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Factors affecting HPV infection in U.S. and Beijing females: A modeling study

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Background: Human papillomavirus (HPV) infection is an important carcinogenic infection highly prevalent among many populations. However, independent influencing factors and predictive models for HPV infection in both U.S. and Beijing females are rarely confirmed. In this study, our first objective was to explore the overlapping HPV infection-related factors in U.S. and Beijing females. Secondly, we aimed to develop an R package for identifying the top-performing prediction models and build the predictive models for HPV infection using this R package.

Methods: This cross-sectional study used data from the 2009–2016 NHANES (a national population-based study) and the 2019 data on Beijing female union workers from various industries. Prevalence, potential influencing factors, and predictive models for HPV infection in both cohorts were explored.

Results: There were 2,259 (NHANES cohort, age: 20–59 years) and 1,593 (Beijing female cohort, age: 20–70 years) participants included in analyses. The HPV infection rate of U.S. NHANES and Beijing females were, respectively 45.73 and 8.22%. The number of male sex partners, marital status, and history of HPV infection were the predominant factors that influenced HPV infection in both NHANES and Beijing female cohorts. However, condom application was not an independent influencing factor for HPV infection in both cohorts. R package Modelbest was established. The nomogram developed based on Modelbest package showed better performance than the nomogram which only included significant factors in multivariate regression analysis.

Conclusion: Collectively, despite the widespread availability of HPV vaccines, HPV infection is still prevalent. Compared with condom promotion, avoidance of multiple sexual partners seems to be more effective for preventing HPV infection. Nomograms developed based on Modelbest can provide improved personalized risk assessment for HPV infection. Our R package Modelbest has potential to be a powerful tool for future predictive model studies.

KEYWORDS

human papillomavirus, vaccine, prevalence, NHANES, Beijing, Modelbest, influencing factor

Introduction

Human papillomavirus (HPV) infection is an important carcinogenic infection highly prevalent among many populations. It is transmitted *via* sexual intercourse or intimate interpersonal contact (1-3). There are over 200 different HPV types. Different HPV subtypes can cause different pathological lesions, and more than 60 types of HPV are found predominantly or exclusively in the anogenital tract (4).

HPVs can be broadly divided into low- and high-risk types. Low-risk HPVs (e.g., HPV 6 and 11) cause nonmalignant lesions (e.g., genital warts and recurrent respiratory papillomatosis). While the high-risk HPVs (e.g., HPV 16 and 18) may lead to various cancers (e.g., nearly all cervical cancers, and most vulvar, vaginal, oropharyngeal, anal, and penile cancers (5)). The oncoproteins E6 and E7 of high-risk HPVs are responsible for HPV oncogenesis (6, 7). The cervical transition zone is extremely vulnerable to HPV infections, and it is the predominant site of origin of cervical cancer (8).

Studies have shown that HPV infection might be associated with sexual behaviors, parity, marital status and age (9-13). However, the relation between condom use and HPV infection remains controversial (13). Exploring HPV-infection associated factors is important for HPV prevention and management. This especially so moving to an era of precision medicine, with artificial intelligence and machine learning finding increasing applications in health care (14-16). Accordingly, prediction models based upon risk factors and/or other parameters, which can serve as a convenient tool for individualized risk estimation as well as risk prevention, have attracted increasing attention (17). Nevertheless, very few studies till date have built a stable and generalizable model for predicting HPV infection risk. Based on the afore-mentioned background, the present study has been executed with two principal goals in mind. One was to identify the overlapping HPV infection-related factors in female populations from the U.S. and Beijing, China. The secondary objective included developing an R package for identifying the top-performing prediction models as well as building predictive models for HPV infection via this R package.

Methods and materials

Study design and participant selection

This was a retrospective cross-sectional analysis in two cohorts. One cohort is from the National Health and Nutrition Examination Survey (NHANES), which is a national crosssectional study designed to provide nationally representative data on the nutritional and health status of the noninstitutionalized, civilian U.S. population. Data in NHANES were collected through participant interviews and physical examinations. All NHANES study protocols were approved by the National Center for Health Statistics (NCHS) Institutional Review Board and all subjects provided written informed consent before data collection. Detailed descriptions of the methodology and data can be accessed at http://www.cdc.gov/ nchs/nhanes/. Since the HPV vaccine was only extensively pursued after 2006 in U.S., we limited our study to female participants in NHANES between 2009 and 2016. The exclusion criteria included: (1) participants aged <18 or \geq 60 years; (2) or with a history of malignancy; (3) or a history of hysterectomy.

