Joe Brew

PHC 6053 - Assignment 8

Part 1: SAS outputs (for each question

Part 2: SAS and R code

Part 3: Full SAS output

*Note: I used R as a supplement to SAS for this assignment. I did the entirety of the assignment in both, and code is side by side. Because I have find certain dataset manipulations to be hard in SAS (subsetting, dealing with categorical data, etc.), I relied largely on R for number 5. I'm worse at SAS than I am at R, and make a note of when I know my SAS output is incorrect (but my R output is correct)

PART 1: SAS OUTPUT FOR HOMEWORK

1. NA

2. **AGE**

Analysis of Maximum Likelihood Estimates						
Parameter	DF	Estimate	Standard Error	Wald Chi-Square		
Intercept	1	-5.4506	0.7277	56.0984	<.0001	
AGE	1	0.0877	0.0119	54.0266	<.0001	

Odds Ratio Estimates					
Point 95% Wald Effect Estimate Confidence Limi					
AGE	1.092	1.066	1.118		

вмі

Analysis of Maximum Likelihood Estimates						
Parameter	Parameter DF Estimate Standard Wald Chi-Square Pr > Ch					
Intercept	1	-2.0949	0.7054	8.8191	0.0030	
BMI	1	0.0745	0.0268	7.7077	0.0055	

Odds Ratio Estimates					
Effect	Point 95% Wald t Estimate Confidence Limits				
BMI	1.077	1.022	1.136		

BMIGROUP

Analysis of Maximum Likelihood Estimates							
Parameter DF Estimate Standard Wald Chi-Square Pr > Ch							
Intercept		1	-0.4875	0.1439	11.4827	0.0007	
BMIGROUP	3	1	0.5323	0.1966	7.3334	0.0068	
BMIGROUP	4	1	0.6698	0.2860	5.4840	0.0192	

Odds Ratio Estimates				
Effect	Point 95% Wald Estimate Confidence Limits			
BMIGROUP 3 vs 2	1.703	1.158	2.503	
BMIGROUP 4 vs 2	1.954	1.115	3.423	

Given my initial difficulty with this one, I wanted to confirm my answer in R.

Category	Estimate	Std. Error	Z	Р	OR	95% CI
Intercept	-0.4875	0.1439	-3.389	0.0007	0.614	0.462-
						0.812
Overweight	0.5323	0.1966	2.708	0.006768	1.703	1.160-
						2.509
Obese	0.6698	0.2860	2.342	0.0192	1.954	1.117-
						3.440

SEX

Analysis of Maximum Likelihood Estimates						
Parameter DF Estimate		Standard Error	Wald Chi-Square			
Intercept	1	-0.0382	0.2990	0.0163	0.8983	
SEX	1	-0.0740	0.1821	0.1651	0.6845	

Odds Ratio Estimates					
Effect	Point Estimate				
SEX	0.929	0.650	1.327		

BPMEDS

Analysis of Maximum Likelihood Estimates						
Parameter	DF	Estimate	Standard Error	Wald Chi-Square		
Intercept	1	-0.3995	0.1000	15.9502	<.0001	
BPMEDS	1	1.6771	0.2919	33.0049	<.0001	

Odds Ratio Estimates					
Effect	Point 95% Wald Estimate Confidence Limits				
BPMEDS	5.350	3.019	9.481		

PREVSTRK

Analysis of Maximum Likelihood Estimates						
Parameter DF Estimate Standard Wald Chi-Square Pr > Chi						
Intercept	1	-0.2138	0.0919	5.4094	0.0200	
PREVSTRK	1	2.8526	1.0391	7.5368	0.0060	

Odds Ratio Estimates					
Effect	Point Estimate				
PREVSTRK	17.332	2.262	132.832		

3.

Analysis of Maximum Likelihood Estimates							
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq	
Intercept		1	-2.4486	0.7671	10.1885	0.0014	
ВМІ		1	0.0791	0.0293	7.3009	0.0069	
BPMEDS	1	1	7.4604	2.7204	7.5207	0.0061	
intBMI_BPMEDS		1	-0.2158	0.0984	4.8141	0.0282	

Odds Ratio Estimates						
Effect	Point Estimate	95% Wald Confidence Limits				
BMI	1.082	1.022	1.146			
BPMEDS 1 vs 0	>999,999	8.402	>999.999			
intBMI_BPMEDS	0.806	0.665	0.977			

4.

