

FMPH 222 - Final Project

Kevin Nguyen

March, 2023

Libraries Used

```
library(ggplot2)
library(car)
library(tableone)
library(dplyr)
library(logistf)
library(generalhoslem)
library(ggfortify)
library(ggpubr)
```

To:Do

```
# Maybe parse the Histology variables differently.

# 0.1 pvalue cut off

# Ignore Treatment
# Ignore other IRAEs

# --- Presentation ---
# Make sure to mention this is an explanatory analysis
```

Data Cleaning

```
# dt <- read.csv("tumor_data.csv")
dt <- read.csv("tumor_data.csv")

# --- Column form changes ---
colnames(dt)[5] <- "Ethnicity"
colnames(dt)[2] <- "Adverse.Events"

# Change to Numeric
dt$Age <- as.numeric(dt$Age) # Row 1 is an NA value
dt$Age[1] <- 90

dt$PFS..months. <- as.numeric(dt$PFS..months.)

# Change to Factor
dt$GENDER <- as.factor(dt$GENDER)
dt$Ethnicity <- as.factor(dt$Ethnicity)
```

```

dt$Blood.type. <- as.factor(dt$Blood.type.)
dt$Histology <- as.factor(dt$Histology)
dt$Adverse.Events <- as.factor(dt$Adverse.Events)

# Change Histology to 3 variables only

dt$Histology2 <-
  ifelse(dt$Histology == "Lung" | dt$Histology == "Breast",
        dt$Histology, "Other")

dt$Histology2 <- as.factor(dt$Histology2)

levels(dt$Histology2) <- c("Breast", "Lung", "Other")

# Change Histology to how the paper did it
list_of_Histologies = as.vector(unique(dt$Histology))
list_of_Histologies = list_of_Histologies[!list_of_Histologies %in% c("Skin", "Lung", "Breast")]

dt$Histology3 = dt$Histology

for (i in list_of_Histologies){
  dt$Histology3 = gsub(i, "Other", dt$Histology3)
}

levels(dt$Histology3) <- c("Lung", "Breast", "Skin", "Other")
dt$Histology3 <- as.factor(dt$Histology3)

# Change O blood vs type A/B/AB blood
dt <- dt %>%
  mutate(
    # Create Categories
    Blood.type.cat = dplyr::case_when(
      Blood.type. %in% c("A-", "A+", "AB+", "B-", "B+") ~ "A/B/AB",
      Blood.type. %in% c("O-", "O+") ~ "O"
    )
  )

dt$Blood.type.cat <- as.factor(dt$Blood.type.cat)

# Change Ethnicity Column

dt <- dt %>%
  mutate(
    # Create Categories
    Ethnicity.2 = dplyr::case_when(
      Ethnicity %in% c("Asian", "Black", "Hispanic") ~ "Other",
      Ethnicity %in% c("White") ~ "White"
    )
  )

dt$Ethnicity.2 <- as.factor(dt$Ethnicity.2)

# Change Age Column

```

```

# Categorize Age
dt <- dt %>%
  mutate(
    # Create Categories
    Age.cat = dplyr::case_when(
      Age >= 28 & Age <= 65 ~ "28-65",
      Age > 65 ~ "65-90"
    )
  )

dt$Age.cat <- factor(dt$Age.cat, levels = c("28-65", "65-90"))

```

Summary Statistics

```

# Variables of Interest
myVars <- c("Age.cat", "Ethnicity.2", "GENDER",
            "Histology3", "Adverse.Events")

# Table Creation
table_3 <- CreateTableOne(vars = myVars, strata = "Blood.type.cat",
                           data = dt)
print(table_3, formatOptions = list(big.mark = ","))

```

```

##                               Stratified by Blood.type.cat
##                               A/B/AB      0      p      test
##  n                               73      47
##  Age.cat = 65-90 (%)             22 (30.1) 14 (29.8) 1.000
##  Ethnicity.2 = White (%)         54 (74.0) 31 (66.0) 0.461
##  GENDER = Male (%)               43 (58.9) 25 (53.2) 0.669
##  Histology3 (%)                                0.488
##    Breast                           2 ( 2.7)  4 ( 8.5)
##    Lung                             16 (21.9)  8 (17.0)
##    Other                            43 (58.9) 26 (55.3)
##    Skin                             12 (16.4)  9 (19.1)
##  Adverse.Events = 1 (%)          24 (32.9) 24 (51.1) 0.073

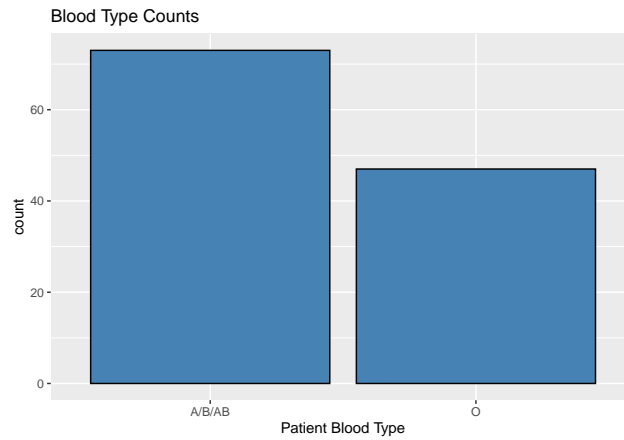
```