Another cohort was from a Beijing Municipal Federation of Trade Unions (BMFTU)-funded cervical screening program for female workers in Beijing. The information needed for this study was sent out to female union members by BMFTU. Females who fulfilled the inclusion/exclusion criteria were invited to take part in this study at Beijing Rehabilitation Hospital. Eligible females were only included as participants in this study after informed consent had been obtained. All participants received questionnaires and HPV genotype testing between April 8, 2020 and May 8, 2020. This study was reviewed and approved by the Ethical Committee of Beijing Rehabilitation Hospital. For this study, the inclusion criteria included: (1) females with a history of having had sex; (2) aged between 20 and 70 years. The exclusion criteria were: (1) history of malignancy; (2) or history of cervical conization or hysterectomy; (3) or history of pelvic radiation therapy; (4) or history of immune disorders; (5) or pregnancy or childbirth within 6 months. For the data analysis of this study, cases with incomplete information were excluded. After completing the questionnaire, the participants received gynecological examinations including HPV genotype test. The detailed procedure of the HPV examination has been described in our previous study (18). Briefly, cervical cells of each participant were sampled using a cervix brush, placed in a vial containing 3 mL of isotonic sodium chloride and then stored at 4°C. Fifteen HPV genotypes included 13 high-risk genotypes (HPV 16, 18, 31, 33, 35, 39, 45, 51, 52, 56, 58, 59, and 68) and two low-risk genotypes (HPV 6 and 11) were examined. This HPV testing method can detect concentrations as low as 1.0×10^3 copies/mL of HPV, and the specificity is more than 98%. Detailed content of the questionnaire and the HPV test are shown in Supplementary Table 1.

Cohort definition and feature evaluation

For the NHANES cohort, the participants were randomly divided in a 7:3 ratio into the training and test-evaluation sets (the test-evaluation set was further divided to test and evaluation sets at a ratio of 2:1) using the "sample" function in R software (version 4.1.3). The training and test sets were used to select features and develop the prediction model. The evaluation set was used to evaluate the performance of prediction model. We also evaluated the prediction model in the combined test-evaluation set. Dividing the data into three data sets (rather

than the training-validation binary divisions) could avoid model overfitting and provide an approach to validate the model (19).

For the feature evaluation, LASSO and multivariate logistic regression analysis were performed successively *via* R packages glmnet (v.4.1-3) and rms (v.6.2-0) on the training set. Features with *P*-values <0.05 in both LASSO and multivariate logistic regression analysis were regarded as independent factors associated with HPV infection. R package forestplot (v.2.0.1) was used to graphically represent results of multivariate logistic regression analysis. Same methods were also performed on the Beijing female cohort.

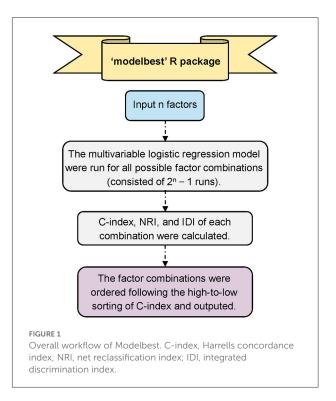
Method and principle of the R package modelbest

The Modelbest package (https://github.com/yhx68/best/ blob/main/Modelbest_0.1.0.tar.gz) iterates through all possible variables' combinations in a multivariate logistic regression model and outputs all models ordered following high-to-low sorted order based on the Harrells concordance index (Cindex); meanwhile, this package can automatically identify the statistically significant variables in multivariate logistic regression among the input variables, and calculate the net reclassification index (NRI) and integrated discrimination index (IDI) of each model compared with the baseline prediction model (i.e., the model that contains only the significant variables in multivariate logistic regression analysis) (Figure 1). The output results include the model formula, number of factors in the model, C-index, NRI and IDI, etc. For the input file of Modelbest, all variables need to be categorical variables. The first column is always the outcome variable, and the following columns are always the candidate factors for model construction. The candidate factors can be the statistically significant variables in logistic regression analysis, beyond that, factors deemed clinically important can also be considered.

