

Agresti Coronary Data

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Needed Packages

```
if(!require(FSA)){install.packages("FSA")}
if(!require(ggplot2)){install.packages("ggplot2")}
if(!require(car)){install.packages("car")}
if(!require(multcompView)){install.packages("multcompView")}
if(!require(lsmmeans)){install.packages("lsmmeans")}
if(!require(grid)){install.packages("grid")}
if(!require(nlme)){install.packages("nlme")}
if(!require(lme4)){install.packages("lme4")}
if(!require(Rmisc)){install.packages("Rmisc")}
if(!require(FSA)){install.packages("FSA")}
#if(!require(lmerTest)){install.packages("lmerTest")}
#if(!require(rcompanion)){install.packages("rcompanion")}
```

Example 1

Read data from SAS input file

```
Input = ("
sex ecg ca total
0 0 4 15
0 1 8 18
1 0 9 18
1 1 21 27
")
coronary1 = read.table(textConnection(Input),header=TRUE)
coronary1 = data.frame(coronary1)

sex = coronary1$sex
ecg = coronary1$ecg
ca = coronary1$ca
total = coronary1$total
per.ca = ca/total
```

Fit Logistic Model with Identity Link

```
mod1 = glm( per.ca ~ sex + ecg,
            family = "gaussian"(link="identity"),
            data=coronary1)
mod1
```

##

```
## Call: glm(formula = per.ca ~ sex + ecg, family = gaussian(link = "identity"),
## data = coronary1)
##
## Coefficients:
## (Intercept)      sex      ecg
## 0.2417      0.2833      0.2278
##
## Degrees of Freedom: 3 Total (i.e. Null); 1 Residual
## Null Deviance: 0.1347
## Residual Deviance: 0.0025 AIC: -10.16
```

Fit Logistic Model with LOGIT link

```
mod2 = glm( ca/total ~ sex + ecg,
            family = "binomial"(link="logit"),
            data=coronary1)
```

```
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
```

```
mod2
```

```
##
## Call: glm(formula = ca/total ~ sex + ecg, family = binomial(link = "logit"),
## data = coronary1)
##
## Coefficients:
## (Intercept)      sex      ecg
## -1.141      1.240      1.017
##
## Degrees of Freedom: 3 Total (i.e. Null); 1 Residual
## Null Deviance: 0.5656
## Residual Deviance: 0.01141 AIC: 9.874
```

```
#plot(mod1)
```

Fit Logistic Model with Cloglog link

```
mod3 = glm( ca/total ~ sex + ecg,
            family = "binomial"(link="cloglog"),
            data=coronary1)
```

```
## Warning in eval(family$initialize): non-integer #successes in a binomial glm!
```

```
mod3
```

```
##
## Call: glm(formula = ca/total ~ sex + ecg, family = binomial(link = "cloglog"),
## data = coronary1)
##
## Coefficients:
## (Intercept)      sex      ecg
## -1.2244      0.8864      0.7247
##
## Degrees of Freedom: 3 Total (i.e. Null); 1 Residual
## Null Deviance: 0.5656
## Residual Deviance: 0.001879 AIC: 9.749
```

```
#plot(mod2)
```

Fit Results

```
p_linear = 0.76 - 0.28*sex - 0.23*ecg
p_logit = exp(1.14-1.24*sex-1.02*ecg)/(1 + exp(1.14-1.24*sex-1.02*ecg))
p_c11 = 1 - exp(-exp(0.36 - 0.84*sex - 0.68*ecg))
p_linear = 1 - p_linear
p_logit = 1 - p_logit
p_c11 = 1 - p_c11
## Probability for CA event
cbind(sex,ecg,per.ca,p_linear,p_logit,p_c11)
```

```
##      sex ecg   per.ca p_linear  p_logit    p_c11
## [1,]  0   0 0.2666667    0.24 0.2423204 0.2385135
## [2,]  0   1 0.4444444    0.47 0.4700359 0.4837684
## [3,]  1   0 0.5000000    0.52 0.5249792 0.5385993
## [4,]  1   1 0.7777778    0.75 0.7539887 0.7308945
```

Odds For Sex and ECG

```
odds_sex = exp(1.24)
odds_sex                                     #male vs female for CA event
```

```
## [1] 3.455613
```

```
odds_ecg = exp(1.02)
odds_ecg                                     #high vs low (ECG) for CA event
```

```
## [1] 2.773195
```

