

Part 1: Logistic Regression Analysis for cross-sectional data

Data: Myocardial Infraction.

Variables: Oral Contraceptive indicator (0: No, 1:Yes)

Age in Years

Smoking Indicator (0: No, 1: Yes)

MI (0: No, 1: Yes)

Question: Is MI related to OC use, age and smoking?

Method: Logistic regression

Exploratory Analysis

Relation between smoking and MI

```
. tabulate smoking MI
```

smoking	MI		Total
	0	1	
0	117	17	134
1	40	26	66
Total	157	43	200

```
. cci 26 17 40 117
```

	Exposed	Unexposed	Total	Proportion Exposed
Cases	26	17	43	0.6047
Controls	40	117	157	0.2548
Total	66	134	200	0.3300
	Point estimate		[95% Conf. Interval]	
Odds ratio	4.473529		2.215584	9.031184
Attr. frac. ex.	.7764629		.5486518	.8892725
Attr. frac. pop	.4694892			
	chi2(1) =		18.69	Pr>chi2 = 0.0000

OR: 4.47, CI: (2.21, 9.03), p-value: 0.00 (Ho: OR = 1)

Conclusion:

Relation between Oral contraceptive use and MI

```
. tabulate OC MI
```

OC	MI		Total
	0	1	
0	93	9	102
1	64	34	98
Total	157	43	200

```
. cci 34 9 64 93
```

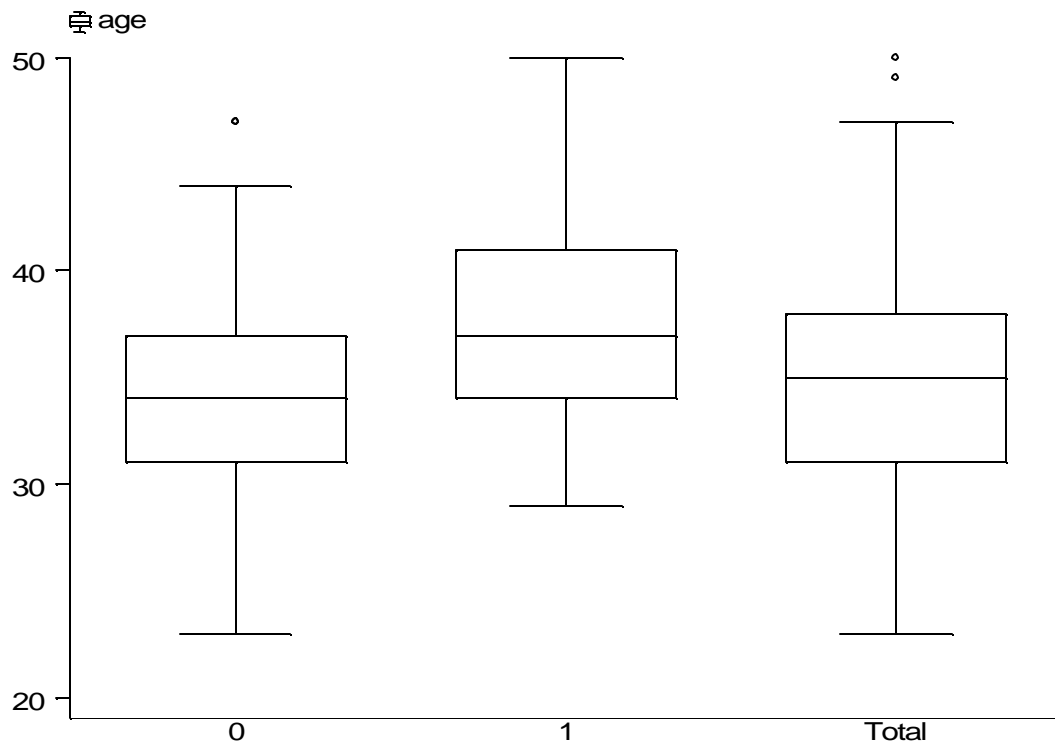
	Exposed	Unexposed	Total	Proportion Exposed
Cases	34	9	43	0.7907
Controls	64	93	157	0.4076
Total	98	102	200	0.4900
	Point estimate		[95% Conf. Interval]	
Odds ratio	5.489583		2.496017	12.0376
Attr. frac. ex.	.8178368		.5993616	.9169269
Attr. frac. pop.	.6466617			

chi2(1) = 19.82 Pr>chi2 = 0.0000

OR: 5.4 , CI: (2.30, 12.04), p-value: 0.00 (Ho: OR = 1)

Age and MI

```
. graph age, box by(MI) total s(o) ylab
```



```
. summ age if MI == 1
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	43	37.90698	5.079435	29	50

```
. summ age if MI == 0
```

Variable	Obs	Mean	Std. Dev.	Min	Max
age	157	34.12102	4.682688	23	47

Now let's perform a two sample t test.

```
. ttesti 43 37.907 5.079 157 34.121 4.683
```

	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
x	43	37.907	.7745402	5.079	36.34391	39.47009
y	157	34.121	.3737441	4.683	33.38275	34.85925
combined	200	34.93499	.3540305	5.006747	34.23686	35.63312
diff		3.786	.8209683		2.167036	5.404964

Degrees of freedom: 198

Ho: mean(x) - mean(y) = diff = 0

Ha: diff < 0	Ha: diff ~= 0	Ha: diff > 0
t = 4.6116	t = 4.6116	t = 4.6116
P < t = 1.0000	P > t = 0.0000	P > t = 0.0000

Now lets perform logistic regression analysis.

