

Solutions for Session 7

22/11/2022

```
. do solution.do
. global basedir http://personalpages.manchester.ac.uk/staff/mark.lunt
. global datadir $basedir/stats/7_binary/data
. use $datadir/epicourse, clear
. tab hip_p sex, co
```

Key
<i>frequency</i> <i>column percentage</i>

hip pain	sex		Total
	M	F	
no	1,969 90.16	1,976 84.77	3,945 87.38
yes	215 9.84	355 15.23	570 12.62
Total	2,184 100.00	2,331 100.00	4,515 100.00

1.1 Prevalence is 9.84% in men, 15.23% in women

. tab hip_p sex, co chi2

Key
frequency
column percentage

hip pain	sex		Total
	M	F	
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yes	215 9.84	355 15.23	570 12.62
Total	2,184 100.00	2,331 100.00	4,515 100.00

Pearson chi2(1) = 29.6438 Pr = 0.000

1.2 The difference in prevalence between men and women is very significant

. cs hip_p sex, or

	sex		Total
	Exposed	Unexposed	
Cases	355	215	570
Noncases	1976	1969	3945
Total	2331	2184	4515
Risk	.1522952	.0984432	.1262458
	Point estimate		[95% Conf. Interval]
Risk difference	.0538519		.0346461 .0730578
Risk ratio	1.547035		1.319614 1.81365
Attr. frac. ex.	.3536024		.2422027 .4486258
Attr. frac. pop	.220226		
Odds ratio	1.645314		1.373815 1.970458 (Cornfield)

chi2(1) = 29.64 Pr>chi2 = 0.0000

1.3 Confidence interval is (1.37, 1.97)

1.4 The odds ratio and the relative risk are very similar

1.5 Yes, the confidence interval does not contain 0, which is the null hypothesis risk difference

```

. logistic hip_p sex
Logistic regression
Log likelihood = -1697.0482
Number of obs = 4515
LR chi2(1) = 29.97
Prob > chi2 = 0.0000
Pseudo R2 = 0.0088

```

hip_p	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
sex	1.645313	.1515298	5.41	0.000	1.373583 1.9708
_cons	.1091925	.0078429	-30.83	0.000	.0948536 .1256989

1.6 The odds ratio is exactly the same as that produced by cs
1.7 The confidence intervals are the same to 3 decimal places (the methods used to calculate them differ, but generally give very similar results)

```

. egen agegp = cut(age), at(0 30(10)100)
. label define age 0 "<30" 30 "30-39" 40 "40-49" 50 "50-59"
. label define age 60 "60-69" 70 "70-79" 80 "80-89" 90 "90+", modify
. label values agegp age
. tab agegp hip_p, chi2

```

agegp	hip pain		Total
	no	yes	
<30	388	9	397
30-39	358	10	368
40-49	498	47	545
50-59	510	90	600
60-69	741	150	891
70-79	987	177	1,164
80-89	415	77	492
90+	48	10	58
Total	3,945	570	4,515

Pearson chi2(7) = 108.8887 Pr = 0.000

2.1 Yes: chi2 is very significant

. logistic hip_p age sex

Logistic regression

Number of obs = 4515
 LR chi2(2) = 125.12
 Prob > chi2 = 0.0000
 Pseudo R2 = 0.0365

Log likelihood = -1649.4709

hip_p	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
age	1.026149	.0028811	9.19	0.000	1.020517 1.031811
sex	1.700317	.1582258	5.70	0.000	1.416837 2.040516
_cons	.0211183	.0042313	-19.25	0.000	.0142597 .0312757

2.2 Yes: $p = 0.000$

2.3 Odds of hip pain increase by 1.03 for each year increase in age

. logistic hip_p i.sex#c.age

Logistic regression

Number of obs = 4515
 LR chi2(3) = 127.54
 Prob > chi2 = 0.0000
 Pseudo R2 = 0.0372

Log likelihood = -1648.26

hip_p	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
sex					
F	.9412622	.3650805	-0.16	0.876	.44011 2.013075
age	1.020415	.0045672	4.52	0.000	1.011503 1.029406
sex#c.age					
F	1.009026	.0057943	1.56	0.118	.9977332 1.020447
_cons	.030562	.0092115	-11.57	0.000	.0169288 .0551744