Plots

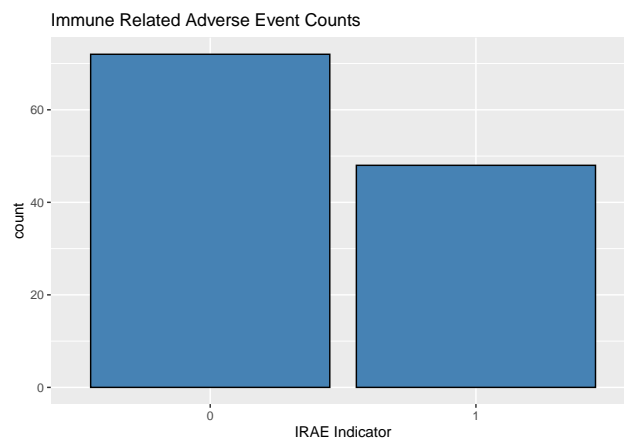
```

# Blood Type Histogram
ggplot(dt, aes(x=Blood.type.cat)) +
  geom_bar(fill="steelblue", color="black") +
  labs(x="Patient Blood Type", title="Blood Type Counts")

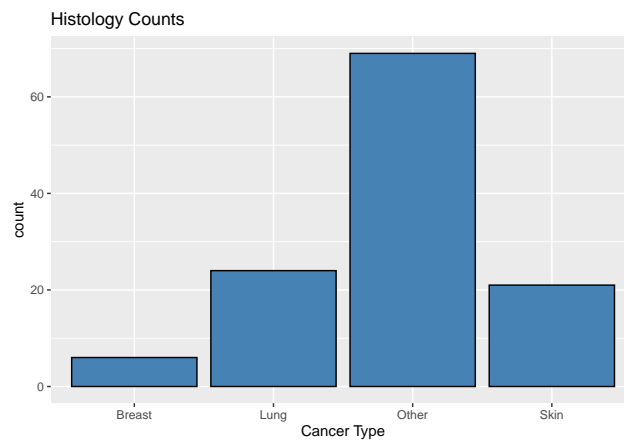
```



```
# Adverse.Events
ggplot(dt,aes(x=Adverse.Events)) +
  geom_bar(fill="steelblue",color="black") +
  labs(x="IRAE Indicator", title="Immune Related Adverse Event Counts")
```



```
# Histology
ggplot(dt,aes(x=Histology3)) +
  geom_bar(fill="steelblue",color="black") +
  labs(x="Cancer Type", title="Histology Counts")
```



```
## could be causing major problems
```