```
. glm MI age smoking OC, f(bin) l(logit)
```

```
Iteration 1 : deviance = 156.9761
Iteration 2 : deviance = 150.7970
Iteration 3 : deviance = 150.3779
Iteration 4 : deviance = 150.3748
Iteration 5 : deviance = 150.3748
```

```
Residual df = 196 No. of obs = 200
Pearson X2 = 177.5298 Deviance = 150.3748
Dispersion = .9057645 Dispersion = .7672184
```

Bernoulli distribution, logit link

MI	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
age	.1625556	.0445356	3.650	0.000	.0752673	.2498438
smoking	1.812237	.429383	4.221	0.000	.9706619	2.653812
OC	1.979896	.4696779	4.215	0.000	1.059344	2.900448
_cons	-9.114047	1.757031	-5.187	0.000	-12.55777	-5.670329

```
. glm, eform
```

```
Residual df =      196          No. of obs =      200
Pearson X2   = 177.5298        Deviance   = 150.3748
Dispersion  =  .9057645        Dispersion =  .7672184
```

```
Bernoulli distribution, logit link
```

MI	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
age	1.176514	.0523968	3.650	0.000	1.078172	1.283825
smoking	6.124132	2.629598	4.221	0.000	2.639691	14.2081
OC	7.241988	3.401402	4.215	0.000	2.884478	18.18228

Part 2: Logistic Regression Analysis for longitudinal data with random effects.

Data: Wheezing

Model:

$$\text{logit Pr}(Y_{ij} = 1 | U_i) = \beta_0 + U_i + \mathbf{b} \mathbf{X}$$

We assume that conditional on the unobservable responses U_i , we have independent responses from a distribution in exponential family.

Interpretation of parameters: Consider the following model, logit for the i th individual in the study.

$$\text{logit Pr}(Y_i = 1 | U_i) = \beta_0^* + U_i + \beta_1^* x_{ij}$$

where x_{ij} is 1 if child i is vitamin A deficient and 0 otherwise. Distribution of U_i is normal with mean 0 and unknown variance v^2 . In this example β_0^* is the log odds of respiratory infection for a typical child with random effect $U_i = 0$. The parameter β_1^* is the log odds for respiratory infection when a child is deficient relative to when that same child is not. The variance v^2 represents the degree of heterogeneity across children in the propensity of disease not attributable to x .

Let's fit the random effects model with one covariate at time

Age and weezing

```
. xtlogit wee age, i(Id) re
```

Fitting comparison model:

```
Iteration 0:  log likelihood = -60.215168
Iteration 1:  log likelihood = -59.821289
Iteration 2:  log likelihood = -59.820619
```

Fitting full model:

```
rho = 0.0      log likelihood = -59.820619
```

```

rho = 0.1    log likelihood = -59.534747
rho = 0.2    log likelihood = -59.313489
rho = 0.3    log likelihood = -59.174587
rho = 0.4    log likelihood = -59.144047
rho = 0.5    log likelihood = -59.264113
Iteration 0: log likelihood = -59.144047
Iteration 1: log likelihood = -59.053072
Iteration 2: log likelihood = -59.052345
Iteration 3: log likelihood = -59.052344

```

```

Random-effects logit                    Number of obs    =    100
Group variable (i) : Id                 Number of groups =     32

Random effects u_i ~ Gaussian           Obs per group: min =     1
                                           avg =           3.1
                                           max =           4

Wald chi2(1) = 0.92
Log likelihood = -59.052344             Prob > chi2      = 0.3385

```

wee	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
age	-.2058959	.2151199	-0.957	0.339	-.6275231	.2157313
_cons	1.13757	2.24621	0.506	0.613	-3.264921	5.540061
/lnsig2u	-.3512171	1.122664	-0.313	0.754	-2.551598	1.849164
sigma_u	.8389463	.4709274			.2792078	2.520814
rho	.4130873	.2721856			.0723192	.8640289

```

Likelihood ratio test of rho=0:      chi2(1) = 1.54   Prob > chi2 = 0.2151

```

Now all the covariates

```
. xtlogit wee smk2 age City2, i(Id) re
```

Fitting comparison model:

```

Iteration 0: log likelihood = -60.215168
Iteration 1: log likelihood = -58.927241
Iteration 2: log likelihood = -58.919236
Iteration 3: log likelihood = -58.919235

```

Fitting full model:

```

rho = 0.0    log likelihood = -58.919235
rho = 0.1    log likelihood = -58.62208
rho = 0.2    log likelihood = -58.391529
rho = 0.3    log likelihood = -58.244856
rho = 0.4    log likelihood = -58.208104
rho = 0.5    log likelihood = -58.324007
Iteration 0: log likelihood = -58.208104
Iteration 1: log likelihood = -58.098593
Iteration 2: log likelihood = -58.097518
Iteration 3: log likelihood = -58.097517

```

```

Random-effects logit                    Number of obs    =    100
Group variable (i) : Id                 Number of groups =     32

```

Random effects u_i ~ Gaussian

Obs per group: min = 1
avg = 3.1
max = 4

Log likelihood = -58.097517

Wald chi2(3) = 2.62
Prob > chi2 = 0.4548

wee	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
smk2	-.6623059	.5200883	-1.273	0.203	-1.68166	.3570485
age	-.198715	.2181834	-0.911	0.362	-.6263465	.2289166
City2	.2836571	.5794732	0.490	0.624	-.8520895	1.419404
_cons	1.216874	2.287129	0.532	0.595	-3.265817	5.699565
/lnsig2u	-.3038675	1.09995	-0.276	0.782	-2.45973	1.851995
sigma_u	.8590452	.4724534			.292332	2.524385
rho	.4246123	.2687362			.0787299	.8643612

Likelihood ratio test of rho=0: chi2(1) = 1.64 Prob > chi2 = 0.1999

What are the interpretations of the parameters?