The C-index has a range from 0.5 to 1.0, 0.5 represents random prediction, while 1.0 indicates perfect prediction. IDI and NRI are newer techniques to quantify the improvement of a new model's predictive performance over a baseline prediction model (20). NRI and IDI values larger than zero indicate the new model has better performance than the baseline prediction model.

Nomogram construction and evaluation

The consensus guidelines (i.e., the number of covariates per model was limited by the number of events, and at least 10 events were required for each covariate) (21–23) were followed in constructing the prediction models for HPV infection. To avoid potentially important variables being missed, the Modelbest package was applied to identify the top-performing



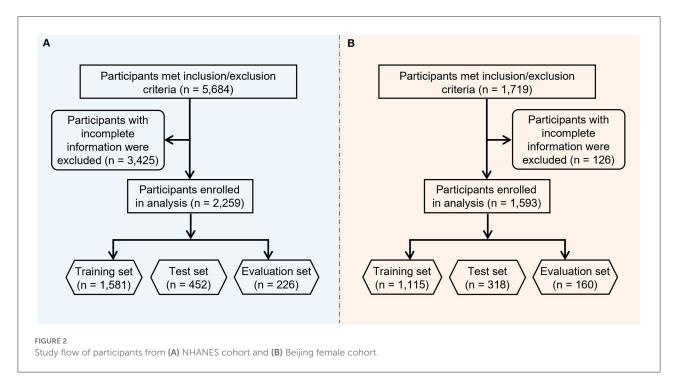
predictive model for HPV infection. Variables included in the Modelbest package were statistically significant variables in the LASSO regression analysis. Results of Modelbest package are shown in Supplementary Table 2 (for NHANES cohort) and Supplementary Table 3 (for Beijing female cohort). The variables' combination with the highest C-index, NRI, and IDI was selected for constructing nomogram. Rule of thumb for multivariable models (i.e., at least 10 outcome events per variable) was considered when choosing the best combination. R packages rms (v.6.2-0) and regplot (v. 1.1) were applied to obtain nomogram.

To evaluate the nomogram's discrimination ability, the receiver operating characteristic (ROC) plot and area under ROC curve (AUC), C-index, NRI, IDI, calibration curves and decision curve analysis (DCA) were performed in R using rms (v.6.2-0), ROCR (v. 1.0-11), Hmisc (v. 4.6-0), PredictABEL (v. 1.2-4) and rmda (v. 1.6) packages. Finally, online calculators for HPV infection were constructed *via* R packages DynNom (v. 5.0.1), shiny (v. 1.7.1) and rsconnect (v. 0.8.25). All analysis in this study was conducted using R software (version 4.1.3). Supplementary Table 4 summarizes the R packages used in this study and the corresponding analysis.

Results

Participant characteristics

The study flowchart is shown in Figure 2. Primary and secondary objectives were fulfilled. For the NHANES



cohort of 2009–2016, a total of 5,684 participants met the inclusion/exclusion criteria (see methods section), following the exclusion of 3,425 participants with incomplete information, 2,259 participants were included in the final analysis. Among these 2,259 participants, 45.73% (1,033/2,259) were HPV positive.

For the Beijing female cohort, 1,719 female union workers from various industries participated in this program, following the exclusion of 126 participants with missing data, 1,593 female workers in total were included in the current study's analysis. These participants came from the security industry, governmental agencies, cleaning departments, construction industry, transportation industry, education industry, express delivery industry, property sector and medical industry. Among these 1,593 participants, 8.22% (131/1,593) were HPV positive. Supplementary Tables 5, 6 summarize the general characteristics of the NHANES and Beijing female cohorts.