Pre-Analysis

```
# --- Old Analysis ---
```

```
m1 <- glm(Adverse.Events ~ Blood.type. + Ethnicity + Age, family = binomial,  
          data = dt )  
summary(m1)
```

```
##  
## Call:  
## glm(formula = Adverse.Events ~ Blood.type. + Ethnicity + Age,  
##      family = binomial, data = dt)  
##  
## Deviance Residuals:  
##      Min       1Q   Median       3Q      Max  
## -1.4384  -0.9987  -0.6886   1.1211   1.8876  
##  
## Coefficients:  
##              Estimate Std. Error z value Pr(>|z|)  
## (Intercept)   -1.06995    1.50323  -0.712   0.4766  
## Blood.type.A+  -1.73253    0.70752  -2.449   0.0143 *  
## Blood.type.AB+ -1.08570    1.06224  -1.022   0.3067  
## Blood.type.B-  -0.98075    1.37152  -0.715   0.4746  
## Blood.type.B+  -1.15779    0.81033  -1.429   0.1531  
## Blood.type.O-  -0.51953    1.00693  -0.516   0.6059  
## Blood.type.O+  -0.45983    0.66380  -0.693   0.4885  
## EthnicityBlack  0.99812    1.38130   0.723   0.4699  
## EthnicityHispanic 0.85546    1.29283   0.662   0.5082  
## EthnicityWhite   0.70278    1.21069   0.580   0.5616  
## Age             0.01396    0.01599   0.873   0.3826  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## (Dispersion parameter for binomial family taken to be 1)  
##  
##      Null deviance: 161.52  on 119  degrees of freedom  
## Residual deviance: 149.79  on 109  degrees of freedom  
## AIC: 171.79  
##  
## Number of Fisher Scoring iterations: 4
```

```
m2 <- glm(Adverse.Events ~ Blood.type. + Ethnicity + Age + Histology2,  
          family = binomial , data = dt )  
summary(m2)
```

```
##  
## Call:  
## glm(formula = Adverse.Events ~ Blood.type. + Ethnicity + Age +  
##      Histology2, family = binomial, data = dt)  
##  
## Deviance Residuals:  
##      Min       1Q   Median       3Q      Max
```

```
## -1.470 -1.042 -0.673 1.147 1.903
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -1.06839    1.67584  -0.638   0.5238
## Blood.type.A+ -1.78748    0.71480  -2.501   0.0124 *
## Blood.type.AB+ -1.06506    1.07035  -0.995   0.3197
## Blood.type.B-  -1.10510    1.38983  -0.795   0.4265
## Blood.type.B+  -1.15051    0.81024  -1.420   0.1556
## Blood.type.O-  -0.46451    1.02375  -0.454   0.6500
## Blood.type.O+  -0.44741    0.66685  -0.671   0.5023
## EthnicityBlack  0.91241    1.38534   0.659   0.5101
## EthnicityHispanic 0.82944    1.30138   0.637   0.5239
## EthnicityWhite  0.65377    1.21552   0.538   0.5907
## Age            0.01012    0.01687   0.600   0.5484
## Histology2Lung  0.60413    1.07334   0.563   0.5735
## Histology2Other 0.20626    0.97841   0.211   0.8330
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 161.52  on 119  degrees of freedom
## Residual deviance: 149.13  on 107  degrees of freedom
## AIC: 175.13
##
## Number of Fisher Scoring iterations: 4
anova(m1,m2)
```

Resid. Df	Resid. Dev	Df	Deviance
109	149.7938	NA	NA
107	149.1257	2	0.6681261

Blood type A+ is significant (THE ONLY ONE too)

Initial Analysis

```
# Main variable of interest:
# Y = Adverse.Events
# X_n = Blood.type.cat, Histology3
# Confounders: Ethnicity, Age, Gender

# DO NOT INCLUDE variables:
# - Treatment..Clean.
# - PFS..months.

m1 <- glm(Adverse.Events ~ Blood.type.cat + Ethnicity.2 + Age + Histology3 +
  GENDER, family = binomial, data = dt)

summary(m1)

##
```

```
## Call:
## glm(formula = Adverse.Events ~ Blood.type.cat + Ethnicity.2 +
##      Age + Histology3 + GENDER, family = binomial, data = dt)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.4638  -1.0185  -0.7623   1.2504   1.7684
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -1.92667     1.22995  -1.566   0.117
## Blood.type.cat0  0.77276     0.39942   1.935   0.053 .
## Ethnicity.2White 0.07563     0.43094   0.175   0.861
## Age            0.01331     0.01644   0.809   0.418
## Histology3Lung   0.83844     1.03340   0.811   0.417
## Histology3Other  0.73765     0.97150   0.759   0.448
## Histology3Skin   0.93112     1.06295   0.876   0.381
## GENDERMale      -0.69448     0.42269  -1.643   0.100
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 161.52  on 119  degrees of freedom
## Residual deviance: 153.43  on 112  degrees of freedom
## AIC: 169.43
##
## Number of Fisher Scoring iterations: 4
(H_L_m1 = generalhoslem::logitgof(dt$Adverse.Events,m1$fitted.values))

##
## Hosmer and Lemeshow test (binary model)
##
## data:  dt$Adverse.Events, m1$fitted.values
## X-squared = 3.4314, df = 8, p-value = 0.9044
```

Linear Model Check

```
# reduced - treat histology as linear term
dt$Histology.lin[dt$Histology3=="Breast"] <- 1
dt$Histology.lin[dt$Histology3=="Lung"] <- 2
dt$Histology.lin[dt$Histology3=="Skin"] <- 3
dt$Histology.lin[dt$Histology3=="Other"] <- 4

m1.lin <- glm(Adverse.Events ~ Blood.type.cat + Ethnicity.2 + Age + Histology.lin+
  GENDER, family = binomial, data = dt)

anova(m1.lin,m1,test="LRT")
```