2.4 No: the interaction term $i.sex#c.age$ is not significant ($p=0.118$)

```
. logistic hip_p sex i.agegp
```

```
Logistic regression                               Number of obs =      4515
                                                    LR chi2(8)       =    173.58
                                                    Prob > chi2      =    0.0000
Log likelihood = -1625.2427                       Pseudo R2       =    0.0507
```

hip_p	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
sex	1.725543	.1611575	5.84	0.000	1.436904 2.072164
agegp					
30-39	1.196695	.5572489	0.39	0.700	.4804145 2.980925
40-49	4.073759	1.509333	3.79	0.000	1.970716 8.421057
50-59	7.740022	2.759065	5.74	0.000	3.848723 15.56567
60-69	9.061111	3.16519	6.31	0.000	4.569232 17.96882
70-79	7.996188	2.777288	5.99	0.000	4.047978 15.7953
80-89	8.16381	2.937001	5.84	0.000	4.033349 16.52418
90+	8.463806	4.109421	4.40	0.000	3.268007 21.9204
_cons	.0167001	.0057293	-11.93	0.000	.0085251 .0327147

2.5 Odds for a man aged 50-60 are 7.74 times the odds for a man aged less than 30

```
. logistic hip_p age sex
```

```
Logistic regression                               Number of obs =      4515
                                                    LR chi2(2)       =    125.12
                                                    Prob > chi2      =    0.0000
Log likelihood = -1649.4709                       Pseudo R2       =    0.0365
```

hip_p	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
age	1.026149	.0028811	9.19	0.000	1.020517 1.031811
sex	1.700317	.1582258	5.70	0.000	1.416837 2.040516
_cons	.0211183	.0042313	-19.25	0.000	.0142597 .0312757

```
. estat gof
```

```
Logistic model for hip_p, goodness-of-fit test
```

```
number of observations =      4515
number of covariate patterns =    162
Pearson chi2(159) =    153.90
Prob > chi2 =    0.5993
```

3.1 Yes. However, this is not really appropriate, since there are so many covariate patterns. It would be better to use only 10 groups

```

. estat gof, group(10)
Logistic model for hip_p, goodness-of-fit test
(Table collapsed on quantiles of estimated probabilities)
      number of observations =      4515
        number of groups   =         10
Hosmer-Lemeshow chi2(8) =      16.32
          Prob > chi2     =       0.0381

```

3.1 In this case, there is evidence that the predicted and observed values differ more than can be explained by random variation

```

. lroc
Logistic model for hip_p
number of observations =      4515
area under ROC curve  =      0.6368

```

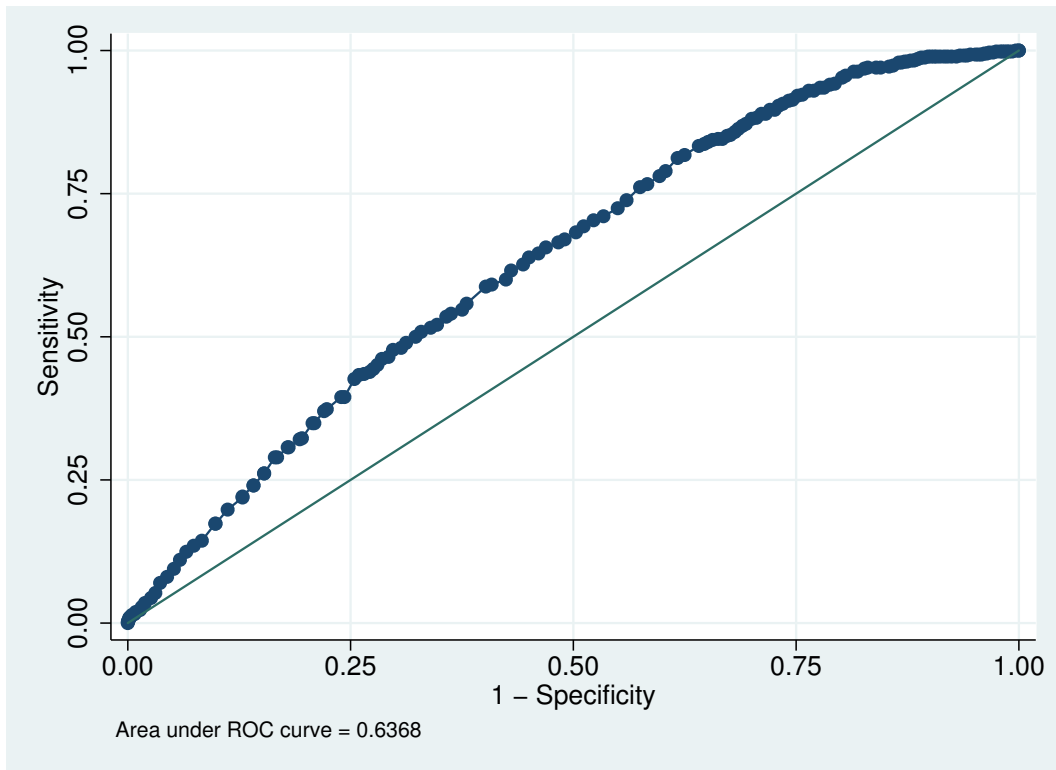


Figure 1: . lroc

```
. graph export graph1.eps replace
(file graph1.eps written in EPS format)
```

```
. logistic hip_p i.agegp sex
```

```
Logistic regression
Log likelihood = -1625.2427
Number of obs = 4515
LR chi2(8) = 173.58
Prob > chi2 = 0.0000
Pseudo R2 = 0.0507
```