Factors associated with HPV infection

In the NHANES cohort, 12 factors were selected for multivariate logistic analysis after LASSO regression, and eight factors (including age, race, number of male sex partners, number of vaginal deliveries, marital status, smoking status, drinking status, and history of HPV infection) were found to be independent factors significantly associated with HPV infection (Figure 3). Notably, factors such as condom use, menstruation status, educational levels, sleep disorder history, chlamydia infection history, and daily sedentary activity were not significantly associated with HPV infection.

In the Beijing female cohort, 13 factors were selected for multivariate logistic analysis after LASSO regression, and five factors (including number of male sex partners, marital status, history of HPV infection, knowledge about HPV prevention and HPV vaccination status) were found to be independently associated with HPV infection (Figure 4). However, factors such as condom use, menstruation status and educational levels were not significantly associated with HPV infection.

Modelbest-models show better performance than log-models for predicting HPV infection

Based on the independently associated factors, the predictive models for HPV infection were constructed (namely log-model). Besides, predictive models based on R package Modelbest were also constructed (namely Modelbest-model). The C-index of the log-models were, respectively 0.792 and 0.691 for the NHANES and Beijing female evaluation sets. While the C-index of the Modelbest-models were, respectively 0.812 and 0.747 for the NHANES and Beijing female evaluation sets. Table 1 shows the NRI and IDI of Modelbest-models than the log-models. The ROC, calibration, and DCA curves of the prediction models (log-models and Modelbest-models) in the NHANES and Beijing female cohorts are shown in Supplementary Figures 1–4.

ge, n (%) 20–29 y 30–39 y 40–49 y 50–59 y ge at first sex, n (%) < 18 y ≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic Other race	488 (21.6) 718 (31.8) 686 (30.4) 367 (16.2) 1,473 (65.2) 786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3) 225 (10)		Reference 0.76 (0.55, 1.05) 0.68 (0.48, 0.96) 0.67 (0.45, 0.99) Reference 0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	0.101 0.029 0.045 0.321 0.976
20–29 y 30–39 y 40–49 y 50–59 y ge at first sex, n (%) < 18 y ≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	718 (31.8) 686 (30.4) 367 (16.2) 1,473 (65.2) 786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		0.76 (0.55, 1.05) 0.68 (0.48, 0.96) 0.67 (0.45, 0.99) Reference 0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	0.029 0.045 0.321
40–49 y 50–59 y ge at first sex, n (%) < 18 y ≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	686 (30.4) 367 (16.2) 1,473 (65.2) 786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		0.68 (0.48, 0.96) 0.67 (0.45, 0.99) Reference 0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	0.029 0.045 0.321
