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Descriptive

codebook chd

```
-----
chd                                     Evidence of Coronary Heart Disease
-----
      type:  numeric (double)
      label:  CHD

      range:  [0,1]
unique values: 2                               units:  1
                                              missing .:  0/100

      tabulation:  Freq.   Numeric   Label
                   57       0       No
                   43       1       Yes
-----
```

tab chd

Evidence of Coronary Heart Disease	Freq.	Percent	Cum.
No	57	57.00	57.00
Yes	43	43.00	100.00
Total	100	100.00	

Logistic

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
-----
```

listcoef

listcoef, percent help

logistic (N=100): Percentage Change in Odds

Odds of: Yes vs No

```
-----
chd |      b          z    P>|z|     %     %StdX     SDofX
-----+-----
age |  0.11092     4.610    0.000    11.7    267.0    11.7213
-----
```

```
b = raw coefficient
z = z-score for test of b=0
P>|z| = p-value for z-test
% = percent change in odds for unit increase in X
%StdX = percent change in odds for SD increase in X
SDofX = standard deviation of X
```

prchange

prchange age, help

logistic: Changes in Probabilities for chd

	min->max	0->1	-+1/2	-+sd/2	MargEfct
age	0.8690	0.0006	0.0267	0.3037	0.0267

	No	Yes
Pr(y x)	0.5955	0.4045

age
 x= 44.38
 sd(x)= 11.7213

Pr(y|x): probability of observing each y for specified x values
 Avg|Chg|: average of absolute value of the change across categories
 Min->Max: change in predicted probability as x changes from its minimum to its maximum
 0->1: change in predicted probability as x changes from 0 to 1
 -+1/2: change in predicted probability as x changes from 1/2 unit below base value to 1/2 unit above
 -+sd/2: change in predicted probability as x changes from 1/2 standard dev below base to 1/2 standard dev above
 MargEfct: the partial derivative of the predicted probability/rate with respect to a given independent variable

prchange age, delta(10)

logistic: Changes in Probabilities for chd

(Note: d = 10)

	min->max	0->1	-+d/2	-+sd/2	MargEfct
age	0.8690	0.0006	0.2612	0.3037	0.0267

	No	Yes
Pr(y x)	0.5955	0.4045

←----- This is the probability for CHD = "Yes" at mean of age

age
 x= 44.38
 sd(x)= 11.7213

prvalue

prvalue, x(age=44.38)

logistic: Predictions for chd

Confidence intervals by delta method

		95% Conf. Interval	
Pr(y=Yes x):	0.4045	[0.2913,	0.5177]
Pr(y=No x):	0.5955	[0.4823,	0.7087]

age
 x= 44.38

Classification table

estat

estat class, cutoff(0.501)

Logistic model for chd

Classified	True		Total
	D	~D	
+	29	12	41
-	14	45	59
Total	43	57	100

Classified + if predicted Pr(D) >= .501
 True D defined as chd != 0

Sensitivity	Pr(+ D)	67.44%
Specificity	Pr(- ~D)	78.95%
Positive predictive value	Pr(D +)	70.73%
Negative predictive value	Pr(~D -)	76.27%
False + rate for true ~D	Pr(+ ~D)	21.05%
False - rate for true D	Pr(- D)	32.56%
False + rate for classified +	Pr(~D +)	29.27%
False - rate for classified -	Pr(D -)	23.73%
Correctly classified		74.00%

Same order as in SPSS:

predict pr

gen predicted_cut05 = pr > 0.5

tab2 chd predicted_cut05, row

Evidence of Coronary Heart Disease	predicted_cut05		Total
	0	1	
No	45 78.95	12 21.05	57 100.00
Yes	14 32.56	29 67.44	43 100.00
Total	59 59.00	41 41.00	100 100.00

display (45+29)/100
 .74

SPSS was:

Classification Table^a

Observed		Predicted		
		Evidence of Coronary Heart Disease		Percentage Correct
		No	Yes	
Step 1	Evidence of Coronary Heart Disease	No	Yes	
		45	12	78,9
		14	29	67,4
	Overall Percentage			74,0

a. The cut value is ,500

prtab

prtab age

logistic: Predicted probabilities of positive outcome for chd

Age (years)	Prediction
20	0.0435
23	0.0596
24	0.0662
25	0.0733
26	0.0812
28	0.0994
29	0.1098
...	
59	0.7747
60	0.7934
61	0.8110
62	0.8274
63	0.8427
64	0.8569
65	0.8699
69	0.9125

age
x= 44.38

Test differences

prvalue, x(age=30) save

logistic: Predictions for chd

Confidence intervals by delta method

		95% Conf. Interval
Pr(y=Yes x):	0.1211	[0.0265, 0.2157]
Pr(y=No x):	0.8789	[0.7843, 0.9735]

age
x= 30

rvalue, x(age=60) dif

logistic: Change in Predictions for chd

Confidence intervals by delta method

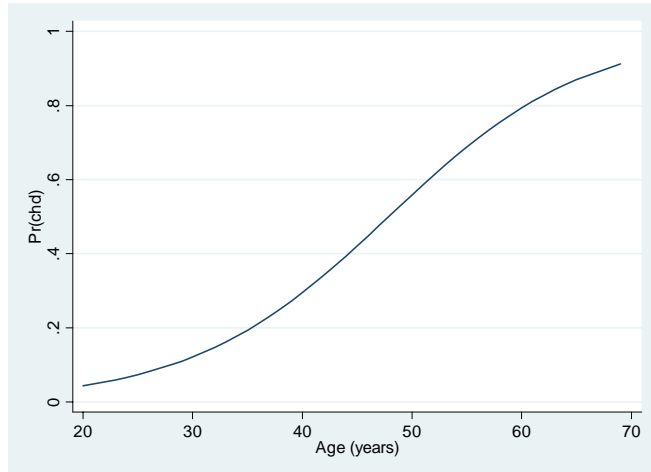
	Current	Saved	Change	95% CI for Change
Pr(y=Yes x):	0.7934	0.1211	0.6723	[0.4822, 0.8625]
Pr(y=No x):	0.2066	0.8789	-0.6723	[-0.8625, -0.4822]

age
Current= 60
Saved= 30
Diff= 30

Plots

postgr3

`postgr3 age`



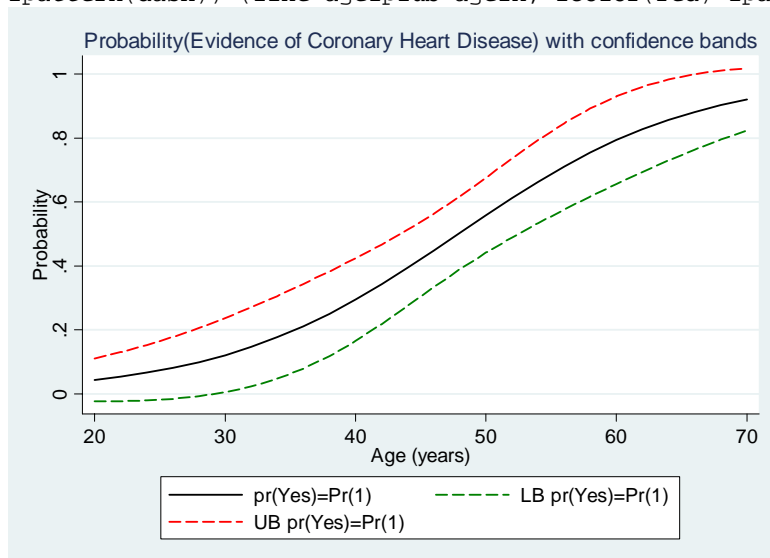
Plot with confidence

`prgen age, from (20) to (70) gap(2) generate (age2) rest(mean) ci`

* creates new variables:

Variable	Label (automatic)	explanation
age2x	Age (years)	takes values from 20 to 70 by 2
age2p0	pr(No)=Pr(0)	prob(0)
age2p1	pr(Yes)=Pr(1)	prob(1)
age2p0lb	LB pr(No)=Pr(0)	low bound of ci 95% for Pr(0)
age2p1lb	LB pr(Yes)=Pr(1)	low bound of ci 95% for Pr(1)
age2p0ub	UB pr(No)=Pr(0)	upper bound of ci 95% for Pr(0)
age2p1ub	UB pr(Yes)=Pr(1)	upper bound of ci 95% for Pr(1)

`twoway (line age2p1 age2x, lcolor(black)) (line age2p1lb age2x, lcolor(green) lpattern(dash)) (line age2p1ub age2x, lcolor(red) lpattern(dash))`



Standardized residuals and leverage

Leverage = $dbeta$ = Pregibon Delta-Beta influence statistic = counterpart to Cook's distance in linear regression

"We have seen quite a few logistic regression diagnostic statistics. Now how large does each one have to be, to be considered influential?"

First of all, we always have to make our judgment based on our theory and our analysis. Secondly, there are some rule-of-thumb cutoffs when the sample size is large. These are shown below.

When the sample size is large, the asymptotic distribution of some of the measures would follow some standard distribution.

That is why we have these cutoff values, and why they only apply when the sample size is large enough.

Usually, we would look at the relative magnitude of a statistic an observation has compared to others.

That is, we look for data points that are farther away from most of the data points. "

Measure	Value
leverage (hat value)	>2 or 3 times of the average of leverage
abs(Pearson Residuals)	> 2
abs(Deviance Residuals)	> 2

<http://www.ats.ucla.edu/stat/Stata/webbooks/logistic/chapter3/statalog3.htm>

Standardized residuals

```
predict prob  
label var prob "Prob(CHD=1)"
```

```
predict stdres, res  
label var std "Std residual"
```

```
gen predict="yes" if prob>0.5  
replace predict="no" if prob<=0.5
```

```
list stdres age if abs(stdres)>2
```

	stdres	age
4.	2.314473	25
5.	2.314473	25

Leverage

```
predict cook, dbeta  
label var cook "Cook's statistics"
```

```
gen obs=_n  
scatter cook obs, mlabel(obs) msize(1) mlabsize(2) jitter(*10)
```