Resid. Df	Resid. Dev	Df	Deviance	Pr(>Chi)
114	154.2508	NA	NA	NA
112	153.4337	2	0.8171169	0.6646076

```
# p-value 0.39 -> linear model is better
```

```
summary(m1.lin)
```

```
##
## Call:
## glm(formula = Adverse.Events ~ Blood.type.cat + Ethnicity.2 +
##      Age + Histology.lin + GENDER, family = binomial, data = dt)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.4037  -1.0051  -0.7766   1.2333   1.7676
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -1.44693    1.14806  -1.260   0.2076
## Blood.type.cat0  0.73290    0.39158   1.872   0.0613 .
## Ethnicity.2White 0.07014    0.42880   0.164   0.8701
## Age           0.01513    0.01536   0.985   0.3245
## Histology.lin  0.04320    0.20635   0.209   0.8342
## GENDERMale    -0.60907    0.39804  -1.530   0.1260
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 161.52  on 119  degrees of freedom
## Residual deviance: 154.25  on 114  degrees of freedom
## AIC: 166.25
##
## Number of Fisher Scoring iterations: 4
```

```
# blood type is no longer significant
# stick with m1
```

Backwards Selection

```
# backward() function
m1.back <- logistf(Adverse.Events ~ Ethnicity.2 + Age + GENDER,
                  family = binomial(link="logit"), data = dt)
mod.m1.back <- backward(m1.back,slstay=0.1)
```

```
## Step 0 : starting model
## Step 1 : removed Ethnicity.2 GENDER (P= 1 )
```

```
summary(mod.m1.back)
```

```
## logistf(formula = Adverse.Events ~ Age + GENDER, data = dt, family = binomial(link = "logit"))
##
## Model fitted by Penalized ML
## Coefficients:
##              coef    se(coef) lower 0.95 upper 0.95    Chisq      p
## (Intercept) -0.99012522 0.89515793 -2.79745993 0.74491528 1.241711 0.2651416
## Age          0.01572507 0.01462231 -0.01262653 0.04521729 1.172459 0.2788965
## GENDERMale  -0.59350925 0.37490212 -1.33671698 0.13842755 2.524117 0.1121175
```



```
##               method
## (Intercept)      2
## Age              2
## GENDERMale       2
##
## Method: 1-Wald, 2-Profile penalized log-likelihood, 3-None
##
## Likelihood ratio test=3.603624 on 2 df, p=0.1649996, n=120
## Wald test = 7.705891 on 2 df, p = 0.02121715
# add main variables
m2 <- glm(Adverse.Events ~ Age + GENDER + Blood.type.cat + Histology3,
          family = binomial(link="logit"), data=dt)
summary(m2)

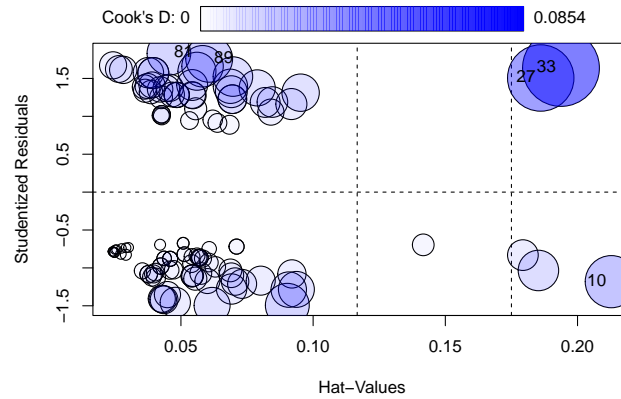
##
## Call:
## glm(formula = Adverse.Events ~ Age + GENDER + Blood.type.cat +
##      Histology3, family = binomial(link = "logit"), data = dt)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.4531  -1.0100  -0.7663   1.2552   1.7801
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -1.88454    1.20494  -1.564   0.1178
## Age             0.01363    0.01637   0.833   0.4051
## GENDERMale     -0.68984    0.42169  -1.636   0.1019
## Blood.type.cat 0.76607    0.39740   1.928   0.0539 .
## Histology3Lung  0.83350    1.03226   0.807   0.4194
## Histology3Other 0.72893    0.96915   0.752   0.4520
## Histology3Skin  0.92430    1.06147   0.871   0.3839
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 161.52  on 119  degrees of freedom
## Residual deviance: 153.46  on 113  degrees of freedom
## AIC: 167.46
##
## Number of Fisher Scoring iterations: 4
```

