second-generation transmission.

Keywords HIV-1, Subtype, Transmission network, Molecular epidemiology, PDR, Pretreatment drug resistance

Background

HIV/AIDS remains one of the most important threats to public health. According to the Joint United Nations Programme on HIV/AIDS (UNAIDS) report in “THE PATH THAT ENDS AIDS”, the number of people living with HIV/AIDS (PLWH) globally by the end of 2022 was 39 million, with 1.3 million new infections and approximately 630,000 people dying from AIDS-related illnesses. According to the infectious disease surveillance data of the China Disease Control and Prevention Information System, by the end of 2022, there were 1.22 million living PLWH in China, among which 107,000 PLWH were newly diagnosed in 2022 [1]. China is facing the challenge of an ageing population. By 2022, individuals aged 50 years and older accounted for 36.8% of the total population [2]. At the same time, the number of PLWH aged 50 years and older continues to increase [3]. The AIDS epidemic in China is also facing challenges related to ageing: the proportion of people with newly reported HIV infections aged older than 50 years increased from 22% in 2011 to 44% in 2020 and reached 48.1% in 2022 [1]. However, there are few studies on HIV transmission among older adults. Men who have sex with men (MSM) are major drivers of HIV transmission in the developed eastern regions of China [4–6]. In contrast, the role of elderly people in HIV and drug resistance transmission might be significantly underestimated.

A molecular transmission network can be used to reveal HIV transmission relationships and patterns among infected individuals via phylogenetic analysis of viral sequences [7] based on the high genetic diversity of HIV-1. In recent years, the application of this network has shown initial results in detecting virus transmission and drug resistance as well as developing precise interventions for targeted populations. Multiple studies have revealed the diversity of HIV-1 prevalence in China, with regional distributions, transmission modes and different prevalences among high-risk groups [8, 9].

This study was designed to reveal the local transmission patterns of HIV through molecular epidemiological

analysis, especially focused on the impact of older people on HIV and PDR transmission. This high proportion of sampling could reflect the characteristics of HIV-1 transmission in Shaoxing more comprehensively and reduce deviation and insufficiency. Our study aimed to provide evidence to inform the development of targeted intervention strategies.

Methods

Study population and data collection

From January 1, 2019, to September 30, 2023, 1048 samples were obtained from 1129 (92.8%) newly confirmed HIV-1-positive and treatment-naïve patients by the Shaoxing Centers for Disease Control and Prevention (Shaoxing CDC) during the first follow-up visit. The plasma was separated on the day after sampling and stored at -80°C . The inclusion criteria were as follows: (1) were aged 16 years and older; (2) were newly diagnosed with HIV-1; and (3) had not received antiretroviral therapy (ART) before enrolment. Sociodemographic information was obtained from the National Center for AIDS/STD Control and Prevention (NCAIDS) information systems.

HIV nucleic acid extraction, amplification and sequencing

HIV RNA was extracted using an automated nucleic acid extraction machine (NP968-C system) and the Tianlong RNA Extraction Kit (Tianlong, Xian, China) according to the manufacturer's standard protocol. If the initial extraction was unsuccessful, another RNA extraction was performed using the High Pure Viral RNA Kit (Roche, Germany). The obtained RNA was amplified using a PrimeScript One Step RT-PCR Kit (Takara, Dalian, China) for reverse transcription polymerase chain reaction (RT-PCR) and nested PCR for the *pol* gene (HXB2: 2253–3306) as previously described [10]. After electrophoretic analysis, the amplified positive products were sent to BioGerm Biotechnology Co., Ltd., for purification and gene sequencing.

HIV subtype analysis

The sequences were assembled and adjusted with Sequencher v5.4.6 software (Genecodes, Ann Arbor, MI). Sequences with $\geq 5\%$ ambiguous bases were excluded from the study. If several sequences from the same patient were available in the database, only the oldest was retained. The assembled sequences were aligned, edited, and analysed with BioEdit 7.2 software using references downloaded from the Los Alamos National Laboratory (LANL) HIV sequence database (<https://www.hiv.lanl.gov>). Maximum-likelihood (ML) phylogenetic trees for different HIV-1 subtypes were constructed with IQ Tree 2.0.6 under the GTR+G+I nucleotide substitution model. Sequences with bootstrap values $\geq 70\%$ were determined to be the same subtype as the reference sequence. Sequences with possible intersubtype recombination were analysed using Simplot v 3.5.1. Then, the web-based tool iTOL v6.0 (<https://itol.embl.de/>) was used to visually edit the sequences for further analysis of geographic locations, phylogenetic clusters, and risk groups.

Genetic transmission network analysis

HIV Trace was used to calculate the genetic distances (GDs) based on the Tamura-Nei 93 (TN93) model. We performed analyses at different GD thresholds (ranging from 0.0025 to 0.015 at intervals of 0.0025) and selected the GD with the maximum number of transmission clusters. Cytoscape v3.7.2 software was used to visualize the HIV molecular transmission network. Each node in the network represents an individual *pol* sequence or case, and each edge represents the propagation correlation between the nodes. The greater the number of edges is, the greater the inferred communication relationship with more people, and the greater the transmission risk. In this study, according to the aggregation degree of nodes, sequences were divided into large clusters (LCs, ≥ 10 nodes), medium clusters (MCs, 4–9 nodes) and small clusters (SCs, 4–9 nodes). In addition, individuals with ≥ 4 links were identified as high-risk transmission individuals [11].

Drug resistance analysis

The *pol* sequences were submitted to the Stanford University HIV Resistance Database HIVdb Program (<https://hivdb.stanford.edu/hivdb/bysequences/>). Drug resistance mutations (DRMs) and their associated resistance levels were determined using the Stanford University HIV Drug Resistance Database as follows: susceptible, potential low-level resistance, low-level resistance, intermediate resistance, and high-level resistance, according to the Stanford penalty.

Statistical analysis

Descriptive statistics were performed on sociodemographic parameters to compare patients in transmission clusters, patients in LCs, and patients with high linkage. Categorical variables were compared using Pearson's chi-square test or Fisher's exact test. Univariate and multivariate logistic regression analyses were performed to identify factors associated with clustering and high-risk individuals. Significant variables identified from the univariate analysis ($P < 0.10$) were included in a multiple logistic regression model to determine the factors that influence HIV clustering. P values less than 0.05 were considered to indicate statistical significance. Data were analysed using IBM SPSS Statistics 26.0 (IBM Corp., Armonk, NY, USA). P values were two-sided with a significance level of 0.05.

Results

Patient demographic characteristics

A total of 933 (89.0%, 933/1048) qualified sequences were obtained, representing 82.6% (933/1129) newly diagnosed HIV-positive patients in Shaoxing city from 2019 to 2023. As shown in Table 1, the patients were mainly male (82.1%, 766/933) and local residents (84.1%, 785/933); 52.9% (494/933) of patients were married, 33.8% (315/933) had a junior middle school education, 61.8% (577/933) were infected through heterosexual contact, and 48.1% (449/933) were infected with the CRF07_BC subtype. A total of 38.3% (357/933) of the individuals were older than 50 years, and the median age for the entire cohort was 44 years (ranging from 18 to 81 years). There was a significant difference in the composition of subtypes between the ≥ 50 -year-old group and the < 50 -year-old group ($\chi^2 = 36.929$, $P < 0.001$).

