

R15 - Interactions in Logistic and Poisson Regression

HCI/PSYCH 522
Iowa State University

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Interactions

Independent variables

- Categorical-continuous
- Categorical-categorical
- Continuous-continuous (not discussed)

Generalized linear models

- Poisson regression
- Logistic regression

Poisson regression

Poisson regression with two (continuous) independent variables:

- Data model

$$Y_i \overset{ind}{\sim} Po(\lambda_i)$$

- Main effects model

$$\log(\lambda_i) = \beta_0 + \beta_1 X_{i,1} + \beta_2 X_{i,2}$$

- Interaction model

$$\log(\lambda_i) = \beta_0 + \beta_1 X_{i,1} + \beta_2 X_{i,2} + \beta_3 X_{i,1} X_{i,2}$$

The interaction occurs on the **log** scale.

Warpbreaks - Descriptive Statistics

```
head(warpbreaks)
```

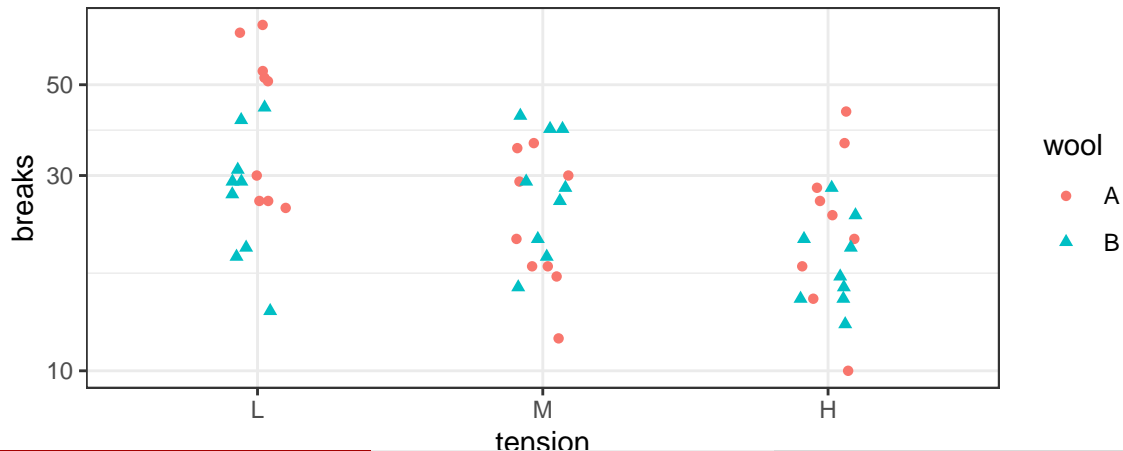
```
##   breaks wool tension
## 1     26    A        L
## 2     30    A        L
## 3     54    A        L
## 4     25    A        L
## 5     70    A        L
## 6     52    A        L
```

```
summary(warpbreaks)
```

```
##      breaks      wool  tension
## Min.   :10.00  A:27    L:18
## 1st Qu.:18.25  B:27    M:18
## Median :26.00             H:18
## Mean   :28.15
## 3rd Qu.:34.00
## Max.   :70.00
```