40–49 y 50–59 y ge at first sex, n (%) < 18 y ≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	367 (16.2) 1,473 (65.2) 786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		0.67 (0.45, 0.99) Reference 0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	0.045 0.321
50–59 y ge at first sex, n (%) < 18 y ≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	367 (16.2) 1,473 (65.2) 786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		0.67 (0.45, 0.99) Reference 0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	0.321
ge at first sex, n (%) < 18 y ≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	1,473 (65.2) 786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		Reference 0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	0.321
< 18 y ≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	
≥ 18 y ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	786 (34.8) 928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		0.88 (0.68, 1.13) Reference 1.01 (0.71, 1.43)	
ace, n (%) Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	928 (41.1) 349 (15.4) 525 (23.2) 232 (10.3)		Reference 1.01 (0.71, 1.43)	
Non-Hispanic White Mexican American Non-Hispanic Black Other Hispanic	349 (15.4) 525 (23.2) 232 (10.3)		1.01 (0.71, 1.43)	0.976
Mexican American Non-Hispanic Black Other Hispanic	349 (15.4) 525 (23.2) 232 (10.3)	⊢∳⊣ │⊢∳⊣	1.01 (0.71, 1.43)	0.976
Non-Hispanic Black Other Hispanic	525 (23.2) 232 (10.3)	'∐⊢•⊣	· · · /	0.070
Other Hispanic	232 (10.3)		1.83 (1.36, 2.45)	< 0.001
	. ,		1.31 (0.90, 1.89)	0.160
		Ц. С	1.21 (0.80, 1.83)	0.365
nnual family income, n (%)	(10)		1.21 (0.00, 1.03)	0.505
< \$20,000	464 (20.5)		Reference	
≥ \$20,000	1,795 (79.5)		0.77 (0.57, 1.03)	0.078
	1,735 (73.5)	•	0.77 (0.57, 1.05)	0.078
umber of male sex partners, n (%) 1	266 (11.8)		Reference	
≥2	1,993 (88.2)			<0.001
	1,995 (00.2)		2.65 (1.74, 4.15)	<0.001
umber of vaginal deliveries, n (%)	1,094 (48.4)		Reference	
< 2	,			0.025
≥2	1,165 (51.6)		0.77 (0.61, 0.97)	0.025
arital status, n (%)	1 400 (66)		Deferrer	
Married/cohabiting	1,492 (66)		Reference	10 001
Living alone	767 (34)		2.18 (1.71, 2.79)	<0.001
ducation levels, n (%)	04 (4 0)		D (
Less than 9th grade	94 (4.2)		Reference	0.050
9-11th grade	269 (11.9)		0.98 (0.52, 1.87)	0.950
High school Grad/GED or equivalent	459 (20.3)		1.03 (0.55, 1.91)	0.938
Some college or AA degree	816 (36.1)		0.88 (0.48, 1.62)	0.667
College graduate or above	621 (27.5)	I - ◆ H	0.66 (0.35, 1.25)	0.201
moking status, n (%)			5 (
Non-current smoker	1,710 (75.7)		Reference	
Current smoker	549 (24.3)	⊢ •⊣	1.37 (1.04, 1.80)	0.027
rinking status, n (%)			D (
Light drinking	825 (36.5)		Reference	
Heavy/Binge drinking	1,434 (63.5)	∣⊷⊣	1.55 (1.22, 1.97)	<0.001
istory of HPV infection, n (%)			- /	
No	2,023 (89.6)		Reference	
Yes	236 (10.4)	⊫⊷⊣	1.63 (1.14, 2.34)	0.008
PV vaccination status, n (%)				
No	2,075 (91.9)		Reference	
Yes	184 (8.1)	H ✦→-1	1.32 (0.86, 2.03)	0.208