Comparing The Two Models

```
anova(m1, m2, test = "LRT")
```

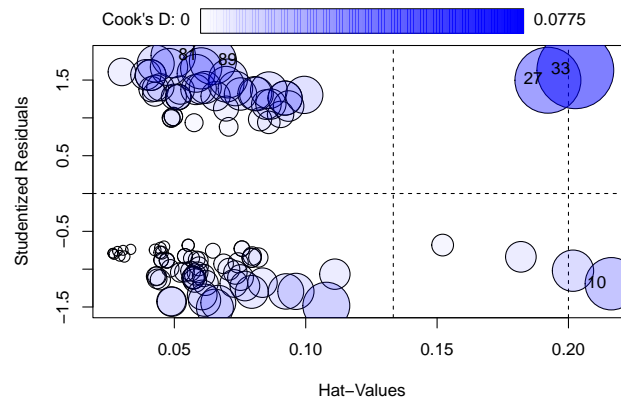
Resid. Df	Resid. Dev	Df	Deviance	Pr(>Chi)
112	153.4337	NA	NA	NA
113	153.4646	-1	-0.0308687	0.8605337

```
influencePlot(m2)
```



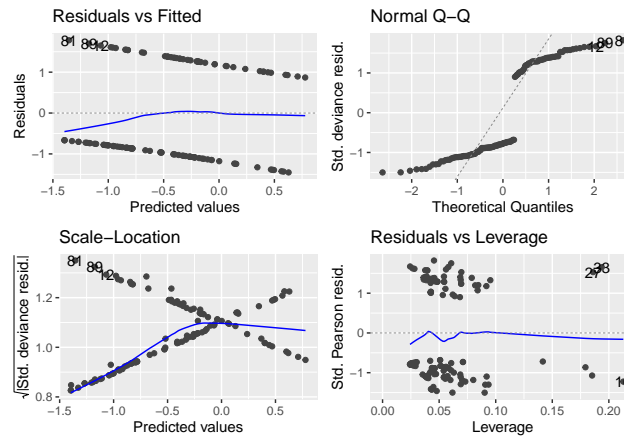
	StudRes	Hat	CookD
10	-1.181731	0.2127751	0.0394053
27	1.505868	0.1860824	0.0638358
33	1.637955	0.1937406	0.0853923
81	1.831165	0.0454517	0.0276204
89	1.758988	0.0606842	0.0317539

```
influencePlot(m1)
```



	StudRes	Hat	CookD
10	-1.192633	0.2163074	0.0360476
27	1.497613	0.1921835	0.0571287
33	1.629499	0.2026208	0.0775135
81	1.823276	0.0495818	0.0259101
89	1.751030	0.0646816	0.0291471

```
autoplot(m2)
```



Hosmer Lemeshow Test

```
(H_L = generalhoslem::logitgof(dt$Adverse.Events,m2$fitted.values))
```

```
##
## Hosmer and Lemeshow test (binary model)
##
## data: dt$Adverse.Events, m2$fitted.values
## X-squared = 2.4051, df = 8, p-value = 0.966
```

- p-value is high
- model is good fit (model is a good fit as the insignificant p value indicates that there are no issues with the model)

Aggregating Data

```
new_dt = dt[,c("Adverse.Events", "Ethnicity.2", "GENDER",
               "Blood.type.cat", "Histology3", "Age.cat")]

new_dt$Histology3 <- relevel(new_dt$Histology3, ref = "Other")

new_dt[] = lapply(new_dt, as.character)
new_dt$Adverse.Events = as.numeric(as.character(new_dt$Adverse.Events))

agg_data = aggregate(Adverse.Events ~ Ethnicity.2 + GENDER +
                     Blood.type.cat + Histology3 + Age.cat,
                     data = new_dt,
                     FUN = sum)
agg_data = cbind(agg_data,
                 aggregate(Adverse.Events ~ Ethnicity.2 + GENDER +
                           Blood.type.cat + Histology3 + Age.cat,
                           data = new_dt,
                           FUN = length))

names(agg_data)[12] = "Adverse.Events_tot"
agg_data = agg_data[, !duplicated(colnames(agg_data))]
head(agg_data)
```