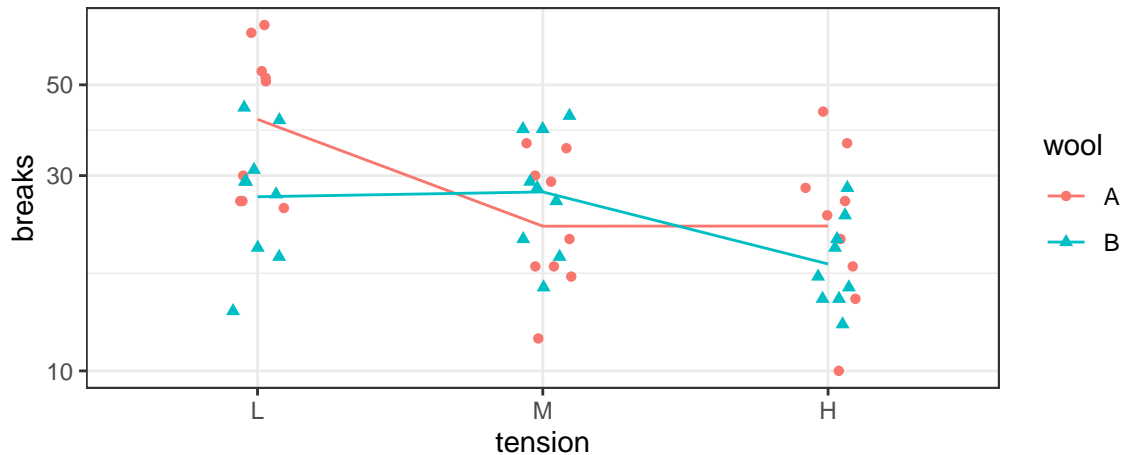
Warpbreaks - Graphical Statistics

```
g <- ggplot(warpbreaks, aes(x = tension, y = breaks,  
                             color = wool, shape = wool, group = wool)) +  
  geom_jitter(width=0.1, height=0) + scale_y_log10()  
g
```



Warpbreaks - Graphical Statistics

```
g + stat_summary(fun = mean, geom = "line")
```



Warpbreaks - Models

```

mM <- glm(breaks ~ wool + tension, data = warpbreaks, family = poisson) # Main effects model
mI <- glm(breaks ~ wool * tension, data = warpbreaks, family = poisson) # Interaction model

drop1(mI, test="Chi")

## Single term deletions
##
## Model:
## breaks ~ wool * tension
##           Df Deviance    AIC    LRT Pr(>Chi)
## <none>           182.31 468.97
## wool:tension  2    210.39 493.06 28.087 7.962e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

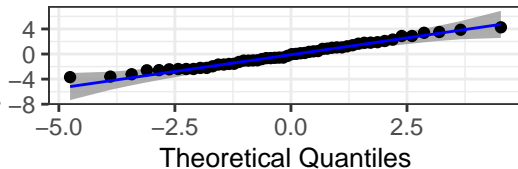
```

Warpbreaks - Diagnostics

```
resid_panel(mM, plots = c("qq", "resid", "index", "cookd"), qqbands = TRUE)
```

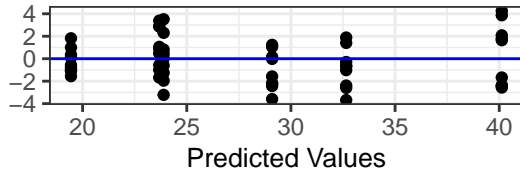
Deviance Residuals Sample Quantiles

Q-Q Plot

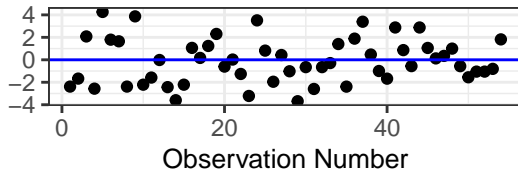


Deviance Residuals

Residual Plot

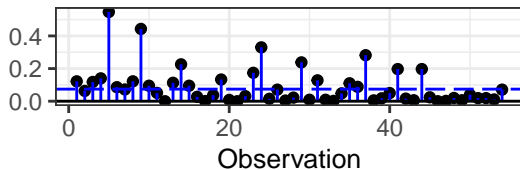


Index Plot



COOK's D

COOK's D Plot



Warpbreaks - Main effects model

```
summary(mM)

##
## Call:
## glm(formula = breaks ~ wool + tension, family = poisson, data = warpbreaks)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -3.6871  -1.6503  -0.4269   1.1902   4.2616
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   3.69196    0.04541  81.302 < 2e-16 ***
## woolB        -0.20599    0.05157  -3.994 6.49e-05 ***
## tensionM     -0.32132    0.06027  -5.332 9.73e-08 ***
## tensionH     -0.51849    0.06396  -8.107 5.21e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
##      Null deviance: 297.37  on 53  degrees of freedom
## Residual deviance: 210.39  on 50  degrees of freedom
## AIC: 493.06
##
## Number of Fisher Scoring iterations: 4
```