Forest plot of multivariable logistic regression analysis in the training set of NHANES cohort. The x-axis represents the odds ratio scale. An odds ratio >1 indicates increased odds of HPV infection, whereas an odds ratio <1 suggests decreased odds of HPV infection. CI, confidence interval.

Figure 5 contains nomograms constructed based on the Modelbest-models for NHANES and Beijing female cohorts. We also constructed online calculators based on the Modelbest-models for predicting probability of HPV infection (https://cervixuteri.shinyapps.io/HPV1/ is for NHANES cohort; https://cervixuteri.shinyapps.io/HPV2/ is for Beijing female cohort).

Discussion

Taken together, we found number of male sex partners, marital status, and history of HPV infection were factors

significantly associated with HPV infection in both the NHANES and Beijing females. Besides, the NHANES cohort revealed that smoking and drinking status were also independently associated with HPV infection. Interestingly, HPV vaccination status in the NHANES cohort showed no significant association with HPV infection. Based on the R package Modelbest developed in this study and the data of NHANES and Beijing females, we built two prediction models, respectively targeted for NHANES and Beijing females, these two models exhibited good prediction performance for HPV infection.

Number of male sex partners, n (%) 1 ≥ 2				
≥ 2	1,049 (94.1)		Reference	
	66 (5.9)		2.68 (1.24, 5.47)	0.009
Number of previous pregnancies, n (%)				
≤1	471 (42.2)		Reference	
2	345 (30.9)	M I	0.74 (0.40, 1.32)	0.313
≥ 3	299 (26.8)	H I	0.72 (0.37, 1.36)	0.323
Number of deliveries, n (%)				
≤ 1	958 (85.9)		Reference	
≥ 2	157 (14.1)	H∳H	0.78 (0.31, 1.77)	0.577
Menstruation status, n (%)				
Regular	673 (60.4)		Reference	
Irregular	442 (39.6)	I III	1.64 (1.00, 2.71)	0.051
Main contraceptive measures, n (%)				
Condom	658 (59)		Reference	
Others	457 (41)	H ∳ -I	1.27 (0.78, 2.08)	0.339
Marital status, n (%)				
Living alone	20 (1.8)		Reference	
Married/cohabiting	1,095 (98.2)	*	0.16 (0.06, 0.49)	<0.00
Smoking status, n (%) Non-current smoker	1,047 (93.9)		Deference	
Current smoker	68 (6.1)		Reference	0.052
History of cervical leisions, n (%)			2.16 (0.94, 4.53)	0.052
No	907 (81.3)		Reference	
Yes	208 (18.7)		0.62 (0.31, 1.18)	0.164
History of HPV infection, n (%)	200 (10.7)		0.02 (0.01, 1.10)	0.101
No	1,019 (91.4)		Reference	
Yes	96 (8.6)		7.54 (4.05, 13.93)	< 0.00
Knowledge about HPV prevention, n (%)	()		1 1.01 (1.00, 10.00)	0.00
No	680 (61)		Reference	
Yes	435 (39)		2.24 (1.36, 3.72)	0.002
Hygiene practices during sex, n (%)				
No	39 (3.5)		Reference	
Yes	1,076 (96.5)	Hell I	0.49 (0.19, 1.49)	0.168
Regular cervical cancer screening, n (%)				
No	326 (29.2)		Reference	
Yes	789 (70.8)	•	0.65 (0.38, 1.12)	0.119
HPV vaccination status, n (%)				
No	1,062 (95.2)		Reference	
Yes	53 (4.8)	₩	0.10 (0.01, 0.52)	0.030
		0 5 1	10	

Forest plot of multivariable logistic regression analysis in the training set of Beijing female cohort. The x-axis represents the odds ratio scale. An odds ratio >1 indicates increased odds of HPV infection, whereas an odds ratio <1 suggests decreased odds of HPV infection.

It is well known that HPV is transmitted predominately through sexual contact, and our findings (i.e., multiple sex partners was independent risk factor for HPV infection in multivariate logistic regression analysis) strongly supported this theory. Earlier studies indicated that two-thirds of women were infected with HPV within 2 years of onset of sexual activity and young women with multiple sex partners are at highest risk of HPV infection (24, 25). Considering the first sexual relationship carries a substantial risk (26, 27), sex education is therefore particularly important for preventing HPV infections. To reduce risk of HPV infection, there is a clear need to provide the general public and especially teenagers with sex education and strategies (such as delayed sexual debut, minimizing the number of sexual partners, and steering away from promiscuous male partners) to prevent HPV infection. Moreover, compared with married/cohabiting women, women living alone may have a higher chance to have a risky sexual partner or multiple sexual partners, therefore, the marital status

	Continuous NRI		IDI		
	OR (95%CI)	P-value	OR (95%CI)	P-value	
NHANES training set	0.128 (0.029-0.227)	0.011	0.008 (0.003-0.012)	<0.001	
NHANES test-evaluation set	0.208 (0.057-0.359)	0.007	0.008 (0.001-0.015)	0.018	
NHANES test set	0.234 (0.049-0.418)	0.013	0.013 (0.003-0.024)	0.013	
NHANES evaluation set	0.328 (0.076- 0.581)	0.011	0.038 (0.013-0.062)	0.003	
Beijing female training set	0.325 (0.104-0.545)	0.004	0.013 (-0.001-0.026)	0.062	
Beijing female test-evaluation set	0.325 (0.040-0.610)	0.025	0.019 (-0.002-0.040)	0.078	
Beijing female test set	0.271 (-0.007-0.549)	0.056	0.013 (-0.010-0.036)	0.261	
Beijing female evaluation set	0.414 (-0.143–0.972)	0.145	0.036 (-0.029-0.101)	0.281	

TABLE 1 NRI and IDI of the Modelbest-developed models for predicting HPV infection than models developed on multivariate logistic regression.