Ethnicity.2	GENDER	Blood.type.cat	Histology3	Age.cat	Adverse.Events	Adverse.Events_tot
Other	Female	A/B/AB	Breast	28-65	0	1
White	Female	A/B/AB	Breast	28-65	0	1
Other	Female	O	Breast	28-65	0	1
White	Female	O	Breast	28-65	2	2
Other	Female	A/B/AB	Lung	28-65	2	2
White	Female	A/B/AB	Lung	28-65	1	6

```
agg_data$Histology3 <- as.factor(x = agg_data$Histology3)
agg_data$Histology3 <- relevel(agg_data$Histology3, ref = "Other")
```

Aggregated Model

```
agg_model = glm(Adverse.Events/Adverse.Events_tot ~ Ethnicity.2 + Age.cat + GENDER +
                Blood.type.cat + Histology3,
                family = binomial(link = "logit"),
                data = agg_data,
                weight = Adverse.Events_tot)
```

```
summary(agg_model)
```

```
##
## Call:
## glm(formula = Adverse.Events/Adverse.Events_tot ~ Ethnicity.2 +
##      Age.cat + GENDER + Blood.type.cat + Histology3, family = binomial(link = "logit"),
##      data = agg_data, weights = Adverse.Events_tot)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.0183  -1.0142  -0.1285   0.9840   1.9787
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.46825    0.48300  -0.969   0.3323
## Ethnicity.2White  0.02881    0.44383   0.065   0.9482
## Age.cat65-90     0.33600    0.45330   0.741   0.4586
## GENDERMale      -0.72152    0.42759  -1.687   0.0915 .
## Blood.type.catO  0.79629    0.39800   2.001   0.0454 *
## Histology3Breast -0.86442    0.95866  -0.902   0.3672
## Histology3Lung    0.12401    0.50756   0.244   0.8070
## Histology3Skin    0.18235    0.55530   0.328   0.7426
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 60.889  on 39  degrees of freedom
## Residual deviance: 52.912  on 32  degrees of freedom
## AIC: 100.14
##
## Number of Fisher Scoring iterations: 4
```

Backwards Selection

```
# using backward() function
agg_back_b = logistf(Adverse.Events/Adverse.Events_tot ~ Ethnicity.2 + Age.cat + GENDER,
                     family = binomial(link="logit"),
                     data = agg_data,
                     weights = Adverse.Events_tot)
agg_reduced_b = backward(agg_back_b, slstay = 0.1)

## Step 0 : starting model
## Step 1 : removed Age.cat (P= 0.362972 )
## Step 2 : removed GENDER (P= 0.2026073 )

summary(agg_reduced_b)

## logistf(formula = Adverse.Events/Adverse.Events_tot ~ Ethnicity.2,
##         data = agg_data, pl = FALSE, weights = Adverse.Events_tot,
##         family = binomial(link = "logit"))
##
## Model fitted by Penalized ML
## Coefficients:
##               coef se(coef) lower 0.95 upper 0.95      Chisq
## (Intercept)   -1.3350011 0.4103913  -2.139353  -0.5306488 10.581978
## Ethnicity.2White -0.8752107 0.5467819  -1.946883   0.1964621  2.562106
##               p method
## (Intercept)    0.001141954      1
## Ethnicity.2White 0.109452715      1
##
## Method: 1-Wald, 2-Profile penalized log-likelihood, 3-None
##
## Likelihood ratio test=2.490393 on 1 df, p=0.1145432, n=120
## Wald test = 48.00105 on 1 df, p = 4.259926e-12

# adding in variables of main interest
agg_final = glm(Adverse.Events/Adverse.Events_tot ~ Ethnicity.2 + Blood.type.cat + Histology3,
                family = binomial(link="logit"),
                data = agg_data,
                weight = Adverse.Events_tot)
summary(agg_final)

##
## Call:
## glm(formula = Adverse.Events/Adverse.Events_tot ~ Ethnicity.2 +
##     Blood.type.cat + Histology3, family = binomial(link = "logit"),
##     data = agg_data, weights = Adverse.Events_tot)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.0905  -0.9843  -0.2473   1.0689   2.1992
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.80572    0.42781  -1.883   0.0597 .
## Ethnicity.2White  0.06164    0.42102   0.146   0.8836
## Blood.type.cat0   0.81326    0.39234   2.073   0.0382 *
## Histology3Breast -0.49849    0.92325  -0.540   0.5892
```

```
## Histology3Lung    0.31844    0.48845    0.652    0.5144
## Histology3Skin   -0.09038    0.52315   -0.173    0.8628
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 60.889  on 39  degrees of freedom
## Residual deviance: 56.018  on 34  degrees of freedom
## AIC: 99.245
##
## Number of Fisher Scoring iterations: 4
```