Example 2

Read data from SAS input file

```
Input = ("
sex ecg age ca
0 0 28 0
1 0 42 1
0 1 46 0
1 1 45 0
0 0 34 0
1 0 44 1
0 1 48 1
1 1 45 1
0 0 38 0
1 0 45 0
0 1 49 0
1 1 45 1
0 0 41 1
1 0 46 0
0 1 49 0
1 1 46 1
0 0 44 0
1 0 48 0
0 1 52 0
1 1 48 1
0 0 45 1
```

```
1 0 50 0
0 1 53 1
1 1 57 1
0 0 46 0
1 0 52 1
0 1 54 1
1 1 57 1
0 0 47 0
1 0 52 1
0 1 55 0
1 1 59 1
0 0 50 0
1 0 54 0
0 1 57 1
1 1 60 1
0 0 51 0
1 0 55 0
0 2 46 1
1 1 63 1
0 0 51 0
1 0 59 1
0 2 48 0
1 2 35 0
0 0 53 0
1 0 59 1
0 2 57 1
1 2 37 1
0 0 55 1
1 1 32 0
0 2 60 1
1 2 43 1
0 0 59 0
1 1 37 0
1 0 30 0
1 2 47 1
0 0 60 1
1 1 38 1
1 0 34 0
1 2 48 1
0 1 32 1
1 1 38 1
1 0 36 1
1 2 49 0
0 1 33 0
1 1 42 1
1 0 38 1
1 2 58 1
0 1 35 0
1 1 43 0
1 0 39 0
1 2 59 1
0 1 39 0
1 1 43 1
```

```

1 0 42 0
1 2 60 1
0 1 40 0
1 1 44 1
")
coronary2 = read.table(textConnection(Input),header=TRUE)
coronary2 = data.frame(coronary2)

sex = coronary2$sex
ecg = coronary2$ecg
ca = coronary2$ca
age = coronary2$age
ca[ca==2] = 1

```

Fit Logistic Model with LOGIT link

```

mod2 = glm( ca ~ sex + ecg + age,
            family = "binomial"(link="logit"),
            data=coronary2)
mod2

##
## Call: glm(formula = ca ~ sex + ecg + age, family = binomial(link = "logit"),
## data = coronary2)
##
## Coefficients:
## (Intercept)      sex      ecg      age
## -5.64176      1.35643      0.87320      0.09285
##
## Degrees of Freedom: 77 Total (i.e. Null); 74 Residual
## Null Deviance:      107.9
## Residual Deviance: 86.81 AIC: 94.81

#plot(mod1)

```

Odds for CA

```

odds_sex = exp(1.36)
odds_sex      #male vs female for CA event

## [1] 3.896193

odds_ecg = exp(0.87)
odds_ecg      #high vs low (ECG) for CA event

## [1] 2.386911

odds_age = exp(0.09)
odds_age      #for each year of age for CA event

## [1] 1.094174

```

Fit Logistic Model with Cloglog link

```

mod3 = glm( ca ~ sex + ecg + age,
            family = "binomial"(link="cloglog"),
            data=coronary2)
mod3

```

```
##
## Call: glm(formula = ca ~ sex + ecg + age, family = binomial(link = "cloglog"),
## data = coronary2)
##
## Coefficients:
## (Intercept)          sex          ecg          age
## -4.48090      0.92131      0.56107      0.06809
##
## Degrees of Freedom: 77 Total (i.e. Null); 74 Residual
## Null Deviance: 107.9
## Residual Deviance: 86 AIC: 94
```

```
#plot(mod2)
```

Fit Results

```
p_logit = exp(-5.64+1.36*sex+0.87*ecg+0.09*age)/(1 + exp(-5.64+1.36*sex+0.87*ecg+0.09*age))
p_cll = 1 - exp(-exp(-4.48+0.92*sex+0.56*ecg+0.07*age))
cbind(sex,ecg,age,p_logit,p_cll)
```