HIV subtype distribution

Twelve HIV-1 subtypes and a proportion of unique recombinant forms (URFs) were identified from the 933 *pol* sequences. CRF07_BC was the predominant subtype, accounting for 48.1% (449/933) of sequences, followed by CRF01_AE (29.4%, 274/933). The other subtypes detected were CRF08_BC (11.3%, 105/933), CRF85_BC (3.3%, 31/933), CRF55_01B (2.5%, 23/933), B (1.4%, 13/933), C (0.8%, 7/933), CRF87_cpx (0.3%, 3/933), CRF57_BC (0.2%, 2/933), 02_AG (0.1%, 1/933), URF (CRF01_AE/CRF07_BC) (2.47%, 23/933), URF (CRF01_AE/B) (0.1%, 1/933), and URF (B/C) (0.1%, 1/933).

By constructing the ML phylogenetic trees of the different HIV-1 subtypes (Fig. 1), we found that there was a large evolutionary clade in the CRF07_BC subtype named C1. A total of 94.2% (65/69) of the CRF07_BC-C1 individuals were infected through heterosexual contact,

Table 1 Sociodemographic Characteristics of newly diagnosed HIV-1 patients in Shaoxing from 2019 to 2023

Categories	Cases(%) (n = 933)	HIV subtypes(%)			χ^2	P
		CRF07_BC (n = 449)	CRF01_AE (n = 274)	Others (n = 210) ^a		
Collection time					16.236	0.039
2019	222(23.8)	128(28.5)	61(22.3)	33(15.7)		
2020	181(19.4)	77(17.1)	53(19.3)	51(24.3)		
2021	210(22.5)	100(22.3)	64(23.4)	46(21.9)		
2022	189(20.3)	88(19.6)	56(20.4)	45(21.4)		
2023	131(14.0)	56(12.5)	40(14.6)	35(16.7)		
Sex					5.427	0.066
Male	766(82.1)	366(81.5)	236(86.1)	164(78.1)		
Female	167(17.9)	83(18.5)	38(13.9)	46(21.9)		
Age(years)					52.198	<0.001
≤25	124(13.3)	75(16.7)	33(12.0)	16(7.6)		
26–39	266(28.5)	127(28.3)	103(37.6)	36(17.2)		
40–49	186(19.9)	89(19.8)	56(20.4)	41(19.5)		
50–59	223(23.9)	103(22.9)	53(19.4)	67(31.9)		
≥60	134(14.4)	55(12.3)	29(10.6)	50(23.8)		
Ethnicity					0.202	0.904
Ethnic Han	882(94.5)	426(94.9)	258(94.2)	198(94.3)		
Others	51(5.5)	23(5.1)	16(5.8)	12(5.7)		
Married status					19.342	0.004 ^b
Single	288(30.9)	149(33.2)	97(35.4)	42(20.0)		
Married	494(52.9)	235(52.3)	132(48.2)	127(60.5)		
Divorced/Widowed	145(15.5)	62(13.8)	42(15.3)	41(19.5)		
Unknow	6(0.6)	3(0.7)	3(1.1)	0		
Education					36.202	<0.001
Junior college or above	163(17.5)	68(15.1)	69(25.2)	26(12.4)		
High or technical secondary school	198(21.2)	113(25.2)	58(21.2)	27(12.9)		
Junior middle school	315(33.8)	157(35.0)	79(28.8)	79(37.6)		
Primary school/illiterate	257(27.5)	111(24.7)	68(24.8)	78(37.1)		
Household registered					5.374	0.251
Other provinces	96(10.3)	41(9.1)	25(9.1)	30(14.3)		
Other cities in Zhejiang	52(5.6)	23(5.1)	18(6.6)	11(5.2)		
Shaoxing city	785(84.1)	385(85.8)	231(84.3)	169(80.5)		
Sexual contact					53.657	<0.001 ^b
Heterosexual	577(61.8)	250(55.7)	157(57.3)	170(80.9)		
Homosexual	338(36.2)	188(41.9)	115(42.0)	35(16.7)		
IDUs	4(0.4)	4(0.9)	0	0		
Unknow	14(1.5)	7(1.5)	2(0.7)	5(2.4)		

^a: Others were CRF08_BC 105, CRF85_BC 31, CRF55_01B 23, B 13, C 7, CRF87_cpx 3, CRF57_BC 2, 02_AG 1, URF(B/C) 1, URF (CRF01_AE/_B) 1 and URF (CRF01_AE/CRF07_BC) 23; ^b: Fisher's exact test

88.4% (61/69) were older than 50 years, and 88.4% (61/69) were drug resistant. CRF01_AE in Shaoxing city was homologous to several CRF01_AE epidemic clusters in China. The major subclusters were CRF01_AE Cluster1 (23.7%, 65/274), CRF01_AE Cluster4 (19.3%, 53/274) and CRF01_AE Cluster5 (8.6%, 24/274). The other detected subclusters were CRF01_AE Cluster8 (1.5%, 4/274), CRF01_AE Cluster2 (1.5%, 4/274), and CRF01_AE

Cluster3 (1.8%, 5/274). Individuals in the largest subcluster, CRF01_AE Cluster1, were mainly infected through heterosexual transmission (95.4%, 62/65), and 63.1% (41/65) of the patients were over 50 years old. In addition to the two major HIV-1 subtypes, the other HIV-1 subtypes were concentrated among heterosexual individuals (81.0%, 170/210).