OR, odds ratio; NRI, net reclassification index; IDI, integrated discrimination index; CI, confident interval.

is also a factor significantly associated with HPV infection in this study. Despite consistent condom use being an effective means against HPV infection, HPV can be transmitted in the genital region through non-penetrative sexual contact (28). Both the NHANES and Beijing female cohorts revealed no significant association between condom use with the probability of cervical HPV infection in the current study. We guess that for the people with high-risk sexual behaviors, the protective effect of condoms against HPV was limited. However, considering the data of the current two cohorts were all retrospective, with the potential risk of confounding and selection bias. Future prospective studies with larger sample sizes are needed to validate this finding (i.e., non-significant association between condom use and HPV infection).

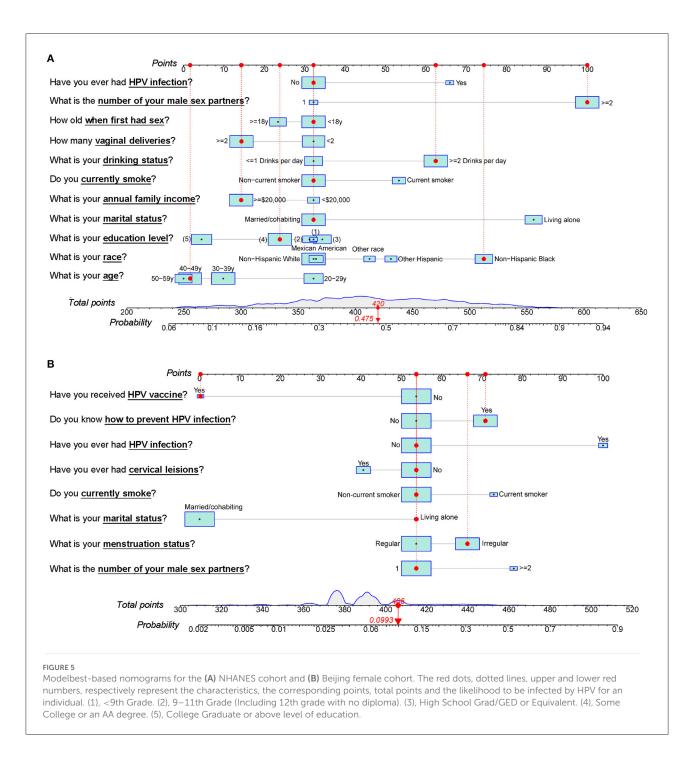
The history of HPV infection was also an independent risk factor for HPV infection in both the NHANES and Beijing female cohorts. It is undeniable that, for most people, HPV infection is transient and can be cleared or become undetectable after 1–2 years of infection (29, 30). Nevertheless, in some instances, HPV infection can be latent and remain at risk for re-activation. This means that women who have ever been infected with HPV (either with or without HPV-vaccination) should pay sufficient attention to HPV and participate in regular re-examinations for HPV.

For the NHANES female cohort, we also found that smoking status and drinking status were the independent risk factors for HPV infection. These findings are in accordance with the previous studies. Smoking has been confirmed in relation to HPV infection (including the HPV incidence, prevalence, persistence, and DNA load) in mounting studies (31–33). Recent study revealed tobacco smoke may interact synergistically with HPV by promoting HPV replication, host DNA damage, HPV E6/E7 expression, and disrupting host immune response against HPV (34). Alcohol drinking behaviors have also been suggested associated with risk of HPV persistence (35). Moreover, another study in U.S. revealed, binge drinkers from young adults were more likely to have HIV/HPV co-infection high-risk behaviors (36).

In the Beijing female cohort, the smoking status showed no significant association with HPV infection in the multivariate logistic regression analysis, this may be partially due to the small sample size, since only 6.10% of the participants (68/1,115) in the training set were current-smokers. In comparison, in the NHANES training set, 24.10% participants (430/1,785) were current-smokers. The statistical results of the Beijing female cohort would probably be more convincing if the sample size was increased. In the Beijing female cohort, we also found that knowledge about HPV prevention was a risk factor for HPV infection. To some extent, this may reflect the lack of public education about HPV in China, as plenty of women learned about HPV and the HPV prevention strategies after the diagnosis of HPV infections.