hip_p	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
agegp						
30-39	1.196695	.5572489	0.39	0.700	.4804145	2.980925
40-49	4.073759	1.509333	3.79	0.000	1.970716	8.421057
50-59	7.740022	2.759065	5.74	0.000	3.848723	15.56567
60-69	9.061111	3.16519	6.31	0.000	4.569232	17.96882
70-79	7.996188	2.777288	5.99	0.000	4.047978	15.7953
80-89	8.16381	2.937001	5.84	0.000	4.033349	16.52418
90+	8.463806	4.109421	4.40	0.000	3.268007	21.9204
sex	1.725543	.1611575	5.84	0.000	1.436904	2.072164
_cons	.0167001	.0057293	-11.93	0.000	.0085251	.0327147

```
. estat gof
```

```
Logistic model for hip_p, goodness-of-fit test
```

```
number of observations = 4515
number of covariate patterns = 16
Pearson chi2(7) = 7.94
Prob > chi2 = 0.3381
```

```
. estat gof, group(10)
```

```
Logistic model for hip_p, goodness-of-fit test
```

```
(Table collapsed on quantiles of estimated probabilities)
```

```
number of observations = 4515
number of groups = 10
Hosmer-Lemeshow chi2(8) = 4.95
Prob > chi2 = 0.7633
```

3.3 Yes, this model is adequate

```
. lroc
```

```
Logistic model for hip_p
```

```
number of observations = 4515
area under ROC curve = 0.6518
```

```
. graph export graph2.eps replace
(file graph2.eps written in EPS format)
```

```
. gen age2= age*age
```

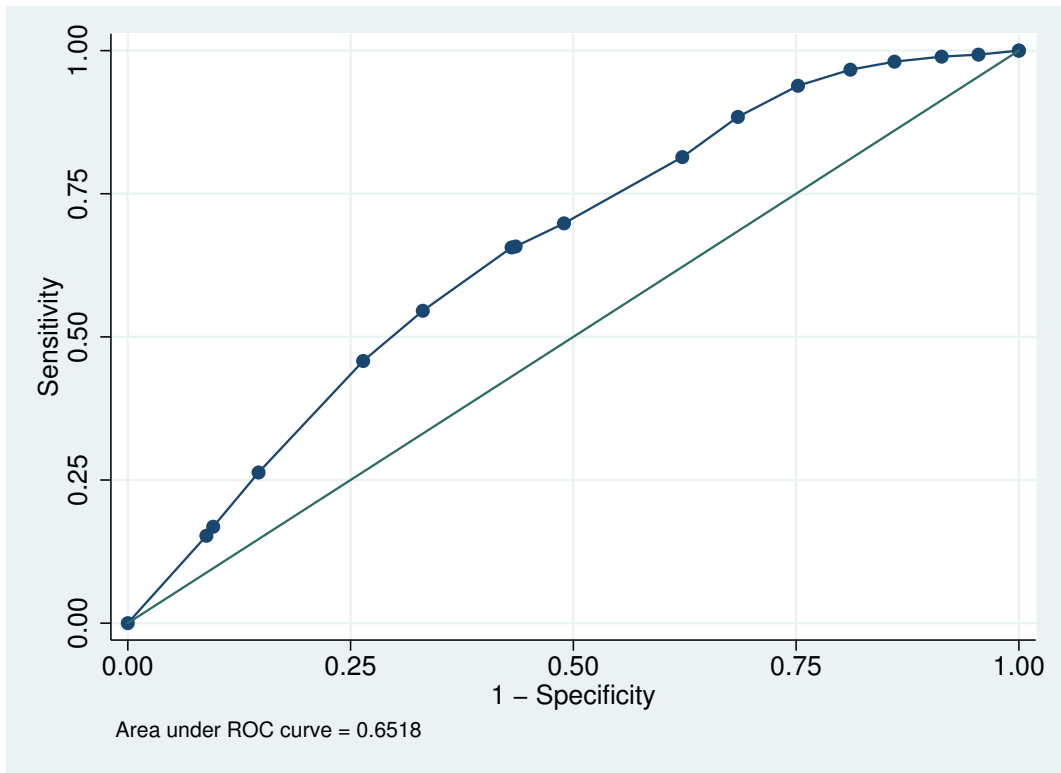


Figure 2: . lroc

```

. logistic hip_p age age2 sex
Logistic regression
Log likelihood = -1629.0242
Number of obs = 4515
LR chi2(3) = 166.01
Prob > chi2 = 0.0000
Pseudo R2 = 0.0485