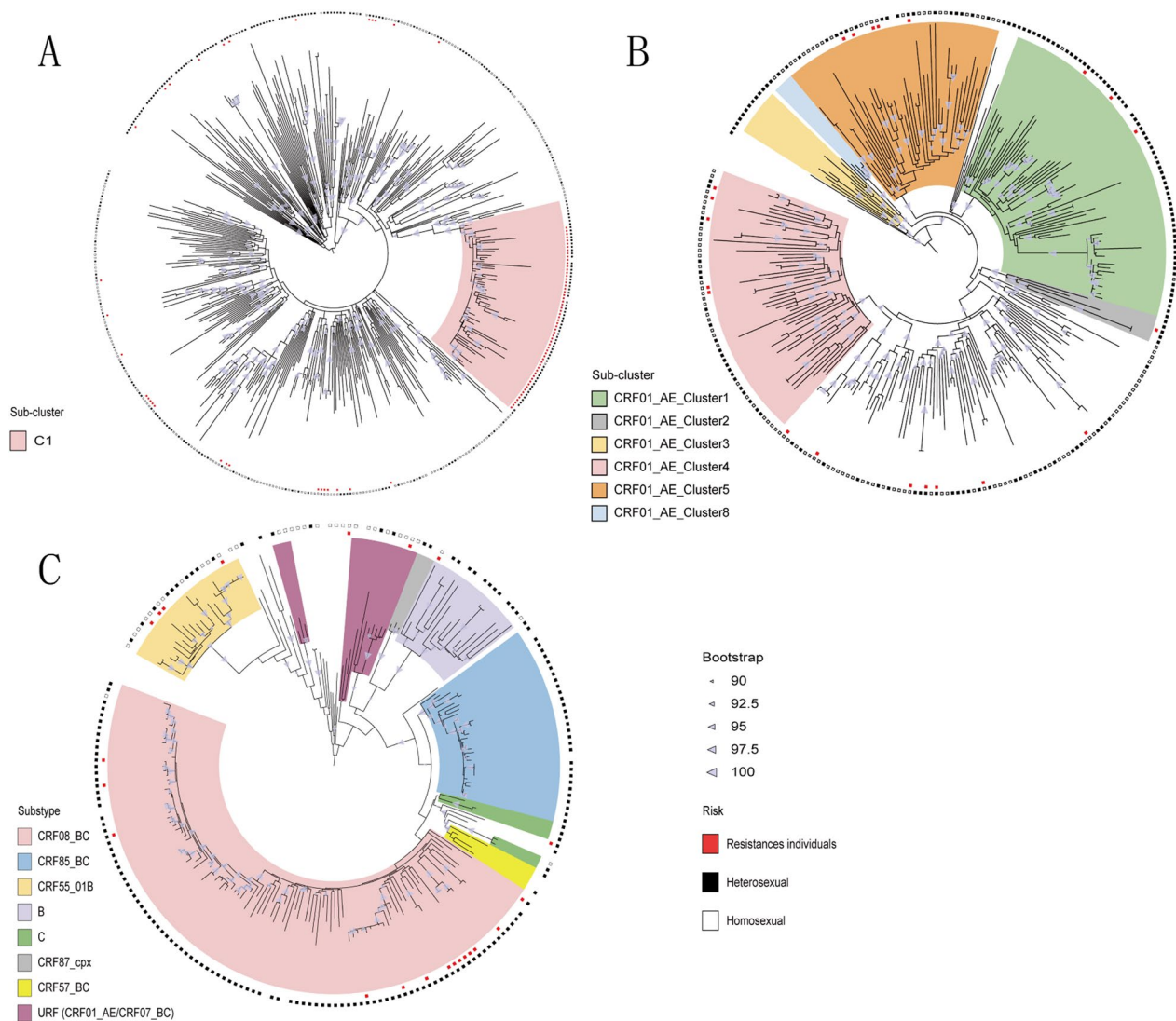


Fig. 1 Maximum-likelihood phylogenetic trees of patients infected with different HIV-1 subtypes in Shaoxing city from 2019 to 2023. The phylogenetic trees based on the pol region of different HIV-1 subtypes were constructed by an approximate maximum likelihood method with IQ Tree 2.0.6 under the GTR+G+I nucleotide substitution model. Clades of different colours represent different HIV-1 subtypes and subclusters. The inner circle red colour square box represents resistances individuals; the outer circles shows the risk group, Black colour square box represents heterosexual individuals, White colour square box represents homosexual individuals. **A** Maximum-likelihood phylogenetic tree of CRF07_BC. **B** Maximum-likelihood phylogenetic tree of CRF01_AE. **C** Maximum-likelihood phylogenetic tree of other HIV-1 subtypes.

Genetic transmission network analysis

At the optimal GD of 0.0075 substitutions/site, 398 individuals (42.7%, 398/933) formed 98 clusters (mean 4.1, median 2.0, range 2–64, interquartile range 2–3.75) and 1534 edges (mean 7.7, median 3.0, range 1–59, interquartile range 1–5) (Table 2 and Fig. 2). Among the 98 clusters, the subtypes included CRF07_BC (51.3%, 204/398), CRF01_AE (22.4%, 89/398), CRF08_BC (15.3%, 61/398), CRF85_BC (6.0%, 24/398), CRF55_01B (2.5%, 10/398), B (1.0%, 4/398), C (1.0%, 4/398) and URF (CRF01_AE/CRF07_BC) (0.5%, 2/398). There were 4 LCs (4.1%,

4/98), 21 MCs (21.4%, 21/98), and 73 SCs (74.5%, 73/98). Eight MCs clustered patients who were diagnosed after 2021, and 18 clusters exhibited an increasing number of cases ≥ 3 from 2021 to 2023. CRF07_BC formed the most clusters (nodes ranging from 2 to 64) in the network. We also observed that 21.4% (85/398) of the PDR patients were included in the transmission network. A total of 65.9% (56/85) of the PDR patients were concentrated in the LC1 cluster (Table 2 and Fig. 2).

The proportions of PDR in patients with different subtypes were 35.8% (73/204), 4.5% (4/89), and 7.6% (8/105)

Table 2 Molecular network clustering of different HIV-1 subtypes in Shaoxing city from 2019 to 2023

Subtypes	Cases(n)	Clusters(n)	SCs(2–3 nodes)	MCs(4–9 nodes)	LCs(≥ 10 nodes)	Maximum cluster cases(n)
Total	398	98	73	21	4	–
CRF07_BC	204	43	30	12	1	64
CRF01_AE	89	31	26	4	1	16
CRF08_BC	61	12	8	3	1	28
CRF85_BC	24	4	2	1	1	16
CRF55_01B	10	3	2	1	0	6
B	4	2	2	0	0	2
C	4	2	2	0	0	2
URF(CRF01_AE/ CRF07_BC)	2	1	1	0	0	2