Comparing Aggregated Models

```
anova(agg_model, agg_final, test = "LRT")
```

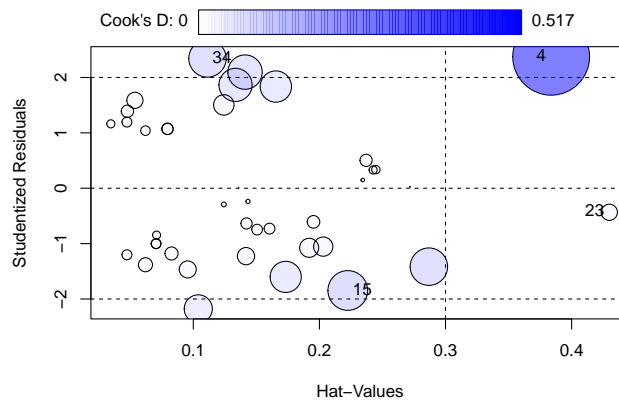
Resid. Df	Resid. Dev	Df	Deviance	Pr(>Chi)
32	52.91191	NA	NA	NA
34	56.01756	-2	-3.105646	0.2116497

Checking For Model Significance And Influential Points

```
pchisq(summary(agg_final)$deviance,
        df = summary(agg_final)$df[2],
        lower.tail = F)
```

```
## [1] 0.01010138
```

```
influencePlot(agg_final)
```

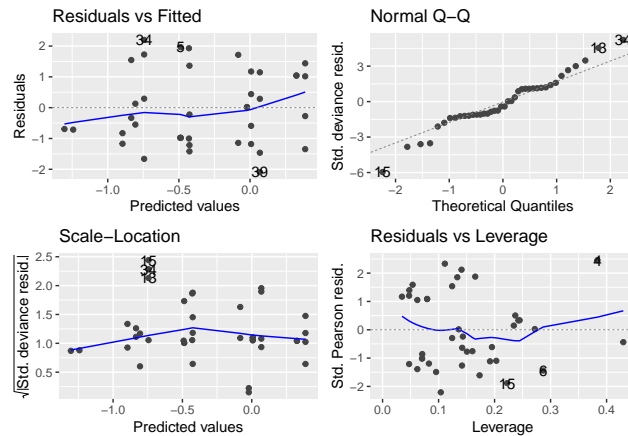


	StudRes	Hat	CookD
4	2.3740184	0.3837633	0.5174754
15	-1.8462425	0.2225077	0.1386077
23	-0.4342275	0.4298657	0.0232038
34	2.3432763	0.1112538	0.1227610

```
Std.res = rstandard(agg_final)
Std.res[Std.res == max(Std.res)]
```

```
##          4
## 2.457829
```

```
autoplot(agg_final)
```



```
CD = cooks.distance((agg_final))
CD[CD==max(CD)]
```

```
##          4
## 0.5174754
```

Determining Outliers

```
outlierTest(agg_final)
```

```
## No Studentized residuals with Bonferroni p < 0.05
## Largest |rstudent|:
##   rstudent unadjusted p-value Bonferroni p
## 4 2.374018      0.017596      0.70383
```

```
(outliers <- match(4, row.names(agg_data)))
```

```
## [1] 4
```

```
agg_data[outliers,]
```

	Ethnicity.2	GENDER	Blood.type.cat	Histology3	Age.cat	Adverse.Events	Adverse.Events_tot
4	White	Female	O	Breast	28-65	2	2