```

hip_p	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
age	1.164533	.0261887	6.77	0.000	1.114319 1.21701
age2	.9989405	.0001825	-5.80	0.000	.9985828 .9992983
sex	1.73314	.1616397	5.90	0.000	1.443601 2.080751
_cons	.0006328	.0004305	-10.83	0.000	.0001668 .0024008


```
. estat gof, group(10) table
```

```
Logistic model for hip_p, goodness-of-fit test
```

```
(Table collapsed on quantiles of estimated probabilities)
```

Group	Prob	Obs_1	Exp_1	Obs_0	Exp_0	Total
1	0.0329	8	8.4	454	453.6	462
2	0.0726	18	25.1	442	434.9	460
3	0.1034	46	39.3	391	397.7	437
4	0.1230	57	56.8	435	435.2	492
5	0.1280	57	52.5	359	363.5	416
6	0.1304	59	61.2	413	410.8	472
7	0.1651	60	61.2	369	367.8	429
8	0.1964	89	84.9	372	376.1	461
9	0.2043	93	90.2	354	356.8	447
10	0.2069	83	90.5	356	348.5	439

```
number of observations = 4515  
number of groups = 10  
Hosmer-Lemeshow chi2(8) = 5.12  
Prob > chi2 = 0.7442
```

3.5 Yes, the coefficient for age2 is highly significant, and there is no longer evidence of lack of fit.

```
. lroc
```

```
Logistic model for hip_p
```

```
number of observations = 4515  
area under ROC curve = 0.6469
```

```
. graph export graph3.eps replace  
(file graph3.eps written in EPS format)
```

3.6 The area under the curve with this model is similar to that use age as a categorical predictor.

```
. predict p  
(option pr assumed; Pr(hip_p))
```

```
. predict db, dbeta
```

```
. scatter db p
```

```
. graph export graph4.eps replace  
(file graph4.eps written in EPS format)
```

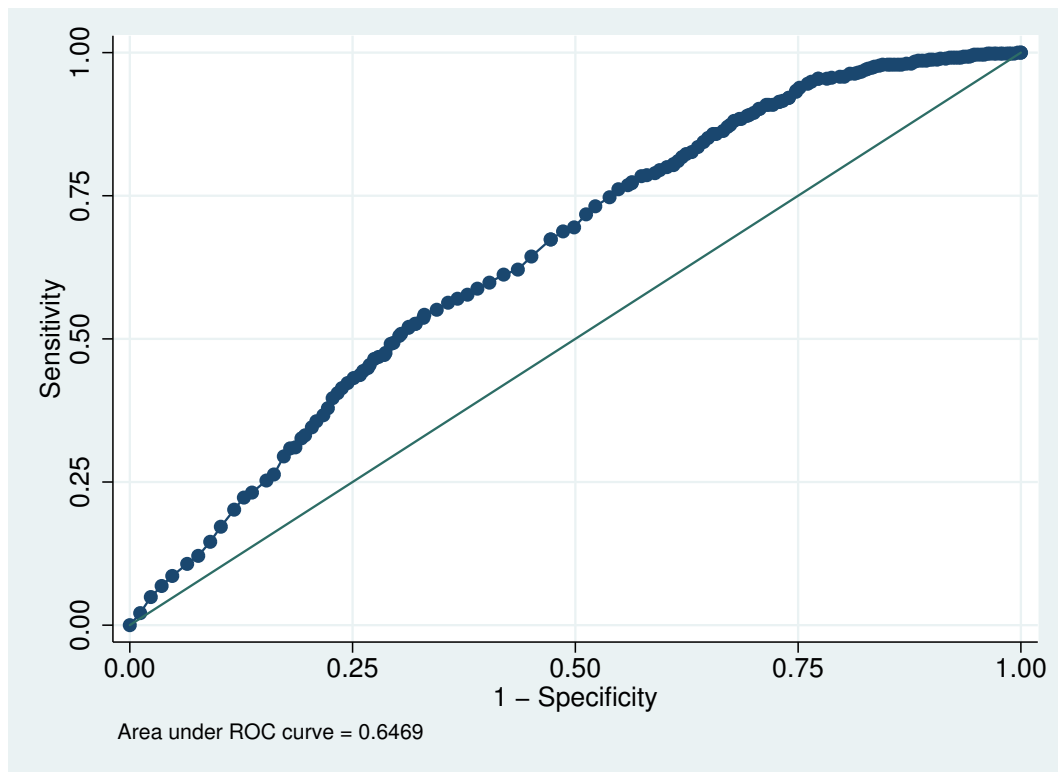


Figure 3: . lroc

*4.1 No, there are no points that are obvious outliers
However, there are 4 points that may be worth checking*

```
. predict d, ddeviance
. scatter d p

. graph export graph5.eps replace
(file graph5.eps written in EPS format)
```