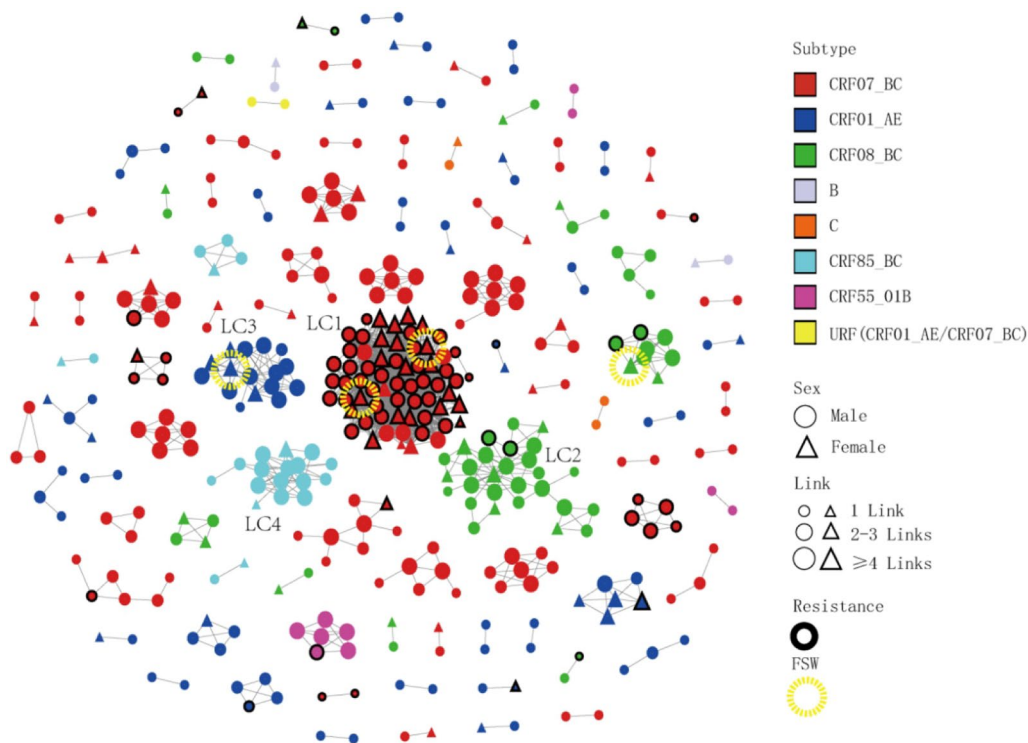



Fig. 2 Molecular transmission network analysis of patients infected with different HIV-1 subtypes in Shaoxing city from 2019 to 2023. All edges represent a genetic distance between nodes of less than 0.0075 substitution/site, and the colour of the node indicates the different HIV-1 subtypes. Different shapes denote sex, and the node size indicates the number of associated links. Thick border represents resistances individuals.  Represents FSW individuals

for CRF07_BC, CRF01_AE, and other subtypes, respectively. The proportion in the CRF07_BC subtype was significantly greater than those of the CRF01_AE and other subtypes ($\chi^2 = 52.148$, $p < 0.001$) (Fig. 2).

Of the four LCs, the CRF07_BC cluster (LC1) had 64 nodes, the CRF08_BC cluster (LC2) had 28 nodes, the CRF01_AE cluster (LC3) had 16 nodes and the

CRF85_BC cluster (LC4) had 16 nodes (Fig. 2). All four LCs were defined as elder heterosexual infection clusters. A total of 124 nodes were identified; 93.5% (116/124) were predominantly infected through heterosexual contact, and 80.6% (100/124) were aged 50 years or older. There were 34 females, 52.9% (18/34) of whom were infected by spouses or regular sexual partners. In

Table 3 Summary statistics of the four largest clusters in Shaoxing city from 2019 to 2023

LC cluster	Cases (n)	subtype	Links Mean [Median, IQR]	Genetic distances Mean [Median, IQR]	Growing case(%)				Age ≥ 50y (%)	Male (%)	Shaoxing household registration(%)	Heterosexual(%)	PDR(%)		
					2019	2020	2021	2022					2023	PI	NNRTI
LC1	64	CRF07_ BC	33 [39,18–47]	0.0051[0.0052, 0.0042–0.0062]	24 (37.5)	10 (15.6)	15 (23.4)	13 (20.3)	2 (3.1)	44 (68.8)	42 (65.6)	58 (90.6)	60 (93.8)	Q58QE 0 1(1.6)	K103N/S/ NS Y188F 56(87.5)
LC2	28	CRF08_ BC	4.2 [3.5,2.8–5.0]	0.0049[0.0054, 0.0035–0.0069]	3 (10.7)	9 (32.1)	11 (39.3)	4 (14.3)	1 (3.6)	19 (67.9)	22 (78.6)	25 (89.3)	26 (92.9)	0	V75A K65KN 2(7.1)
LC3	16	CRF01_ AE	7.0 [6.0,5.0–9.0]	0.0049[0.0052, 0.0035–0.0060]	4 (25.0)	2 (12.5)	5 (31.3)	2 (12.5)	3 (18.8)	14 (87.5)	12 (75.0)	13 (81.3)	15 (93.8)	0	0
LC4	16	CRF85_ BC	6.0 [5.5,3.0–8.3]	0.0050[0.0054, 0.0042–0.0067]	2 (12.5)	5 (31.3)	4 (25.0)	2 (12.5)	3 (18.8)	13 (81.3)	14 (87.5)	14 (87.5)	14 (87.5)	0	0

addition, 4 female sex workers (FSWs) from other provinces were included in these clusters, including two patients in LC1, one patient in LC2 and one patient in LC3. The largest cluster (LC1) contained 64 nodes, a mean of 33 edges (median 39, range 1–59, interquartile range 18–47), and a mean GD of 0.0051 (median 0.0052, interquartile range 0.0042–0.0062). Among the 64 patients, 90.6% (58/64) were local residents, 85.9% (55/64) had a junior middle school education or below, 93.8% (60/64) were infected through heterosexual transmission, 87.5% (56/64) were drug resistant, and 32.8% (21/64) were female; most females (66.7%, 14/21) were detected after their spouses or regular sexual partners were confirmed to be HIV positive. The characteristics of the four LCs are shown in Table 3.

Factors associated with clustering and high linkages

Of the individuals included in the clusters, 78.6%(313/398) were male, 50.0%(199/398) were ≥ 50 years old, 59.5%(237/398) were married, 51.3%(204/398) were infected with CRF07_BC, 88.2%(351/398) were local residents, and 67.8%(270/398) were infected through heterosexual transmission (Table 4). Multivariate logistic regression analysis revealed that age, nationality, subtype, and PDR were the most significant factors associated with clustering in the transmission network (Table 4). Patients aged ≥ 50 years were more likely to cluster within the network (adjusted odds ratio [AOR]=1.868, 95% confidence interval [CI] 1.329–2.625). We also found that individuals with PDR were more likely to cluster than those without PDR (AOR=2.219, 95% CI 1.491–3.301). Individuals in other ethnic groups (AOR=0.464, 95% CI 0.232–0.925) and with CRF01_AE (AOR=0.678, 95% CI 0.487–0.944) had significantly lower odds of clustering than Han individuals and patients with CRF07_BC.

We defined nodes with ≥ 4 links as high-risk individuals. Multivariate logistic regression analysis also revealed that age, subtype, and PDR were significant factors associated with high-risk individuals (Table 4). The results showed that the effect of age ≥ 50 years was 5.149 times greater than that of age < 50 years (95% CI=2.689–9.859), that of individuals with PDR was 3.899 times greater than that of individuals without PDR (95% CI=1.971–7.711), and that of CRF01_AE and other subtypes were 0.303 times (95% CI=0.148–0.622) and 0.410 times (95% CI=0.214–0.786) greater than that of CRF07_BC.