Warpbreaks - Main effects model

```
em <- emmeans(mM, pairwise ~ wool)
cm <- confint(em, type = "unlink"); cm
```

```
## $emmeans
##   wool rate      SE  df asymp.LCL asymp.UCL
##   A    30.3 1.061 Inf      28.3      32.5
##   B    24.7 0.955 Inf      22.9      26.6
##
## Results are averaged over the levels of: tension
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
##
## $contrasts
##   contrast ratio      SE  df asymp.LCL asymp.UCL
##   A / B      1.23 0.0634 Inf      1.11      1.36
##
## Results are averaged over the levels of: tension
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
```

Warpbreaks - Main effects model

```
et <- emmeans(mM, pairwise ~ tension)
ct <- confint(et, type = "unlink"); ct
```



```
## $emmeans
##   tension rate    SE df asymp.LCL asymp.UCL
##   L         36.2 1.42 Inf    33.5    39.1
##   M         26.2 1.21 Inf    24.0    28.7
##   H         21.6 1.09 Inf    19.5    23.8
##
## Results are averaged over the levels of: wool
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
##
## $contrasts
##   contrast ratio    SE df asymp.LCL asymp.UCL
##   L / M      1.38 0.0831 Inf     1.20     1.59
##   L / H      1.68 0.1074 Inf     1.45     1.95
##   M / H      1.22 0.0832 Inf     1.04     1.43
##
## Results are averaged over the levels of: wool
## Confidence level used: 0.95
## Conf-level adjustment: tukey method for comparing a family of 3 estimates
## Intervals are back-transformed from the log scale
```

Warpbreaks - Main effects model

```
g + geom_line(mapping=aes(y=predict(mM, warpbreaks, type = "unlink")))  
  
## Error in match.arg(type): 'arg' should be one of "link", "response", "terms"
```

Warpbreaks - Interaction model

```
summary(mI)

##
## Call:
## glm(formula = breaks ~ wool * tension, family = poisson, data = warpbreaks)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -3.3383  -1.4844  -0.1291   1.1725   3.5153
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   3.79674    0.04994  76.030 < 2e-16 ***
## woolB         -0.45663    0.08019  -5.694 1.24e-08 ***
## tensionM      -0.61868    0.08440  -7.330 2.30e-13 ***
## tensionH      -0.59580    0.08378  -7.112 1.15e-12 ***
## woolB:tensionM  0.63818    0.12215   5.224 1.75e-07 ***
## woolB:tensionH  0.18836    0.12990   1.450   0.147
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
##      Null deviance: 297.37  on 53  degrees of freedom
## Residual deviance: 182.31  on 48  degrees of freedom
## AIC: 468.97
##
## Number of Fisher Scoring iterations: 4
```

Warpbreaks - Interaction model

```
em <- emmeans(mI, pairwise ~ wool | tension)
cm <- confint(em, type = "unlink"); cm$emmeans
```



```
## tension = L:
##   wool rate   SE  df asymp.LCL asymp.UCL
##   A    44.6 2.22 Inf     40.4     49.1
##   B    28.2 1.77 Inf     25.0     31.9
##
## tension = M:
##   wool rate   SE  df asymp.LCL asymp.UCL
##   A    24.0 1.63 Inf     21.0     27.4
##   B    28.8 1.79 Inf     25.5     32.5
##
## tension = H:
##   wool rate   SE  df asymp.LCL asymp.UCL
##   A    24.6 1.65 Inf     21.5     28.0
##   B    18.8 1.44 Inf     16.1     21.8
##
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
```

Warpbreaks - Interaction model

```
cm$contrasts

## tension = L:
## contrast ratio      SE df asymp.LCL asymp.UCL
## A / B      1.579 0.1266 Inf      1.349      1.847
##
## tension = M:
## contrast ratio      SE df asymp.LCL asymp.UCL
## A / B      0.834 0.0768 Inf      0.696      0.999
##
## tension = H:
## contrast ratio      SE df asymp.LCL asymp.UCL
## A / B      1.308 0.1336 Inf      1.070      1.598
##
## Confidence level used: 0.95
## Intervals are back-transformed from the log scale
```