4.2 Again, there is no evidence of any outliers

```
. scatter p age
```

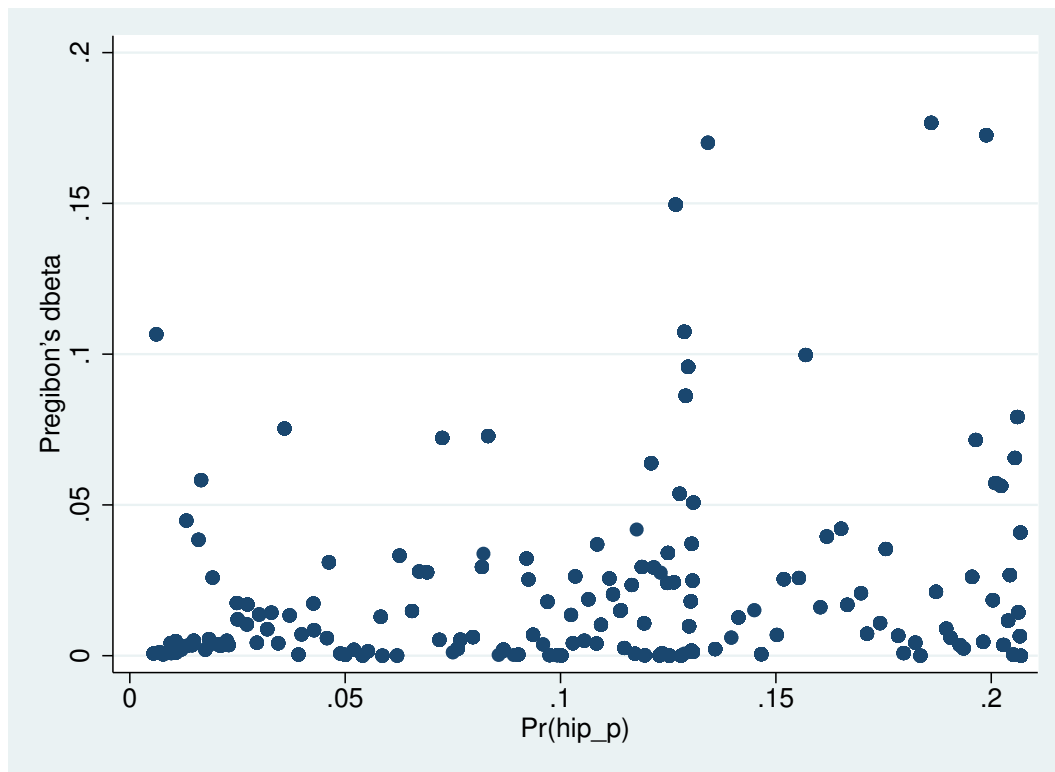


Figure 4: . scatter db p

```
. graph export graph6.eps replace
(file graph6.eps written in EPS format)
```

4.3 the two lines are the prevalences in men and women

```
. graph twoway scatter p age || lowess hip_p age if sex == 1 || lowess hip_p age if sex == 0
```

```
. graph export graph7.eps replace
(file graph7.eps written in EPS format)
```

*4.4 the fit is good for men, but fits poorly to women over 80
The quadratic model is reasonable for men, not women*

```
. use $datadir/chd, clear
```