Analysis of Maximum Likelihood Estimates							
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq	
Intercept		1	-8.5994	1.2702	45.8354	<.0001	
AGE		1	0.0884	0.0129	47.0424	<.0001	
SEX	2	1	-0.1323	0.2070	0.4085	0.5227	
PREVSTRK	1	1	1.9337	1.0817	3.1959	0.0738	
ВМІ		1	0.1124	0.0324	12.0083	0.0005	
BPMEDS	1	1	8.1861	2.6862	9.2868	0.0023	
intBMI_BPMEDS		1	-0.2479	0.0974	6.4735	0.0109	

Odds Ratio Estimates						
Effect	Point Estimate	95% Wald Confidence Limits				
AGE	1.092	1.065	1.120			
SEX 2 vs 1	0.876	0.584	1.314			
PREVSTRK 1 vs 0	6.915	0.830	57.610			
ВМІ	1.119	1.050	1.192			
BPMEDS 1 vs 0	>999.999	18.563	>999.999			
intBMI_BPMEDS	0.780	0.645	0.945			

5.

Variable	Unadjusted OR (95% CI)	Adjusted OR (95% CI)
AGE	1.092 (1.066-1.118)	1.092 (1.066-1.121)
SEX	0.929 (0.650-1.327)	0.876 (0.583-1.314)
PREVSTRK	17.332 (2.262-132.832)	6.923 (0.583-1.314)
BMI (BPMEDS=No)	1.082 (1.022-1.470)	1.116 (1.048-1.190)
BMI (BPMEDS=Yes)	0.872 (0.719-1.044)	0.866 (0.713-1.037)

(The following table uses values obtained from the models from question 2 only)

Variable	Unadjusted OR (95% CI)
BMI (single variable model)	1.077 (1.023-1.136)
BPMEDS (single variable model)	5.350 (3.084-9.746)
Obese vs. Normal	1.954 (1.117-3.344)
Overweight vs. Normal	1.703 (1.160-2.509)