	sex	ecg	age	p_logit	p_cll
## [1,]	0	0	28	0.04228977	0.07730783
## [2,]	1	0	42	0.37754067	0.41605264
## [3,]	0	1	46	0.34751054	0.39139468
## [4,]	1	1	45	0.65475346	0.68708830
## [5,]	0	0	34	0.07043673	0.11525555
## [6,]	1	0	44	0.42067575	0.46140070
## [7,]	0	1	48	0.38936077	0.43515790
## [8,]	1	1	45	0.65475346	0.68708830
## [9,]	0	0	38	0.09796880	0.14958069
## [10,]	1	0	45	0.44275215	0.48503186
## [11,]	0	1	49	0.41095957	0.45807431
## [12,]	1	1	45	0.65475346	0.68708830
## [13,]	0	0	41	0.12455336	0.18117723
## [14,]	1	0	46	0.46505705	0.50922540
## [15,]	0	1	49	0.41095957	0.45807431
## [16,]	1	1	46	0.67480527	0.71236896
## [17,]	0	0	44	0.15709547	0.21854442
## [18,]	1	0	48	0.50999867	0.55900897
## [19,]	0	1	52	0.47751518	0.53035761
## [20,]	1	1	48	0.71300016	0.76148651
## [21,]	0	0	45	0.16938390	0.23239290
## [22,]	1	0	50	0.55477924	0.61006083
## [23,]	0	1	53	0.50000000	0.55540176
## [24,]	1	1	57	0.84812884	0.93220281
## [25,]	0	0	46	0.18242552	0.24697287
## [26,]	1	0	52	0.59868766	0.66151892
## [27,]	0	1	54	0.52248482	0.58077950
## [28,]	1	1	57	0.84812884	0.93220281
## [29,]	0	0	47	0.19623406	0.26230238
## [30,]	1	0	52	0.59868766	0.66151892
## [31,]	0	1	55	0.54487889	0.60638965
## [32,]	1	1	59	0.86989153	0.95475470
## [33,]	0	0	50	0.24232036	0.31292450
## [34,]	1	0	54	0.64106741	0.71236896

## [35,]	0	1	57	0.58904043	0.65785073
## [36,]	1	1	60	0.87974314	0.96385140
## [37,]	0	0	51	0.25922510	0.33136986
## [38,]	1	0	55	0.66150316	0.73721721
## [39,]	0	2	46	0.55971365	0.58077950
## [40,]	1	1	63	0.90550963	0.98336016
## [41,]	0	0	51	0.25922510	0.33136986
## [42,]	1	0	59	0.73691590	0.82937158
## [43,]	0	2	48	0.60348325	0.63212056
## [44,]	1	2	35	0.64794080	0.63579929
## [45,]	0	0	53	0.29525430	0.37061559
## [46,]	1	0	59	0.73691590	0.82937158
## [47,]	0	2	57	0.77381857	0.84704486
## [48,]	1	2	37	0.68783133	0.68708830
## [49,]	0	0	55	0.33403307	0.41291860
## [50,]	1	1	32	0.37051689	0.37353754
## [51,]	0	2	60	0.81757448	0.90136873
## [52,]	1	2	43	0.79084063	0.82937158
## [53,]	0	0	59	0.41824062	0.50573728
## [54,]	1	1	37	0.48001066	0.48503186
## [55,]	1	0	30	0.17079548	0.20724121
## [56,]	1	2	47	0.84422416	0.90363834
## [57,]	0	0	60	0.44028635	0.53035761
## [58,]	1	1	38	0.50249998	0.50922540
## [59,]	1	0	34	0.22793645	0.26455442
## [60,]	1	2	48	0.85569687	0.91867407
## [61,]	0	1	32	0.13124447	0.17003685
## [62,]	1	1	38	0.50249998	0.50922540
## [63,]	1	0	36	0.26114999	0.29774218
## [64,]	1	2	49	0.86645828	0.93220281
## [65,]	0	1	33	0.14185106	0.18117723
## [66,]	1	1	42	0.59145898	0.61006083
## [67,]	1	0	38	0.29733935	0.33406929
## [68,]	1	2	58	0.93583612	0.99361044
## [69,]	0	1	35	0.16520487	0.20540719
## [70,]	1	1	43	0.61301418	0.63579929
## [71,]	1	0	39	0.31647911	0.35341414
## [72,]	1	2	59	0.94103299	0.99557055
## [73,]	0	1	39	0.22097389	0.26230238
## [74,]	1	1	43	0.61301418	0.63579929
## [75,]	1	0	42	0.37754067	0.41605264
## [76,]	1	2	60	0.94583330	0.99700987
## [77,]	0	1	40	0.23685498	0.27839674
## [78,]	1	1	44	0.63413559	0.66151892