Drug resistance mutation analysis

The prevalence of PDR among newly diagnosed patients was 14.6% (136/933). Eleven protease inhibitor (PI) resistance-associated mutation patterns, 10 nucleoside reverse transcriptase inhibitor (NRTI) mutation patterns, and 13 non-nucleoside reverse transcriptase inhibitor

(NNRTI) mutation patterns were identified. The overall prevalence of HIV-1 PDR to NNRTIs (10.0%, 93/933) was much greater than that to NRTIs (1.8%, 17/933) and PIs (3.2%, 30/933) ($\chi^2=77.961, p<0.001$). In individuals with NNRTI resistance, the most frequent mutations were K103N/S/KN/NS (52.2%, 71/136), which were responsible for the greatest proportion of high-level resistance to nevirapine (NVP) and efavirenz (EFV) (52.2%). In individuals with PI resistance, resistance to tipranavir (TPV/r) was the most common (16.2%, 22/136), which was mainly related to the Q58E/QE mutation (14.7, 20/136)(Fig. 3). Four (2.9%, 4/136) strains harboured two classes of drug resistance mutations. No HIV-1 strains with PDR mutations to three classes of drugs were found in this study.

Discussion

A genetic transmission network based on the HIV-1 *pol* gene is widely used in HIV-targeted precision intervention [4, 12–14]. A high coverage rate of sample collection ($\geq 60\%$) is conducive to avoiding bias caused by insufficient sampling [15]. Via consecutive cross-sectional sampling of a high proportion of newly diagnosed HIV-1 patients between 2019 and 2023 and a detailed molecular epidemiological approach, we constructed a molecular network for Shaoxing, a developed city in eastern China. We detected a high rate of infection, clustering in the transmission network, and a high number of linkages in elderly people (≥ 50 years) in Shaoxing. In addition, four large molecular clusters were identified in people over 50 years old, and the largest molecular cluster had a high proportion of patients with PDR (87.5%, 56/64), which has never been observed before, suggesting that people over 50 years of age are the main drivers of HIV transmission in Shaoxing. The spread of drug-resistant molecular clusters resulted in a high rate of PDR in Shaoxing, a typical epitome of the HIV epidemic in ageing people in China. Our study provides a basis for the assessment of epidemic development and the formulation of targeted prevention and intervention strategies.

The CRF07_BC subtype formed the largest number of clusters in the network (Table 2) and was present in significantly greater proportions in different years, male patients, and elderly people over 50 years old compared with the other subtypes (Table 1). These findings indicated that CRF07_BC spreads rapidly through people over 50 years old, which may be related to the transmission characteristics of CRF07_BC [16]. Its pathogenicity is less pronounced than that of the CRF01_AE and B subtypes [17], but it has greater transmission efficiency and slower disease progression in group of men who have sex with men. Therefore, CRF07_BC represents a new form

Table 4 Factor associated with clustering and high linkages of newly diagnosed HIV-1 patients in Shaoxing from 2019 to 2023

Categories	Cluster cases(%)(n = 398)	Univariate analysis		Multivariate analysis		High-risk cases(%)(n = 157)	Univariate analysis		Multivariate analysis		
		OR (95% CI)	P	AOR (95% CI)	P		OR (95% CI)	P	AOR (95% CI)	P	
<i>Age(years)</i>											
<50	199(50.0)	1.000		1.000		44(28.0)	1.000		1.000		
≥50	199(50.0)	2.386(1.821–3.127)	0.000	1.868(1.329–2.625)	0.000	113(72.0)	4.629(2.991–7.164)	0.000	5.149(2.689–9.859)	0.000	
<i>Sex</i>											
Male	313(78.6)	1.000		1.000		123(78.3)	1.000				
Female	85(21.4)	1.500(1.072–2.099)	0.018	1.282(0.867–1.896)	0.213	34(21.7)	1.030(0.631–1.680)	0.906			
<i>Ethnicity</i>											
Ethnic Han	398(96.7)	1.000		1.000		154(98.1)	1.000				
Others	13(3.3)	0.442(0.232–0.841)	0.013	0.464(0.232–0.925)	0.029	3(1.9)	0.450(0.122–1.661)	0.231			
<i>Married status</i>											
Single	100(25.1)	1.000		1.000		25(15.9)	1.000		1.000		
Married	237(59.5)	1.734(1.284–2.340)	0.000	1.083(0.750–1.563)	0.671	101(64.3)	2.228(1.324–3.750)	0.003	0.573(0.264–1.245)	0.160	
Divorced/Widowed	58(14.6)	1.253(0.831–1.891)	0.282	0.803(0.507–1.273)	0.351	30(19.1)	3.214(1.619–6.381)	0.001	1.533(0.646–3.639)	0.333	
Unknown	3(0.8)	1.880(0.373–9.486)	0.445	1.095(0.176–6.830)	0.922	1(0.6)	1.500(0.130–17.257)	0.745	0.076(0.004–1.288)	0.074	
<i>Education</i>											
Junior college or above	51(12.8)	1.000		1.000		11(7.0)	1.000		1.000		
High or technical secondary school	82(20.6)	0.152(1.004–2.399)	0.048	1.439(0.914–2.265)	0.117	24(15.3)	1.503(0.663–3.415)	0.328	1.407(0.547–3.617)	0.478	
Junior middle school	138(34.7)	1.712(1.149–2.552)	0.008	1.281(0.829–1.981)	0.265	57(36.3)	2.559(1.211–5.408)	0.014	1.510(0.611–3.733)	0.372	
Primary school/illiterate	127(31.9)	2.145(1.422–3.238)	0.000	1.331(0.819–2.164)	0.249	65(41.4)	3.812(1.796–8.091)	0.000	1.528(0.604–3.866)	0.371	
<i>Household registered</i>											
Other provinces	32(8.0)	1.000		1.000		15(9.6)	1.000				
Other cities in Zhejiang	15(3.8)	0.811(0.389–1.691)	0.576	0.702(0.325–1.517)	0.368	3(1.9)	0.283(0.067–1.200)	0.087			
Shaoxing city	351(88.2)	1.618(1.034–2.529)	0.035	1.481(0.923–2.376)	0.103	139(88.5)	0.746(0.359–1.537)	0.423			
<i>Subtype</i>											
CRF07_BC	204(51.3)	1.000		1.000		101(64.3)	1.000		1.000		
CRF01_AE	89(22.4)	0.578(0.422–0.791)	0.001	0.678(0.487–0.944)	0.021	18(11.5)	0.259(0.144–0.464)	0.000	0.303(0.148–0.622)	0.001	
Other	105(26.4)	1.201(0.865–1.667)	0.274	1.185(0.829–1.683)	0.353	38(24.2)	0.578(0.357–0.938)	0.026	0.410(0.214–0.786)	0.007	
<i>Sexual contact</i>											
Heterosexual	270(67.8)	1.000		1.000		125(79.6)	1.000		1.000		
Homosexual	121(30.4)	0.634(0.481–0.836)	0.001	1.053(0.744–1.489)	0.771	25(15.9)	0.302(0.183–0.498)	0.000	0.494(0.244–0.999)	0.050	
Unknown	7(1.8)	0.724(0.277–1.893)	0.510	0.632(0.222–1.799)	0.390	7(4.5)	/				
<i>PDR</i>											
No	313(78.6)	1.000		1.000		95(60.5)	1.000		1.000		
Yes	85(21.4)	2.577(1.771–3.750)	0.000	2.219(1.491–3.301)	0.000	62(39.5)	6.186(3.620–10.570)	0.000	3.899(1.971–7.711)	0.000	