Joe Brew - SAS and R code

PHC 6053 – Assignment 8

```
Question
        SAS
                                                   R
        *READ IN THE DATA FROM ASSIGNMENT 1;
                                                   setwd("C:/Users/BrewJR/Desktop/ass8")
Prep
        LIBNAME ass8
        'C:\Users\BrewJR\Desktop\ass8';
                                                   library(sas7bdat)
        DATA ass8.mydata;
                                                   library(foreign)
        SET ass8.fghm122;
        RUN;
                                                   dat <- read.ssd(</pre>
                                                     "C:/Users/BrewJR/Desktop/ass8",
        *ASSIGN THE DATA TO &DAT;
                                                     "fghm122",
        %let dat=ass8.mydata;
                                                     sascmd="C:/Program
        RUN;
                                                   Files/SASHome93/SASFoundation/9.3/sas.exe")
         /* TAKE A LOOK AT THE DATA */
        proc print data=&dat (obs=7);
        run;
         *OPTIONAL: OUTPUT DIRECTLY AS A PDF
        FILE/;
        ods pdf file =
        "C:\Users\BrewJR\Desktop\ass8.pdf"
```

```
/* 1(a) CREATE A BMIGROUP VARIABLE*/
data &dat;
                                             # 1(a) CREATE A BMIGROUP VARIABLE*/
set &dat;
                                           dat$BMIGROUP <- ifelse(dat$BMI<18.5,</pre>
BMIGROUP=.;
                                                                   1,
if BMI<18.5 then BMIGROUP=1;</pre>
                                                                   ifelse(dat$BMI >= 18.5
if BMI >= 18.5 & BMI < 25 then BMIGROUP | & dat$BMI < 25,
if BMI >= 25 & BMI < 30 then BMIGROUP =
                                                                           ifelse(dat$BMI
3;
                                           >= 25 & dat$BMI < 30,
if BMI >= 30 then BMIGROUP = 4;
                                                                                  3,
RUN;
                                           ifelse(dat$BMI >= 30,
/* 1(b) CREATE HBP */
                                                                                          4,
data &dat;
set &dat;
                                           NA))))
HBP = .;
if SYSBP >= 0 & SYSBP < 140 then HBP =
                                           # 1(b) CREATE HBP */
0;
                                           dat$HBP <- ifelse(dat$SYSBP >=0 & dat$SYSBP <</pre>
if SYSBP >= 140 then HBP = 1;
                                           140,
RUN;
                                                              Ο,
                                                              ifelse(dat$SYSBP >= 140,
/* 1(c) Remove all the individuals in
                                                                     1,
the underweight BMI group AND all the
                                                                     NA))
individuals with a BMI which is 40 or
larger*/
data dat2;
                                           # 1(c) Remove all the individuals in the
set &dat;
                                           underweight BMI group AND
if BMIGROUP = 1 then delete;
                                           #all the individuals with a BMI which is 40 or
if BMI > 40 then delete;
                                           larger*/
run;
                                           dat2 <-dat[which(dat$BMIGROUP != 1 &</pre>
                                                               dat$BMI <=40),]</pre>
```

```
/* AGE */
                                              # AGE */
proc logistic data = dat2;
                                           modelAGE <- glm(dat2$HBP ~ dat2$AGE,</pre>
model HBP(event="1") = AGE;
                                           family="binomial")
run;
                                           exp(cbind(OR = coef(modelAGE),
                                           confint(modelAGE)))
/* BMI */
proc logistic data = dat2;
                                           # BMI */
                                           modelBMI \leftarrow glm(dat2$HBP \sim dat2$BMI,
model HBP(event="1") = BMI;
run;
                                           family="binomial")
                                           exp(cbind(OR = coef(modelBMI),
/* BMIGROUP */
                                           confint(modelBMI)))
proc logistic data = dat2;
class BMIGROUP (ref="2") / param=ref;
                                           # BMIGROUP */
model HBP(event="1") = BMIGROUP;
                                           dat2$BMIGROUP <- factor(dat2$BMIGROUP,</pre>
run;
                                           levels=c(2,3,4)
                                           modelBMIGROUP <- glm(dat2$HBP ~ dat2$BMIGROUP,</pre>
/* SEX */
                                           family="binomial")
proc logistic data = dat2;
                                           exp(cbind(OR = coef(modelBMIGROUP),
model HBP(event="1") = SEX;
                                           confint(modelBMIGROUP)))
run;
                                           # SEX */
/* BPMEDS */
                                           modelSEX <- glm(dat2$HBP ~ dat2$SEX,</pre>
proc logistic data = dat2;
                                           family="binomial")
model HBP(event="1") = BPMEDS;
                                           exp(cbind(OR = coef(modelSEX),
run;
                                           confint(modelSEX)))
/* PREVSTRK */
                                           # BPMEDS */
proc logistic data = dat2;
                                           modelBPMEDS <- glm(dat2$HBP ~ dat2$BPMEDS,</pre>
model HBP(event="1") = PREVSTRK;
                                           family="binomial")
run;
                                           exp(cbind(OR = coef(modelBPMEDS),
                                           confint(modelBPMEDS)))
                                           # PREVSTRK */
                                           modelPREVSTRK <- glm(dat2$HBP ~ dat2$PREVSTRK,</pre>
                                           family="binomial")
                                           exp(cbind(OR = coef(modelPREVSTRK),
                                           confint(modelPREVSTRK)))
```

```
3
         /* FIRST, CREATE AN INTERACTION TERM
                                                   dat2$intBMI_BPMEDS <- dat2$BMI*dat2$BPMEDS</pre>
        BETWEEN BMI AND BPMEDS */
        data dat2;
                                                   #NOW RUN THE MODEL */
                                                   model3 <- glm(dat2$HBP ~</pre>
         set dat2;
         intBMI BPMEDS = BMI * BPMEDS;
                                                                    dat2$BMI +
        run;
                                                                    dat2$BPMEDS +
                                                                    dat2$intBMI_BPMEDS,
         /*NOW RUN THE MODEL */
                                                   family="binomial")
                                                   exp(cbind(OR = coef(model3), confint(model3)))
        proc logistic data = dat2;
        class BPMEDS(ref="0") / param=ref;
        model HBP(event="1") = BMI BPMEDS
        intBMI BPMEDS;
        run;
        proc logistic data = dat2;
                                                   model4 <- glm(dat2$HBP ~</pre>
4
        class BPMEDS(ref="0") / param=ref;
                                                                    dat2$AGE +
        class SEX(ref="1") / param=ref;
                                                                    dat2$SEX +
        class PREVSTRK(ref="0") / param=ref;
                                                                    dat2$PREVSTRK +
        model HBP(event="1") = AGE SEX PREVSTRK
                                                                    dat2$BMI +
        BMI BPMEDS intBMI_BPMEDS;
                                                                    dat2$BPMEDS +
                                                                    dat2$intBMI_BPMEDS,
        run;
        ods pdf close;
                                                   family="binomial")
                                                   exp(cbind(OR = coef(model4), confint(model4)))
```

```
#Subset dat2 into bpmeds=yes and bpmeds=no
5
                                                    dat2y <- dat2[which(dat2$BPMEDS == 1),]</pre>
                                                    dat2n <- dat2[which(dat2$BPMEDS == 0),]</pre>
                                                    #GET UNADJSUTED FOR BOTH
                                                    #BPMEDS=Yes
                                                    modelBMI_BPMEDSY <- glm(dat2y$HBP ~ dat2y$BMI,</pre>
                                                    family="binomial")
                                                    exp(cbind(OR = coef(modelBMI_BPMEDSY),
                                                    confint(modelBMI_BPMEDSY)))
                                                    #BPMEDS-No
                                                    modelBMI_BPMEDSN <- glm(dat2n$HBP ~ dat2n$BMI,</pre>
                                                    family="binomial")
                                                    exp(cbind(OR = coef(modelBMI_BPMEDSN),
                                                    confint(modelBMI_BPMEDSN)))
                                                    #GET ADJUSTED FOR BOTH
                                                    #BPMEDS=YES
                                                    modelBMI_BPMEDSYadj <- glm(dat2y$HBP ~</pre>
                                                                                   dat2y$AGE +
                                                                                   dat2y$SEX +
                                                                                   dat2y$PREVSTRK +
                                                                                   dat2y$BMI,
                                                                                 family="binomial")
                                                    exp(cbind(OR = coef(modelBMI_BPMEDSYadj),
                                                    confint(modelBMI_BPMEDSYadj)))
                                                    #BPMEDS=NO
                                                    modelBMI_BPMEDSNadj <- glm(dat2n$HBP ~</pre>
                                                                                   dat2n$AGE +
                                                                                   dat2n$SEX +
                                                                                   dat2n$PREVSTRK +
                                                                                   dat2n$BMI,
                                                                                 family="binomial")
                                                    exp(cbind(OR = coef(modelBMI_BPMEDSNadj),
                                                    confint(modelBMI_BPMEDSNadj)))
```