Warpbreaks - Interaction model

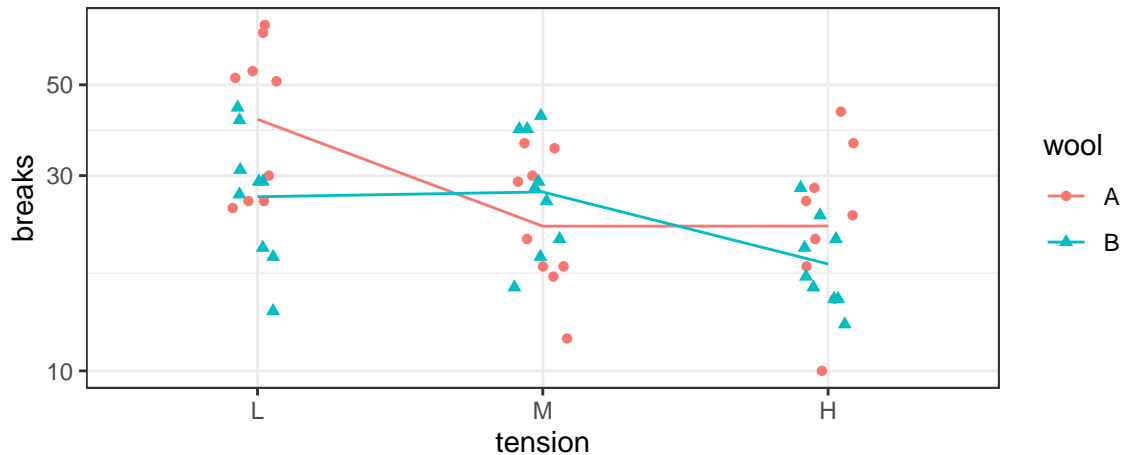
```
et <- emmeans(mI, pairwise ~ tension | wool)
ct <- confint(et, type = "unlink"); ct$contrasts
```



```
## wool = A:
## contrast ratio      SE df asymp.LCL asymp.UCL
## L / M      1.856 0.1567 Inf      1.523      2.26
## L / H      1.814 0.1520 Inf      1.491      2.21
## M / H      0.977 0.0935 Inf      0.781      1.22
##
## wool = B:
## contrast ratio      SE df asymp.LCL asymp.UCL
## L / M      0.981 0.0866 Inf      0.797      1.21
## L / H      1.503 0.1492 Inf      1.191      1.90
## M / H      1.533 0.1515 Inf      1.216      1.93
##
## Confidence level used: 0.95
## Conf-level adjustment: tukey method for comparing a family of 3 estimates
## Intervals are back-transformed from the log scale
```


Warpbreaks - Interaction model

```
g + stat_summary(fun = mean, geom = "line")
```



Logistic regression

Logistic regression with two (continuous) independent variables.

- Data model

$$Y_g \stackrel{ind}{\sim} \text{Bin}(n_g, \theta_g)$$

- Main effects model

$$\log \left(\frac{\theta_g}{1 - \theta_g} \right) = \beta_0 + \beta_1 X_{g,1} + \beta_2 X_{g,2}$$

- Interaction model

$$\log \left(\frac{\theta_g}{1 - \theta_g} \right) = \beta_0 + \beta_1 X_{g,1} + \beta_2 X_{g,2} + \beta_3 X_{g,1} X_{g,2}$$

The interaction occurs on the **log-odds** scale.