Table 4 (continued)

Categories	Cluster cases(%) (n = 398)	Univariate analysis		Multivariate analysis		High-risk cases(%) (n = 157)	Univariate analysis		Multivariate analysis	
		OR (95% CI)	P	AOR (95% CI)	P		OR (95% CI)	P	AOR (95% CI)	P
Current place of residence										
Yuecheng	127(31.9)	1.000				67(42.7)	1.000			1.000
Keqiao	113(28.4)	1.521(0.767–3.018)	0.230			48(30.6)	0.661(0.397–1.102)	0.112		1.232(0.654–2.320)
Zhuji	84(21.1)	1.169(0.589–2.317)	0.656			25(15.9)	0.379(0.212–0.680)	0.001		0.822(0.401–1.686)
Shangyu	22(5.5)	0.833(0.417–1.664)	0.604			3(1.9)	0.141(0.040–0.502)	0.002		0.469(0.112–1.967)
Shengzhou	36(9.0)	0.518(0.232–1.157)	0.109			13(8.3)	0.506(0.236–1.087)	0.081		1.528(0.599–3.900)
Xingchang	16(4.0)	1.101(0.509–2.382)	0.807			1(0.6)	0.060(0.008–0.466)	0.007		0.121(0.014–1.073)

Others genotypes were CRF08_BC, CRF85_BC, CRF55_01B, B, C, CRF87_cpX, CRF57_BC, 02_AG, URF(B/C), URF(CRF01_AE/B), URF(CRF01_AE/B_C), PDR: pretreatment drug resistance; OR: odds ratio; AOR: adjusted odds ratio; red data represent $P \leq 0.05$

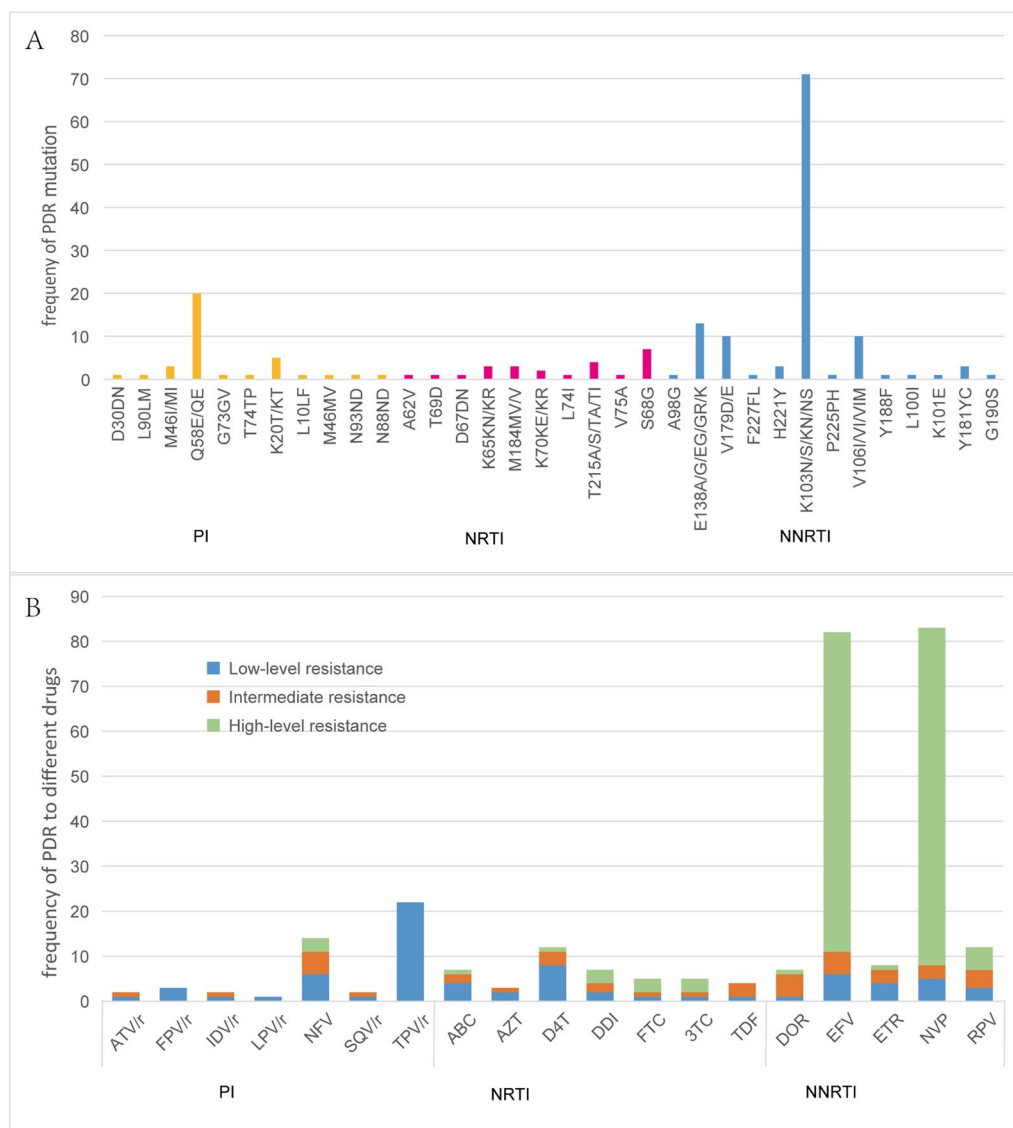


Fig. 3 **A:** Frequency of HIV-1 pretreatment drug resistance (PDR) mutations to PIs, NRTIs, and NNRTIs. **B:** Level of HIV-1 PDR associated mutations to different ART drug among 136 newly diagnosed HIV/AIDS patients. PI, Protease inhibitor; NRTI, Nucleoside reverse transcriptase inhibitor; NNRTI, Non-nucleoside reverse transcriptase inhibitor; ATV/r:Atazanavir/ritonavir; FPV/r:Fosamprenavir/ritonavir; IDV/r:Indinavir/ritonavir; LPV/r:Lopinavir/ritonavir; NFV:Nelfinavir; SQV/r:Saquinavir/ritonavir; TPV/r: Ritonavir/ritonavir; ABC: Abacavir; AZT: Zidovudine; D4T: Stavudine; DDI: Didanosine; FTC: Emtricitabine; 3TC, Lamivudine; TDF, Tenofovir; DOR, Doravirine; EFV, Efavirenz; ETR: Etravirine; NVP: Nevirapine; RPV: Rilpivirine

of rapid expansion in China [16, 18, 19] and has become the most commonly detected subtype in many areas.