Obs	SEX	RANDID	TOTCHOL	AGE	SYSBP	DIABP	CURSMOKE	CIGPDAY	BMI
1	2	6238	237	58	108	66	0	0	28.5
2	1	14367	280	64	168	100	0	0	25.72
3	1	16365	211	55	173	123	0	0	29.11
4	1	82425	226	67	157	95	0	0	29.86
5	1	101990	230	49	142	90.5	1	35	24.33
6	2	123622	241	51	145	85	1	20	25.66
7	1	147250	229	68	145	77	1	20	23.09

Obs	DIABETES	BPMEDS	HEARTRTE	GLUCOSE	PREVCHD	PREVAP	PREVMI
1	0	0	80	71	0	0	0
2	0	0	92	82	0	0	0
3	0	1	75	85	0	0	0
4	0	0	88	99	0	0	0
5	0	0	70	61	0	0	0
6	0	0	96	102	0	0	0
7	0	0	72	83	0	0	0

Obs	PREVSTRK	PREVHYP	PERIOD	HDLC	LDLC	BMIGROUP	НВР
1	0	0	3	54	141	3	0
2	0	1	3	44	236	3	1
3	0	1	3	48	163	3	1
4	0	1	3	61	165	3	1
5	0	1	3	30	200	2	1
6	0	0	3	68	173	3	1
7	0	0	3	39	170	2	1

Model Information				
Data Set	WORK.DAT2			
Response Variable	НВР			
Number of Response Levels	2			
Model	binary logit			
Optimization Technique	Fisher's scoring			

Number of Observations Read	494
Number of Observations Used	494

Response Profile			
Ordered Value	НВР	Total Frequency	
1	0	266	
2	1	228	

Probability modeled is HBP=1.

Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics			
Criterion	Intercept Only	Intercept and Covariates	
AIC	683.903	625.023	
SC	688.106	633.428	
-2 Log L	681.903	621.023	

Testing Global Null Hypothesis: BETA=0				
Test	Chi-Square	DF	Pr > ChiSq	
Likelihood Ratio	60.8803	1	<.0001	
Score	58.7136	1	<.0001	
Wald	54.0266	1	<.0001	

Analysis of Maximum Likelihood Estimates					
Parameter	DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept	1	-5.4506	0.7277	56.0984	<.0001
AGE	1	0.0877	0.0119	54.0266	<.0001

Odds Ratio Estimates			
Point 95% Wald Confidence Limits			
AGE	1.092	1.066	1.118

Association of Predicted Probabilities and Observed Responses			
Percent Concordant	67.9	Somers' D	0.390
Percent Discordant	29.0	Gamma	0.402
Percent Tied	3.1	Tau-a	0.194
Pairs	60648	С	0.695

Model Information		
Data Set WORK.DAT		
Response Variable	НВР	
Number of Response Levels	2	
Model	binary logit	
Optimization Technique	Fisher's scoring	

Number of Observations Read	494
Number of Observations Used	494

Response Profile			
Ordered Value	НВР	Total Frequency	
1	0	266	
2	1	228	

Probability modeled is HBP=1.

Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics			
Criterion	Intercept Only	Intercept and Covariates	
AIC	683.903	678.029	
SC	688.106	686.434	
-2 Log L	681.903	674.029	

Testing Global Null Hypothesis: BETA=0						
Test Chi-Square DF Pr > ChiSq						
Likelihood Ratio	7.8744	1	0.0050			
Score	7.8323	1	0.0051			
Wald	7.7077	1	0.0055			

Analysis of Maximum Likelihood Estimates							
Parameter DF Estimate Standard Chi-Square Pr > C							
Intercept	1	-2.0949	0.7054	8.8191	0.0030		
BMI	1	0.0745	0.0268	7.7077	0.0055		

Odds Ratio Estimates				
Point 95% Wald Confidence Limits				
BMI	1.077	1.022	1.136	

Association of Predicted Probabilities and Observed Responses						
Percent Concordant 56.7 Somers' D 0.141						
Percent Discordant	42.5	Gamma	0.143			
Percent Tied 0.8 Tau-a						
Pairs	60648	С	0.571			

Model Information				
Data Set	WORK.DAT2			
Response Variable	НВР			
Number of Response Levels	2			
Model	binary logit			
Optimization Technique	Fisher's scoring			

Number of Observations Read	494
Number of Observations Used	494

Response Profile				
Ordered Value	Total Frequency			
1	0	266		
2	1	228		

Probability modeled is HBP=1.

Class Level Information					
Class	Class Value Design Variables				
BMIGROUP	2	0 0			
	3	1	0		
	4	0	1		

Model Convergence Status

Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics						
Criterion	Intercept Only	Intercept and Covariates				
AIC	683.903	678.345				
SC	688.106	690.952				
-2 Log L	681.903	672.345				

Testing Global Null Hypothesis: BETA=0						
Test Chi-Square DF Pr > ChiSo						
Likelihood Ratio	9.5587	2	0.0084			
Score	9.5029	2	0.0086			
Wald	9.4260	2	0.0090			

Type 3 Analysis of Effects					
Effect DF Chi-Square Pr > ChiSq					
BMIGROUP	2	9.4260	0.0090		

Analysis of Maximum Likelihood Estimates						
Parameter DF Estimate Standard Chi-Square Pr > 0				Pr > ChiSq		
Intercept		1	-0.4875	0.1439	11.4827	0.0007
BMIGROUP	3	1	0.5323	0.1966	7.3334	0.0068
BMIGROUP	4	1	0.6698	0.2860	5.4840	0.0192

Odds Ratio Estimates			
Point 95% Wald Estimate Confidence Limits			
BMIGROUP 3 vs 2	1.703	1.158	2.503
BMIGROUP 4 vs 2	1.954	1.115	3.423

Association of Predicted Probabilities and Observed Responses				
Percent Concordant 37.9 Somers' D 0.144				
Percent Discordant 23.5 Gamma		0.234		
Percent Tied 38.6 Tau-a 0.0				
Pairs	60648	С	0.572	

Model Information		
1710del Illiol Illatio	, ii	
Data Set	WORK.DAT2	
Response Variable	НВР	
Number of Response Levels	2	
Model	binary logit	
Optimization Technique	Fisher's scoring	

Number of Observations Read	494
Number of Observations Used	494

Response Profile			
Ordered Tota Value HBP Frequency			
1	0	266	
2	1	228	

Probability modeled is HBP=1.

Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics			
Criterion	Intercept Only	Intercept and Covariates	
AIC	683.903	685.738	
SC	688.106	694.143	
-2 Log L	681.903	681.738	

Testing Global Null Hypothesis: BETA=0				
Test Chi-Square DF Pr > ChiSq				
Likelihood Ratio	0.1651	1	0.6845	
Score	0.1651	1	0.6845	
Wald	0.1651	1	0.6845	

Analysis of Maximum Likelihood Estimates					
Parameter DF Estimate Standard Chi-Square Pr > Ch					Pr > ChiSq
Intercept	1	-0.0382	0.2990	0.0163	0.8983
SEX	1	-0.0740	0.1821	0.1651	0.6845

Odds Ratio Estimates			
Point 95% Wald Confidence Limits			
SEX	0.929	0.650	1.327

Association of Predicted Probabilities and Observed Responses					
Percent Concordant25.5Somers' D0.018					
Percent Discordant	23.7	Gamma	0.037		
Percent Tied 50.9 Tau-a 0.009					
Pairs 60648 c 0.50					

Model Information			
Data Set	WORK.DAT2		
Response Variable	НВР		
Number of Response Levels	2		
Model	binary logit		
Optimization Technique	Fisher's scoring		

Number of Observations Read	494
Number of Observations Used	494

Response Profile		
Ordered Value	НВР	Total Frequency
1	0	266
2	1	228

Probability modeled is HBP=1.

Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics		
Criterion	Intercept Only	Intercept and Covariates
AIC	683.903	646.219
SC	688.106	654.624
-2 Log L	681.903	642.219

Testing Global Null Hypothesis: BETA=0			
Test	Chi-Square	DF	Pr > ChiSq
Likelihood Ratio	39.6841	1	<.0001
Score	38.2875	1	<.0001
Wald	33.0049	1	<.0001

Analysis of Maximum Likelihood Estimates					
Parameter	DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq
Intercept	1	-0.3995	0.1000	15.9502	<.0001
BPMEDS	1	1.6771	0.2919	33.0049	<.0001

Odds Ratio Estimates			
Effect	Point Estimate	95% Confiden	Wald ice Limits
BPMEDS	5.350	3.019	9.481

Association of Predicted Probabilities and Observed Responses			
Percent Concordant	25.0	Somers' D	0.204
Percent Discordant	4.7	Gamma	0.685
Percent Tied	70.3	Tau-a	0.101
Pairs	60648	С	0.602

Model Information		
Data Set	WORK.DAT2	
Response Variable	НВР	
Number of Response Levels	2	
Model	binary logit	
Optimization Technique	Fisher's scoring	

Number of Observations Read	494
Number of Observations Used	494

Response Profile		
Ordered Value	НВР	Total Frequency
1	0	266
2	1	228

Probability modeled is HBP=1.

Model Convergence Status
Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics			
Criterion	Intercept Only	Intercept and Covariates	
AIC	683.903	669.943	
SC	688.106	678.348	
-2 Log L	681.903	665.943	

Testing Global Null Hypothesis: BETA=0				
Test	Chi-Square	DF	Pr > ChiSq	
Likelihood Ratio	15.9609	1	<.0001	
Score	13.8556	1	0.0002	
Wald	7.5368	1	0.0060	

Analysis of Maximum Likelihood Estimates					
Parameter DF Estimate Standard Chi-Square					Pr > ChiSq
Intercept	1	-0.2138	0.0919	5.4094	0.0200
PREVSTRK	1	2.8526	1.0391	7.5368	0.0060

Odds Ratio Estimates			
Effect	Point 95% Wald Estimate Confidence Limits		
PREVSTRK	17.332	2.262	132.832

Association of Predicted Probabilities and Observed Responses					
Percent Concordant	6.1	Somers' D	0.058		
Percent Discordant 0.4 Gamma 0.891					
Percent Tied	93.5 Tau-a 0.029				
Pairs	irs 60648 c 0.529				

Model Information			
Data Set	WORK.DAT2		
Response Variable	НВР		
Number of Response Levels	2		
Model	binary logit		
Optimization Technique	Fisher's scoring		

Number of Observations Read	494
Number of Observations Used	494

Response Profile			
Ordered Value	Total Frequency		
1	0	266	
2	1	228	

Probability modeled is HBP=1.