Donner Party - Descriptive Statistics

```
head(case2001)
```

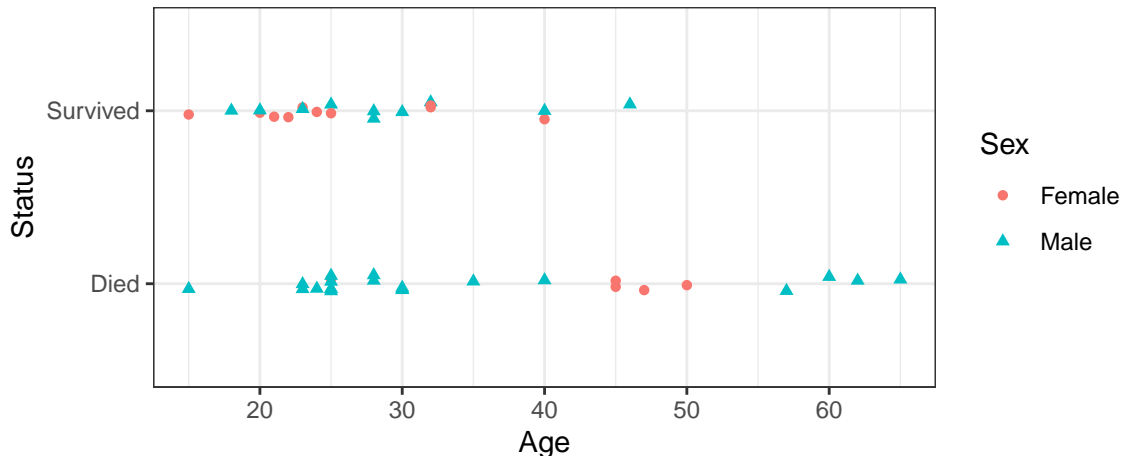
```
##   Age    Sex   Status
## 1  23  Male    Died
## 2  40 Female Survived
## 3  40  Male Survived
## 4  30  Male    Died
## 5  28  Male    Died
## 6  40  Male    Died
```

```
summary(case2001)
```

```
##           Age           Sex           Status
## Min.   :15.0  Female:15  Died    :25
## 1st Qu.:24.0  Male  :30  Survived:20
## Median :28.0
## Mean   :31.8
## 3rd Qu.:40.0
## Max.   :65.0
```

Donner Party - Graphical Statistics

```
g <- ggplot(case2001, aes(x = Age, y = Status, color = Sex, shape = Sex)) +  
  geom_jitter(width=0, height = 0.05)  
g
```



Donner Party - Models

```
mM <- glm(Status == "Survived" ~ Sex + Age, data = case2001, family = binomial) # Main effects model
mI <- glm(Status == "Survived" ~ Sex * Age, data = case2001, family = binomial) # Interaction model

drop1(mI, test="Chi")

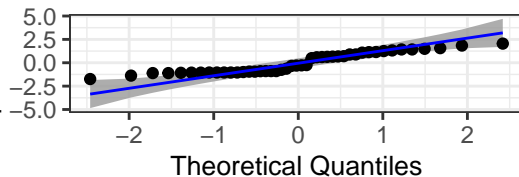
## Single term deletions
##
## Model:
## Status == "Survived" ~ Sex * Age
##           Df Deviance   AIC    LRT Pr(>Chi)
## <none>          47.346 55.346
## Sex:Age   1    51.256 57.256 3.9099   0.048 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Donner Party - Diagnostics

```
resid_panel(mM, plots = c("qq", "resid", "index", "cookd"), qqbands = TRUE)
```

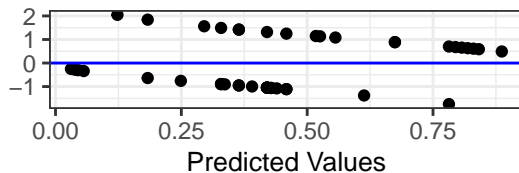
Deviance Residuals Sample Quantiles

Q-Q Plot

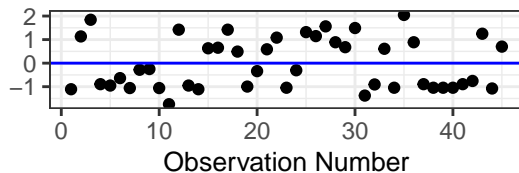


Deviance Residuals

Residual Plot

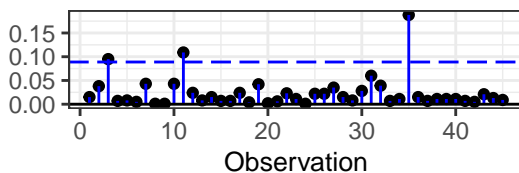


Index Plot



COOK's D

COOK's D Plot



Donner Party - Main effects model