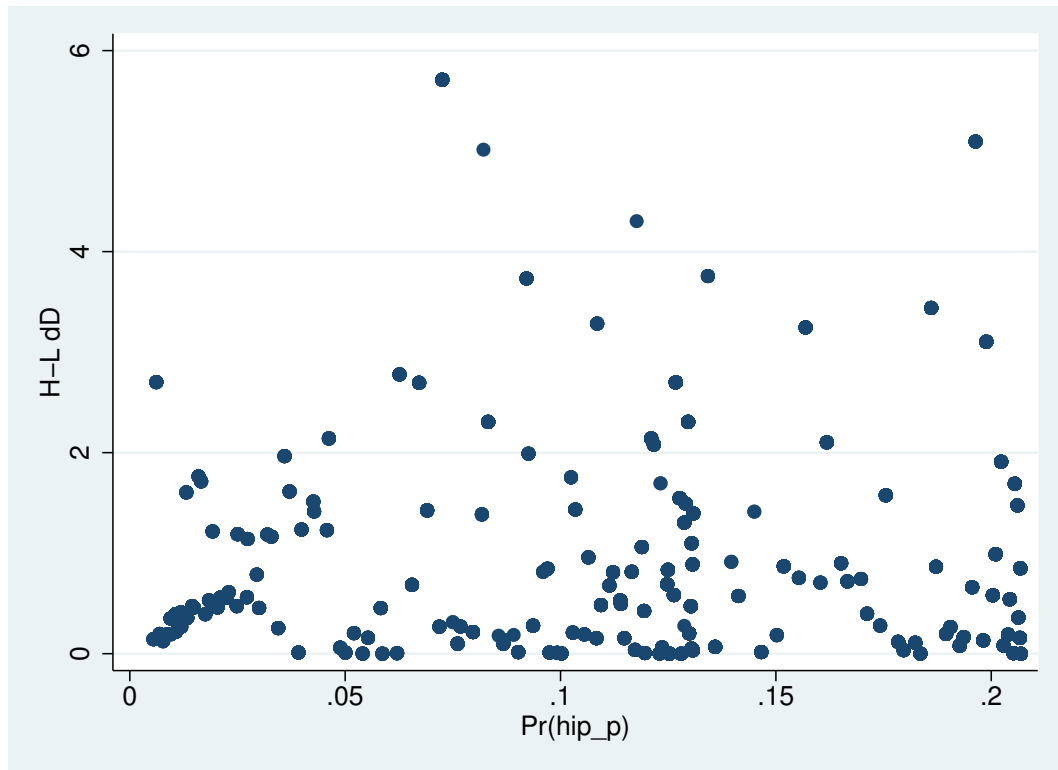


Figure 5: . scatter d p

```

. sort agegrp
. by agegrp: egen agemean = mean(age)
. by agegrp: egen chdprop = mean(chd)
. label var agemean "Mean age"
. label var chdprop "Proportion of subjects with CHD"
. scatter chdprop agemean

. graph export graph8.eps replace
(file graph8.eps written in EPS format)

```

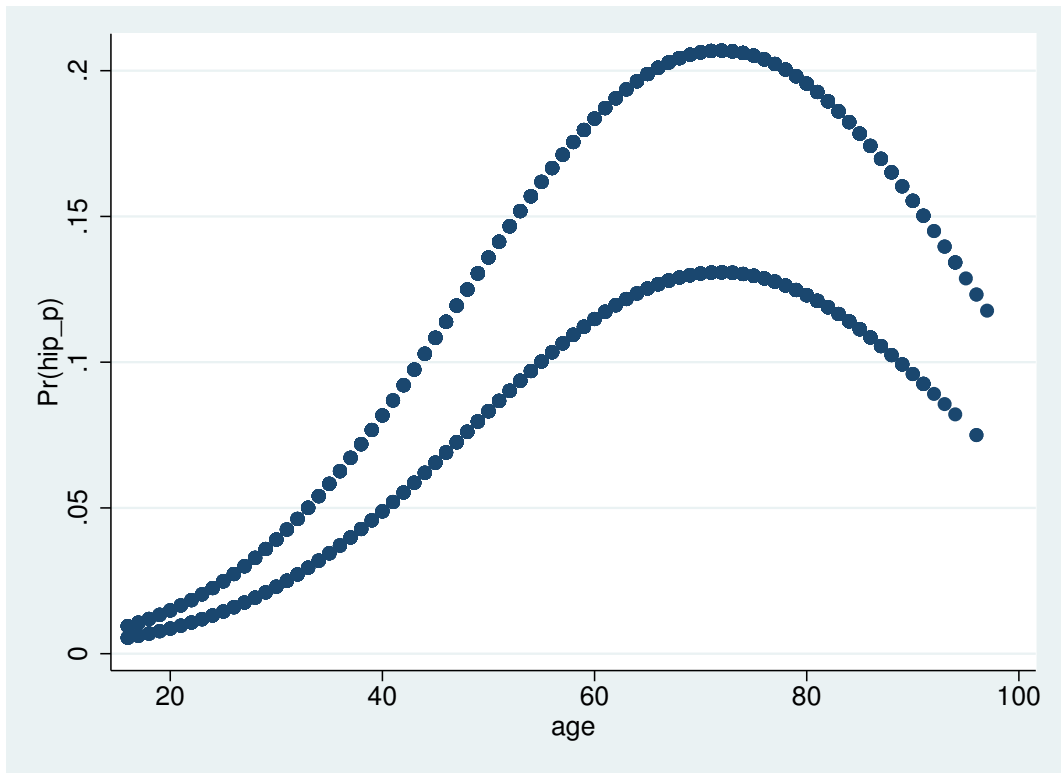


Figure 6: . scatter p age

```
. logistic chd age
Logistic regression               Number of obs   =       100
                                LR chi2(1)       =       29.31
                                Prob > chi2       =       0.0000
Log likelihood = -53.676546      Pseudo R2      =       0.2145
```

	chd	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
	age	1.117307	.0268822	4.61	0.000	1.065842 1.171257
	_cons	.0049446	.0056055	-4.68	0.000	.000536 .0456144