Based on the NHANES females, we observed no significant correlation between HPV infection and vaccination. This may be due to the HPV vaccine being a newer vaccine with limited application time. The Food and Drug Administration (FDA) only licensed the first HPV vaccine on June, 2006 for females 9-26 years, and the most recent nine-valent HPV vaccine was not approved by the FDA until 2014. Some women may have been infected with HPV prior to HPV vaccination, nevertheless, the vaccine is not therapeutic and may only prevent future infection. In addition, the Roche HPV linear array applied in the NHANES females can detect 37 HPV types, but currently available HPV vaccines can prevent at most of nine types of high-risk HPV infection. It is also possible that a few women did not develop effective immunity response after vaccination. This suggests that, for sexually active women, HPV vaccination is only part of a successful strategy for preventing HPV infection and cervical cancer, and that regular cervical screening and safer sex approaches still play an integral role.

Nomograms, as a reliable and convenient tool for individual risk prediction, have been widely used in the medical field and for guiding clinical decision (37, 38). The C-index and AUC are



the main reference for evaluating the predictive performance of a nomogram. Typically, AUC and C-index values larger than 0.7 are considered acceptable discrimination. Additionally, other methods such as calibration and DCA curves, NRI and IDI have also been applied for evaluating the reliability and improvements of the nomogram. To date, there is no uniform screening standard for variables incorporated into the clinical prediction model, but it is recommended to consider both statistically significant results and universally accepted clinical knowledge when constructing a clinical prediction model (39–41). The R package Modelbest developed in this study could help researchers evaluate the performance of every possible combination of variables and choose the best performing combination based on objective evaluation indicators (C-index, NRI, and IDI). Based on the data of NHANES and Beijing female cohorts, we developed nomograms for predicting HPV infection

using R package Modelbest. Modelbest-based predictive models showed better performance than the models developed only based on significant variables in multivariate logistic regression analysis. To our knowledge, this is the first study to predict HPV infection through nomograms and, in addition, to establish a powerful R package for future predictive model studies.

However, there are several limitations to this study: First, its main limitation is that it is retrospective. Second, the predictive models built in this study lack external validation. For example, the research on the Beijing female cohort was conducted among Beijing female union workers who live in areas with a great availability of medical services and are provided ThinPrep cytology test and HPV testing yearly. This characteristic of the cohort limits the generalizability of the Beijing-female-based predictive model to other regions/populations. In comparison, the NHANES are a series of nationally representative crosssectional surveys for community-dwelling U.S. population using a complex multistage probability sampling design, which allows for generalizability of the NHANES-based predictive model to the U.S. population. Third, the study covers a small sample size, which may lead to some factors being insignificant in the present study. The heterogeneity in study design and populations between NHANES and Beijing female cohorts is another limitation, which was partly reflected in the differences in HPV detection methodologies (i.e., HPV genotype testing in NHANES cohort could detect 37 types of HPV, while testing in the Beijing female cohort could detect 15 types). The enormous population heterogeneity prevents any standardization of data across these two populations. However, this study was not designed to compare the HPV infection rates between different cohorts, but to find the overlapping risk factors for HPV infection among different regions and construct targeted prediction models for different regions.

Conclusion

Collectively, in this study, we found number of male sex partners, marital status and history of HPV infection were significantly correlated with HPV infection in both NHANES and Beijing female cohorts, and the multi-cohort findings increase the credibility and generalizability of our conclusion. Beyond that, we developed an R package (Modelbest) which could serve as a powerful tool for identifying the top-performing prediction models in future studies. The Modelbest-developed nomograms in this study showed good predictive performance and can provide risk assessments for HPV infection that are both personalized and evidence-based.

Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

Ethics statement

The study involving Beijing female participants was reviewed and approved by the Ethical Committee of Beijing Rehabilitation Hospital. The participants provided their written informed consent to participate in this study. Ethics approval and informed consent were not required for the analysis of the publicly available NHANES data.

Author contributions

HYu, BZ, YZho, and HYa designed the research. YZha performed HPV testing. WL, YL, FZ, XG, ZT, and HLi conducted questionnaire survey. HYa, YX, RG, and HLiu analyzed data and developed R package Modelbest. HYa wrote the manuscript. HYu, BZ, and YZho proofread and revised the manuscript. All authors contributed to the article and approved submitted version.

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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/fpubh. 2022.1052210/full#supplementary-material

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