```
summary(mM)

##
## Call:
## glm(formula = Status == "Survived" ~ Sex + Age, family = binomial,
##      data = case2001)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.7445  -1.0441  -0.3029   0.8877   2.0472
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  3.23041    1.38686   2.329   0.0198 *
## SexMale      -1.59729    0.75547  -2.114   0.0345 *
## Age          -0.07820    0.03728  -2.097   0.0359 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 61.827  on 44  degrees of freedom
## Residual deviance: 51.256  on 42  degrees of freedom
## AIC: 57.256
##
## Number of Fisher Scoring iterations: 4
```

Donner Party - Main effects model

```
em <- emmeans(mM, pairwise ~ Sex | Age, at = list(Age = c(15,28,65)))
cm <- confint(em, type = "unlink"); cm$emmeans
```

```
## Age = 15:
## Sex      response      SE  df asymp.LCL asymp.UCL
## Female   0.8867 0.0918 Inf   0.56620   0.979
## Male     0.6130 0.1484 Inf   0.31742   0.844
##
## Age = 28:
## Sex      response      SE  df asymp.LCL asymp.UCL
## Female   0.7390 0.1217 Inf   0.45129   0.907
## Male     0.3643 0.0949 Inf   0.20429   0.561
##
## Age = 65:
## Sex      response      SE  df asymp.LCL asymp.UCL
## Female   0.1355 0.1539 Inf   0.01181   0.673
## Male     0.0308 0.0433 Inf   0.00185   0.353
##
## Confidence level used: 0.95
## Intervals are back-transformed from the logit scale
```


Donner Party - Main effects model

```
cm$contrasts
```

```
## Age = 15:
## contrast      odds.ratio    SE  df asymp.LCL asymp.UCL
## Female / Male      4.94 3.73 Inf      1.12     21.7
##
## Age = 28:
## contrast      odds.ratio    SE  df asymp.LCL asymp.UCL
## Female / Male      4.94 3.73 Inf      1.12     21.7
##
## Age = 65:
## contrast      odds.ratio    SE  df asymp.LCL asymp.UCL
## Female / Male      4.94 3.73 Inf      1.12     21.7
##
## Confidence level used: 0.95
## Intervals are back-transformed from the log odds ratio scale
```

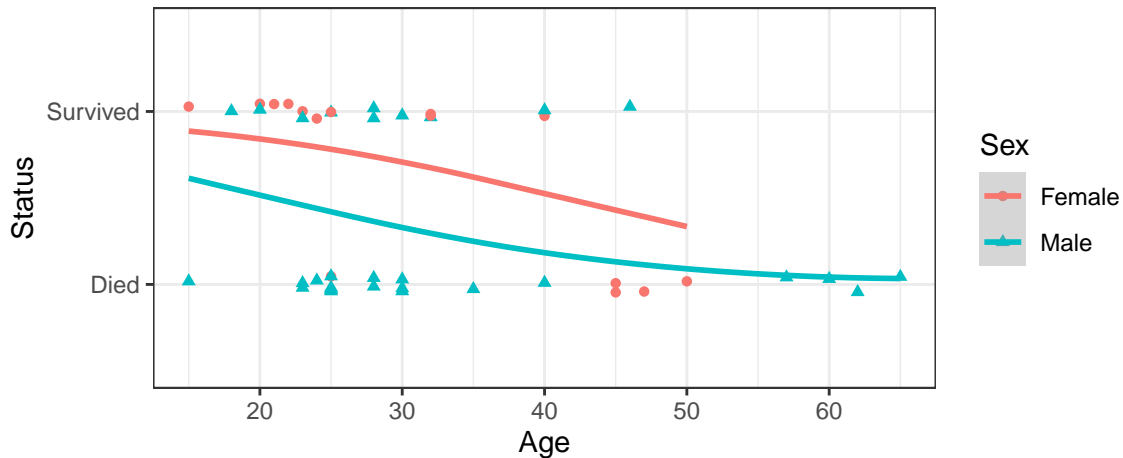
Donner Party - Main effects model