5.1 Odds ratio is about 1.12 per year

```
. predict p
(option pr assumed; Pr(chd))

. predict db, dbeta
```

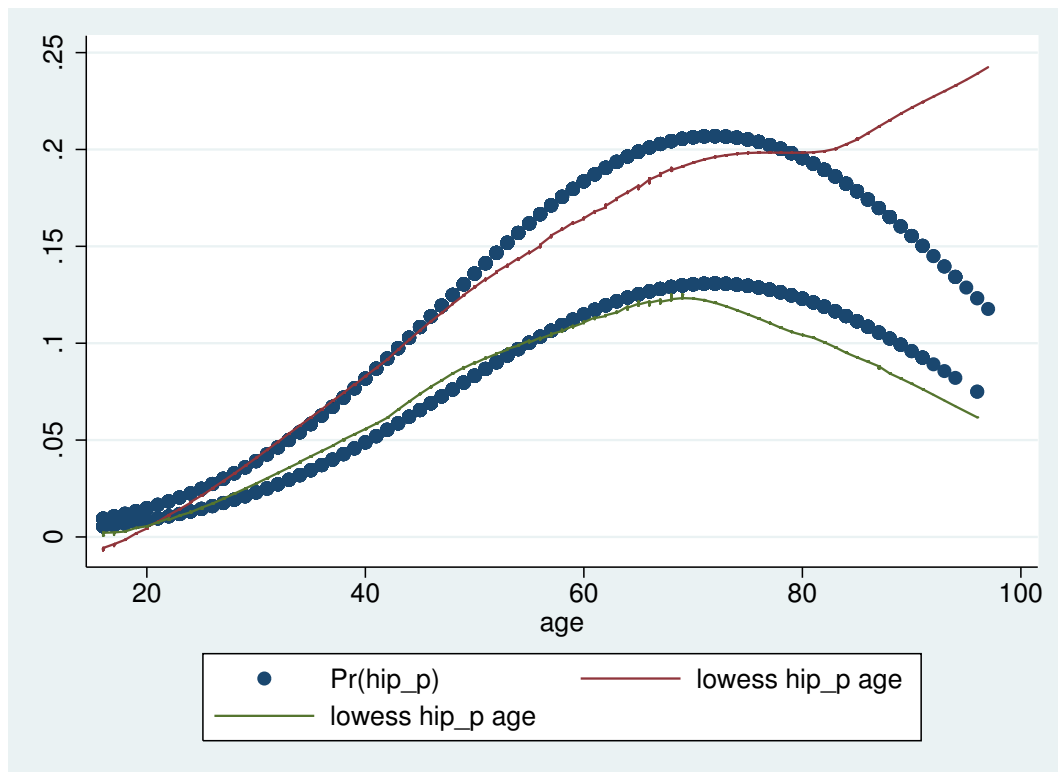


Figure 7: . graph twoway scatter p age — lowess hip'p age if sex == 1 — lowess hip'p age if sex == 0

```
. predict d, ddeviance
. scatter db p

. graph export graph9.eps replace
(file graph9.eps written in EPS format)
```