We found that older people (aged ≥ 50 years) were more likely to form clusters in the transmission network (50%, 199/398), and high-risk linkages (72.0%, 113/157) and individuals who carried the resistance mutation, i.e., K103N (87.5%, 56/64), were part of the largest CRF07_BC cluster (LC1) (Tables 4 and 2). Most male (94.4%, 85/90) individuals in the four LCs self-reported nonmarital heterosexual sexual contact, which reflected elderly men, especially younger ones, still have

psychological and physical needs [20]. Rich economic conditions, easier communication, convenient transportation and more open minds allow these individuals to experience nonmarital heterosexual sex through commercial or noncommercial sources for thrill seeking. All of these factors maintain the growth of commercial sex [21]. Most of the patients in clusters (67.8%, 204/398) and in high-risk groups (79.6%, 125/157) were infected through heterosexual transmission. Moreover, poor knowledge of AIDS prevention has resulted in decreased condom use, increasing the likelihood of

HIV transmission between older male clients and FSWs [22] and of transmission from males to their spouses or regular sexual partners [23]. Our study conclusively confirmed this finding; we detected 4 FSWs from other provinces within the 4 LCs (Fig. 2), and 52.9% (18/34) of the female individuals in the four LCs were infected by their spouses or permanent partners (Table 3). All of these findings indicated that patients aged older than 50 years play an important role in HIV transmission in Shaoxing and should be monitored.

Since China had offered a free ART program for AIDS patients for more than 20 years, the rapid growth of PDR in recent years may have been underestimated due to a lack of resistance monitoring and testing before ART. Our study showed that PDR was significantly greater in Shaoxing (14.6%) than in Yunnan [24], Guangxi [25] and other severe HIV epidemic areas [26, 27]. According to the WHO definition of the prevalence of HIV drug resistance, our study revealed moderate (5–15%) rates of PDR in the Shaoxing region, which is extremely close to cut-off for a high rate of resistance (>15%). Some studies have suggested that PDR is primarily driven by resistance to NNRTIs [28], which is consistent with our study results. The overall prevalence of HIV-1 PDR to NNRTIs reached 10.0% (93/933), which significantly differed from that of PI and NRTI ($\chi^2=77.961$, $P<0.001$) in Shaoxing. Our study revealed a specific regional characteristic: 78.9% (56/71) of patients with K103N/S/KN/NS mutations in Shaoxing were older than 50 years, and 77.5% (55/71) were concentrated in the largest molecular cluster (LC1). The difference was significant ($\chi^2=10.000$, $P=0.001$). These findings suggested that K103N/S/KN/NS mutations spread rapidly through the large molecular transmission cluster LC1 in patients over 50 years old, resulting in a significantly high proportion of PDR in Shaoxing. In addition to LC1, two drug resistance genes in CRF07_BC contained 4 and 5 nodes. The aggregation of drug-resistant strains is a target for intervention priority. Effective monitoring and intervention of the LC1 cluster could significantly reduce PDR in Shaoxing. Due to the lack of pretreatment resistance testing, patients who are experiencing drug resistance before treatment may be more prone to antiviral failure. Specifically, elderly people need more safe and complex treatment because of poor health.

Our study provided some insights and information to explore the transmission characteristics and epidemic patterns of HIV in ageing individuals. This study also inevitably has several limitations. First, our results represent the Shaoxing epidemic alone. Due to the limited activities of elderly people, it is believed that they are infected mainly by heterosexual contact in local places. Our sample had little influence from MSM or injection drug use. Although the epidemic among elderly people

continues to grow across China, there are still significant differences between different regions. Therefore, future research should focus on a larger geographic range of older individuals with infections. Second, all of the patients were newly diagnosed, but we do not know exactly when they were infected. Elderly people have a higher rate of late diagnosis [29]. Thus, the network constructed may not reflect the true prevalence status. Finally, given limited time and funds, subtypes were identified based only on the *pol* gene sequence, which may not represent the exact subtype.

Conclusion

In summary, our study revealed the important influence of elderly people on the transmission of the CRF07_BC subtype and the high prevalence of PDR. Four clusters confirmed the transmission relationship between elderly men and FSWs and between elderly men and their spouses. The clustering of drug-resistant cases is significant, leading to the potential for localized widespread transmission of drug-resistant strains. Finally, a high prevalence of PDR could compromise the efficacy of ART and impact the overall management of the epidemic. Therefore, it is essential to conduct PDR testing for HIV patients and strengthen the dissemination of AIDS knowledge to older people.

Abbreviations

HIV	Human immunodeficiency virus
AIDS	Acquired immunodeficiency syndrome
PLWHA	People living with HIV/AIDS
ART	Antiretroviral therapy
WHO	World Health Organization
PDR	Pretreatment drug resistance
PCR	Polymerase chain reaction
PI	Protease inhibitor
NRTI	Nucleoside reverse transcriptase inhibitor
NNRTI	Non-nucleoside reverse transcriptase inhibitor
ATV/r	Atazanavir/ritonavir
FPV/r	Fosamprenavir /ritonavir
IDV/r	Indinavir/ritonavir
LPV/r	Lopinavir/ritonavir
NFV	Nelfinavir
SQV/r	Saquinavir/ritonavir
TPV/r	Ritonavir/ritonavir
ABC	Abacavir
AZT	Zidovudine
D4T	Stavudine
DDI	Didanosine
FTC	Emtricitabine
3TC	Lamivudine
TDF	Tenofovir
DOR	Doravirine
EFV	Efavirenz
ETR	Etravirine
NVP	Nevirapine
RPV	Rilpivirine

Acknowledgements

We are thankful to the participants, primary clinicians and nurses for sample collection.

Author contributions

DC is responsible for sample detection, study design, analysis and writing article. HX, YF, JZ, and TH are responsible for guiding the whole study and revising the article. JL, JC, and JZ are responsible for performed experiments, statistics and figures, translation. All authors read and approved the final version of the paper.

Funding

This work was supported by Projects of Science Technology Bureau of Shaoxing (Grant number: 2018C30051, 2022A14007), and the 2022 Zhejiang Province Higher Education Curriculum Ideological and Political Education Research Project (Project Number: ZJSJ2022406).

Data availability

The datasets used and/or analyzed during this study are available from the corresponding author upon reasonable request, and with permission from Shaoxing Center for Disease Control and Prevention.

Declarations

Ethics approval and consent to participate

The study protocol was reviewed and approved by the Institutional Review Board of the Shaoxing CDC in China (SXCDC2018-002 and SXCDC2021-001). Participants provided written informed consent to use sequencing data both for HIV genetic transmission network and drug resistance studies. The study was conducted in accordance with the «Helsinki Declaration».

Consent for publication

Not applicable.

Competing interests

No potential conflict of interest was reported by the authors.