Class Level Information				
Class Value Design Variables				
BPMEDS	0	0		
	1	1		

Model Convergence Status Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics			
Criterion	Intercept Only	Intercept and Covariates	
AIC	683.903	640.563	
SC	688.106	657.373	
-2 Log L	681.903	632.563	

Testing Global Null Hypothesis: BETA=0				
Test	Chi-Square	DF	Pr > ChiSq	
Likelihood Ratio	49.3403	3	<.0001	
Score	46.9814	3	<.0001	
Wald	39.6163	3	<.0001	

Type 3 Analysis of Effects					
Effect DF Chi-Square Pr > ChiSq					
BMI	1	7.3009	0.0069		
BPMEDS	1	7.5207	0.0061		
intBMI_BPMEDS	1	4.8141	0.0282		

Analysis of Maximum Likelihood Estimates							
Parameter		DF Estimate Standard Chi-Square Pr >					
Intercept		1	-2.4486	0.7671	10.1885	0.0014	
BMI		1	0.0791	0.0293	7.3009	0.0069	
BPMEDS	1	1	7.4604	2.7204	7.5207	0.0061	
intBMI_BPMEDS		1	-0.2158	0.0984	4.8141	0.0282	

Odds Ratio Estimates							
Effect	Point 95% Wald Confidence Limits						
BMI	1.082	1.022 1.146					
BPMEDS 1 vs 0	>999.999	8.402 >999.999					
intBMI_BPMEDS	0.806	0.665	0.977				

Association of Predicted Probabilities and Observed Responses					
Percent Concordant	65.3	Somers' D	0.311		
Percent Discordant	34.2	Gamma	0.313		
Percent Tied	0.6	Tau-a	0.155		
Pairs	60648	С	0.655		

Model Information				
Data Set	WORK.DAT2			
Response Variable	НВР			
Number of Response Levels	2			
Model	binary logit			
Optimization Technique	Fisher's scoring			

Number of Observations Read	494
Number of Observations Used	494

Response Profile					
Ordered Value	НВР	Total Frequency			
1	0	266			
2	1	228			

Probability modeled is HBP=1.

Class Level Information				
Class	Value	Design Variables		
BPMEDS	0	0		
	1	1		
SEX	1	0		
	2	1		
PREVSTRK	0	0		
	1	1		

Model Convergence Status Convergence criterion (GCONV=1E-8) satisfied.

Model Fit Statistics					
Criterion	Intercept Only	Intercept and Covariates			
AIC	683.903	582.454			
SC	688.106	611.872			
-2 Log L	681.903	568.454			

Testing Global Null Hypothesis: BETA=0							
Test	Chi-Square DF Pr > Chi						
Likelihood Ratio	113.4495	6	<.0001				
Score	100.9107	6	<.0001				
Wald	79.2805	6	<.0001				

Type 3 Analysis of Effects							
Effect	DF	Wald Chi-Square	Pr > ChiSq				
AGE	1	47.0424	<.0001				
SEX	1	0.4085	0.5227				
PREVSTRK	1	3.1959	0.0738				
BMI	1	12.0083	0.0005				
BPMEDS	1	9.2868	0.0023				
intBMI_BPMEDS	1	6.4735	0.0109				

Analysis of Maximum Likelihood Estimates							
Parameter		DF	Estimate	Standard Error	Wald Chi-Square	Pr > ChiSq	
Intercept		1	-8.5994	1.2702	45.8354	<.0001	
AGE		1	0.0884	0.0129	47.0424	<.0001	
SEX	2	1	-0.1323	0.2070	0.4085	0.5227	
PREVSTRK	1	1	1.9337	1.0817	3.1959	0.0738	
BMI		1	0.1124	0.0324	12.0083	0.0005	
BPMEDS	1	1	8.1861	2.6862	9.2868	0.0023	
intBMI_BPMEDS		1	-0.2479	0.0974	6.4735	0.0109	

Odds Ratio Estimates							
Effect	Point Estimate						
AGE	1.092	1.065	1.120				
SEX 2 vs 1	0.876	0.584	1.314				
PREVSTRK 1 vs 0	6.915	0.830	57.610				
BMI	1.119	1.050	1.192				
BPMEDS 1 vs 0	>999.999	18.563	>999.999				
intBMI_BPMEDS	0.780	0.645	0.945				

Association of Predicted Probabilities and Observed Responses			
Percent Concordant	75.2	Somers' D	0.506
Percent Discordant	24.6	Gamma	0.507
Percent Tied	0.2	Tau-a	0.252
Pairs	60648	С	0.753