```
et <- emtrends(mM, pairwise ~ Sex, var = "Age")
ct <- confint(et); ct
```

```
## $emtrends
##   Sex    Age.trend      SE df asymp.LCL asymp.UCL
## Female -0.0782 0.0373 Inf   -0.151  -0.00513
## Male   -0.0782 0.0373 Inf   -0.151  -0.00513
##
## Results are given on the == (not the response) scale.
## Confidence level used: 0.95
##
## $contrasts
##   contrast      estimate SE  df asymp.LCL asymp.UCL
## Female - Male      0 0 Inf      0          0
##
## Note: contrasts are still on the == scale
## Results are given on the == (not the response) scale.
## Confidence level used: 0.95
```

Donner Party - Main effects model

```
g + geom_smooth(mapping=aes(y=predict(mM, case2001, type="response")+1))
```



Donner Party - Interaction model

```
summary(mI)

##
## Call:
## glm(formula = Status == "Survived" ~ Sex * Age, family = binomial,
##      data = case2001)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2279  -0.9388  -0.5550   0.7794   1.6998
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  7.24638     3.20517   2.261  0.0238 *
## SexMale      -6.92805     3.39887  -2.038  0.0415 *
## Age          -0.19407     0.08742  -2.220  0.0264 *
## SexMale:Age   0.16160     0.09426   1.714  0.0865 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 61.827  on 44  degrees of freedom
## Residual deviance: 47.346  on 41  degrees of freedom
## AIC: 55.346
##
## Number of Fisher Scoring iterations: 5
```

Donner Party - Interaction model

```
em <- emmeans(mI, pairwise ~ Sex | Age, at = list(Age = c(15,28,65)))
cm <- confint(em); cm$emmeans
```

```
## Age = 15:
## Sex      emmean    SE df asymp.LCL asymp.UCL
## Female  4.335  1.965 Inf     0.484   8.1870
## Male   -0.169  0.661 Inf    -1.465   1.1269
##
## Age = 28:
## Sex      emmean    SE df asymp.LCL asymp.UCL
## Female  1.812  1.043 Inf    -0.231   3.8559
## Male   -0.591  0.400 Inf    -1.375   0.1929
##
## Age = 65:
## Sex      emmean    SE df asymp.LCL asymp.UCL
## Female -5.368  2.707 Inf   -10.674  -0.0633
## Male   -1.793  1.293 Inf    -4.327   0.7420
##
## Results are given on the logit (not the response) scale.
## Confidence level used: 0.95
```

Donner Party - Interaction model

```
cm$contrasts
```

```
## Age = 15:
## contrast      estimate    SE  df asymp.LCL asymp.UCL
## Female - Male      4.50  2.07 Inf      0.440      8.57
##
## Age = 28:
## contrast      estimate    SE  df asymp.LCL asymp.UCL
## Female - Male      2.40  1.12 Inf      0.215      4.59
##
## Age = 65:
## contrast      estimate    SE  df asymp.LCL asymp.UCL
## Female - Male     -3.58  3.00 Inf     -9.455      2.30
##
## Results are given on the log odds ratio (not the response) scale.
## Confidence level used: 0.95
```

Donner Party - Interaction model

```
et <- emtrends(mI, pairwise ~ Sex, var = "Age")
ct <- confint(et); ct
```

```
## $emtrends
## Sex    Age.trend      SE df asymp.LCL asymp.UCL
## Female -0.1941 0.0874 Inf   -0.365   -0.0227
## Male   -0.0325 0.0353 Inf   -0.102    0.0366
##
## Results are given on the == (not the response) scale.
## Confidence level used: 0.95
##
## $contrasts
## contrast      estimate      SE df asymp.LCL asymp.UCL
## Female - Male -0.162 0.0943 Inf   -0.346    0.0232
##
## Note: contrasts are still on the == scale
## Results are given on the == (not the response) scale.
## Confidence level used: 0.95
```

Donner Party - Interaction model

```
g + geom_smooth(mapping=aes(y=predict(mI, case2001, type="response")+1))
```