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Received: 2 June 2024 Accepted: 31 July 2024

Published online: 26 August 2024

References

- Han MJ. Analysis of the epidemic situation and prospects of HIV/AIDS prevention and control in China. *Chin J AIDS STD*. 2023;29(3):247–50.
- China NBoSo. China statistical yearbook; 2023.
- Yue Q, Liu YF, Li H, Zhao Y. Characteristics and the first CD4(+)T lymphocytes test of newly-reported HIV/AIDS cases aged 50 years and above in the third round of China comprehensive AIDS response program. *Zhonghua Liu Xing Bing Xue Za Zhi*. 2021;42(10):1823–8.
- Qin F, Zhang JF, Luo MY, Feng Y, Ge R, Yan Y, et al. Molecular genetics and epidemiological characteristics of HIV-1 epidemic strains in various sexual risk behaviour groups in developed Eastern China, 2017–2020. *Emerg Microbes Infect*. 2022;11(1):2326–39.
- Zhang J, Guo Z, Pan X, Zhang WJ, Yang JZ, Ding XB, et al. Highlighting the crucial role of Hangzhou in HIV-1 transmission among men who have sex with men in Zhejiang, China. *Sci Rep*. 2017;7(1):13892.
- Wang ZY, Zhang M, Zhang RF, Liu L, Shen YZ, Wang JR, et al. Diversity of HIV-1 genotypes and high prevalence of pretreatment drug resistance in newly diagnosed HIV-infected patients in Shanghai, China. *BMC Infect Dis*. 2019;19(1):313.
- Hassan AS, Pybus OG, Sanders EJ, Albert J, Esbjörnsson J. Defining HIV-1 transmission clusters based on sequence data. *AIDS*. 2017;31(9):1211–22.
- Zhong P. Progress in research and practice of molecular epidemiology of HIV-1. *Electron J Emerg Infect Diseases*. 2019;4(3):137–44.
- Vrancken B, Zhao B, Li X, Han X, Liu H, Zhao J, et al. Comparative circulation dynamics of the five main HIV types in China. *J Virol*. 2020;94(23):e00683–e720.
- Zhang J, Guo Z, Yang J, Pan X, Jiang J, Ding X, et al. Genetic diversity of HIV-1 and transmitted drug resistance among newly diagnosed individuals with HIV infection in Hangzhou. *China J Med Virol*. 2015;87(10):1668–76.
- Oster AM, Wertheim JO, Hernandez AL, Ocfemia MC, Saduvala N, Hall HI. Using molecular HIV surveillance data to understand transmission between subpopulations in the United States. *JAIDS J Acquir Immune Defic Syndr*. 2015;70(4):444–51.
- Zhang JF, Xu K, Jiang J, Qin F, Ding XB, Zhong P, et al. Combining molecular network analysis and field epidemiology to quantify local HIV transmission and highlight ongoing epidemics. *Int J Infect Dis*. 2023;128:187–93.
- Zhou Y, Lu J, Zhang Z, Sun Q, Xu XQ, Hu HY. Characteristics of the different HIV-1 risk populations based on the genetic transmission network of the newly diagnosed HIV cases in Jiangsu, Eastern China. *Heliyon*. 2023;9(12):e22927.
- Zhang F, Yang Y, Liang N, Liang HY, Chen YZ, Lin ZS, et al. Transmission network and phylogenetic analysis reveal older male-centered transmission of CRF01_AE and CRF07_BC in Guangxi, China. *Emerg Microbes Infect*. 2023;12(1):2147023.
- Prevention detecting and responding to HIV transmission clusters: A guide for health department. 2018.
- Ye JR, Chen J, Wang J, Wang YC, Xing H, Yu FT, et al. CRF07_BC is associated with slow HIV disease progression in Chinese patients. *Sci Rep*. 2022;12(1):3773.
- Cheng Z, Yan H, Li Q, Ablan SD, Kleinpeter A, Freed EO, et al. Enhanced transmissibility and decreased virulence of HIV-1 CRF07_BC may explain its rapid expansion in China. *Microbiol Spectr*. 2022;10(4):e0014622.
- Lin PH, Lai CC, Yang JL, Huang HL, Huang MS, Tsai MS, et al. Slow immunological progression in HIV-1 CRF07_BC-infected injecting drug users. *Emerg Microbes Infect*. 2013;2(12):e83.
- Ge ZW, Feng Y, Li K, Lv B, Zaongo SD, Sun J, et al. CRF01_AE and CRF01_AE cluster 4 are associated with poor immune recovery in Chinese patients under combination antiretroviral therapy. *Clin Infect Dis*. 2021;72(10):1799–809.
- Sun X, Feng C, Feng L, Xiao TC, Zhao XR, Liu H, et al. An exploratory transmission mode of HIV/AIDS among older people based on data from multiple sources in China. *Sci Rep*. 2022;12(1):16077.
- Odimegwu CO, Mutanda N. Covariates of high-risk sexual behaviour of men aged 50 years and above in sub-Saharan Africa. *SAHARA J*. 2017;14(1):162–70.
- Chen X, Qin CW, Chen RF, Huang YX. Epidemiological profile and molecular genetic characterization of HIV-1 among female sex workers and elderly male clients in Guangxi. *China Emerg Microbes Infect*. 2021;10(1):384–95.
- Zhu J, Hu D, Yin YQ, Zhu ZB, Wang N, Wang B. HIV prevalence and correlated factors among male clients of female sex workers in a border region of China. *PLoS ONE*. 2019;14(11):e0225072.
- Chen M, Zhu QM, Xing H, Chen HC, Jin XM, Dong LJ, et al. The characteristics of pretreatment HIV-1 drug resistance in western Yunnan, China. *Epidemiol Infect*. 2020;8(148):e102.
- Xu XS, Luo LH, Song C, Li JJ, Chen HH, Zhu QY, et al. Survey of pretreatment HIV drug resistance and the genetic transmission networks among HIV-positive individuals in southwestern China, 2014–2020. *BMC Infect Dis*. 2021;21(1):1153.
- Liu L, Dong AB, Liao LJ, Feng Y, Shao YM, Shu L, et al. Survey of pretreatment HIV drug resistance and genetic transmission network analysis among HIV patients in a high drug-use area of Southwest China. *Curr HIV Res*. 2019;17:441–51.
- Kang RH, Liang SJ, Ma YL, Liang S, Xiao L, Zhang XH, et al. Pretreatment HIV drug resistance in adults initiating antiretroviral therapy in China, 2017. *Infect Dis Poverty*. 2020;9(1):54.
- Zuo LL, Liu K, Liu HL, Hu YH, Zhang ZJ, Qin JR, et al. Trend of HIV-1 drug resistance in China: a systematic review and meta-analysis of data accumulated over 17 years (2001–2017). *EclinicalMedicine*. 2020;5(18):100238.

29. Sun CQ, Li JJ, Liu XY, Zhang Z, Qiu T, Hu HY, et al. HIV/AIDS late presentation and its associated factors in China from 2010 to 2020: a systematic review and meta-analysis. *AIDS Res Ther.* 2021;18(1):96.

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