5.3 Yes, there is one influential point, with $db \sim 0.25$

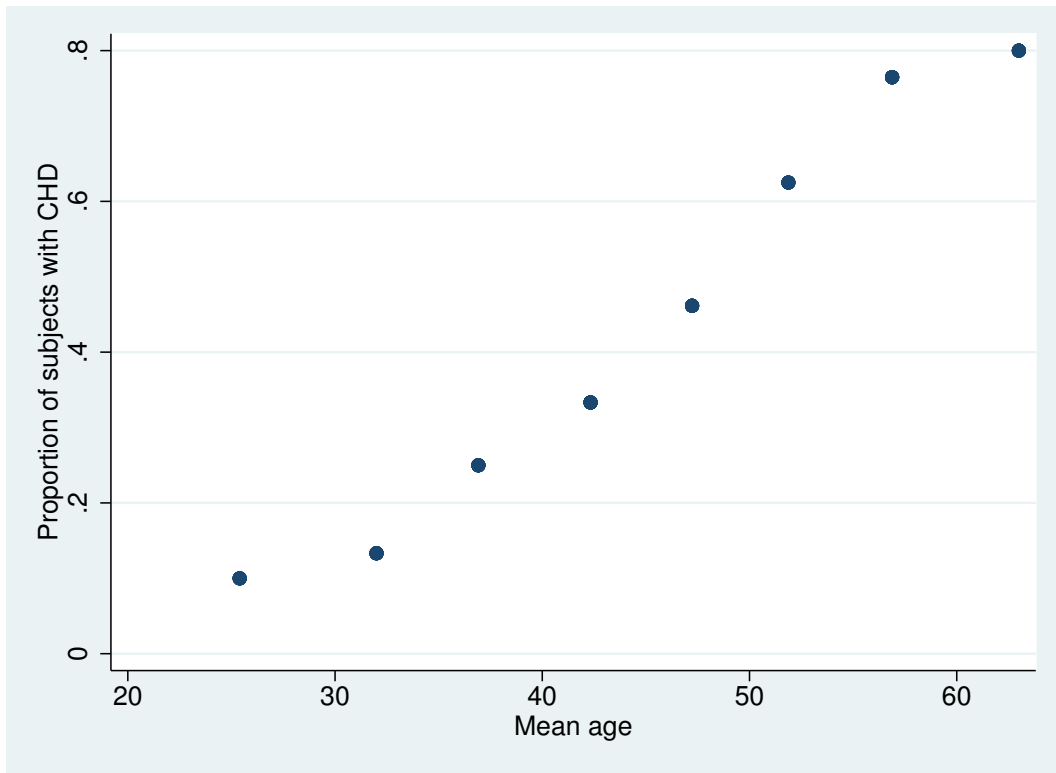


Figure 8: . scatter chdprop agemean

. summ db, detail

Pregibon's dbeta

Percentiles		Smallest		
1%	.0004816	.0004816	Obs	100
5%	.0006804	.0004816	Sum of Wgt.	100
10%	.0009566	.0004816	Mean	.0247488
25%	.0052691	.000628	Std. Dev.	.0409846
50%	.0138935		Variance	.0016797
		Largest	Skewness	3.889923
75%	.0250462	.1390621	Kurtosis	19.78815
90%	.0471688	.1390621		
95%	.0975644	.2487164		
99%	.2487164	.2487164		

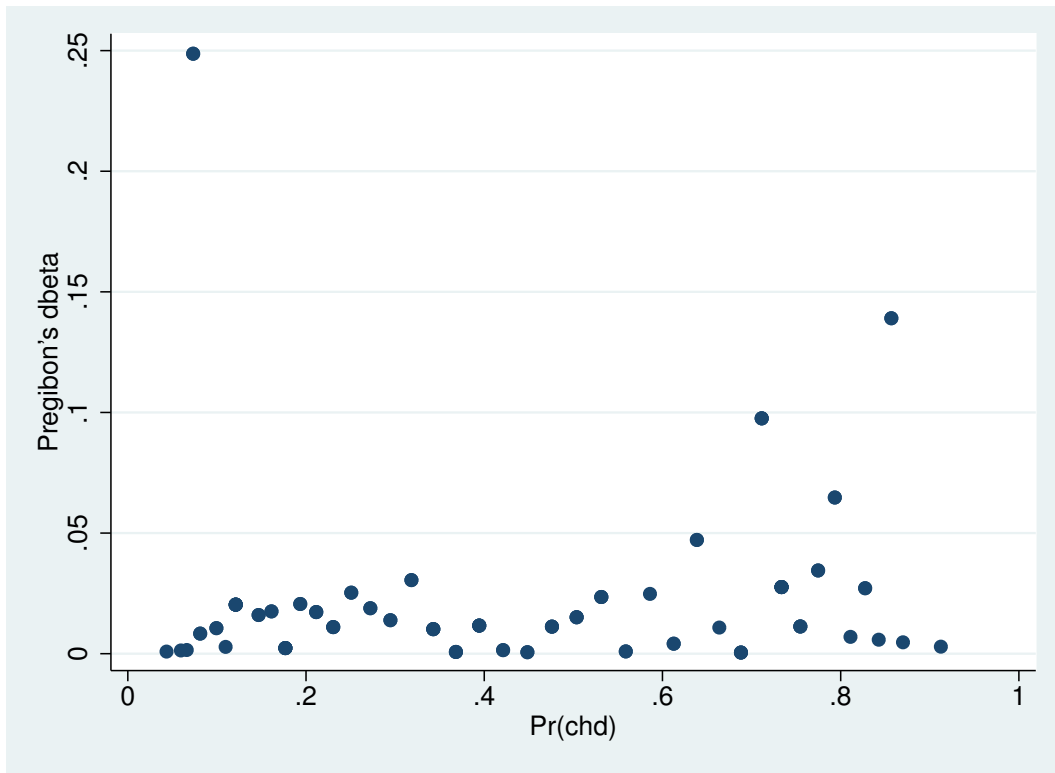


Figure 9: . scatter db p

```
. logistic chd age if db < 0.2
Logistic regression                Number of obs   =       98
                                   LR chi2(1)        =       32.12
                                   Prob > chi2       =       0.0000
Log likelihood = -50.863658        Pseudo R2      =       0.2400
```

	chd	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
	age	1.130329	.0293066	4.73	0.000	1.074324 1.189254
	_cons	.0027493	.0033856	-4.79	0.000	.0002461 .0307199

5.4 The effect on the odds ratio is small: a very slight increase to 1.13
end of do-file