Removing Outliers And Recreating Model

```
agg_data = agg_data[-4,]
```

```
agg_final_adjusted = glm(Adverse.Events/Adverse.Events_tot ~ Ethnicity.2 + Blood.type.cat +
  Histology3,
  family = binomial(link="logit"),
  data = agg_data,
```

```

weight = Adverse.Events_tot)
summary(agg_final_adjusted)

##
## Call:
## glm(formula = Adverse.Events/Adverse.Events_tot ~ Ethnicity.2 +
##      Blood.type.cat + Histology3, family = binomial(link = "logit"),
##      data = agg_data, weights = Adverse.Events_tot)
##
## Deviance Residuals:
##      Min        1Q      Median        3Q        Max
## -2.02865  -1.02405  -0.00025   1.03971   2.20286
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.69151    0.43147  -1.603   0.1090
## Ethnicity.2White -0.05596    0.42896  -0.130   0.8962
## Blood.type.cat0  0.73293    0.39925   1.836   0.0664 .
## Histology3Breast -17.23422  1934.16698  -0.009   0.9929
## Histology3Lung    0.31950    0.48688   0.656   0.5117
## Histology3Skin   -0.08322    0.52151  -0.160   0.8732
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 57.173  on 38  degrees of freedom
## Residual deviance: 49.245  on 33  degrees of freedom
## AIC: 92.472
##
## Number of Fisher Scoring iterations: 16
pchisq(summary(agg_final_adjusted)$deviance,
        df = summary(agg_final_adjusted)$df[2],
        lower.tail = F)

## [1] 0.03427618

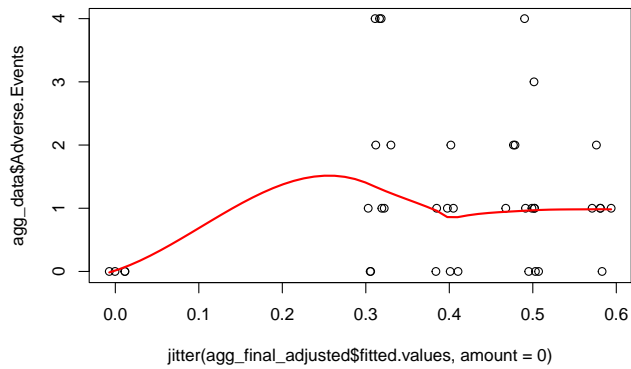
```

Final Model Diagnostics

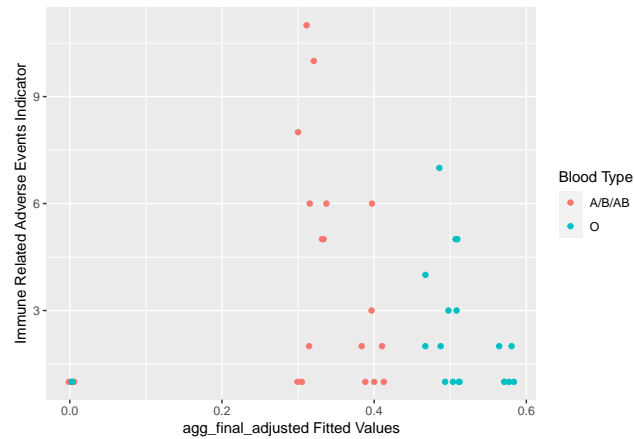
```

# PLOT AFTER AGGREGATING DATA
scatter.smooth(jitter(agg_final_adjusted$fitted.values, amount=0),
               agg_data$Adverse.Events,
               lpars = list(col = "red", lwd = 2))

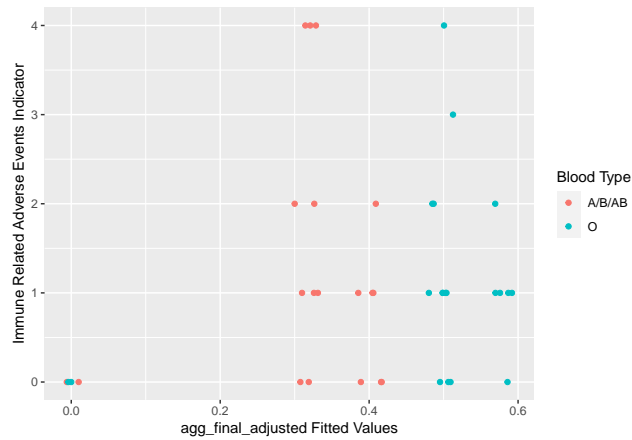
```

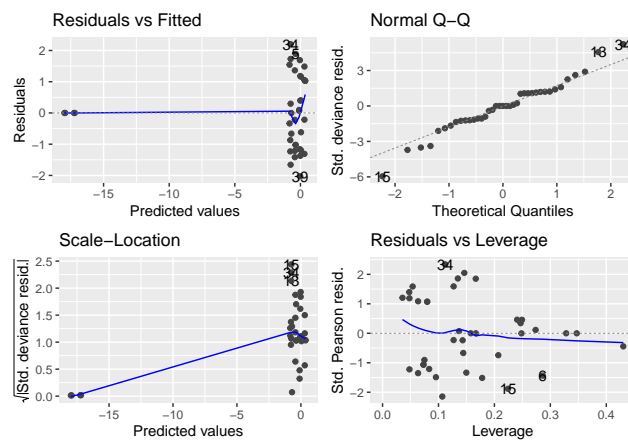
```
# Adverse.Events_tot
ggplot(agg_data,aes(x=jitter(agg_final_adjusted$fitted.values,amount=0),
                        y=agg_data$Adverse.Events_tot)) +
  geom_point(aes(color=Blood.type.cat))+
  labs(y="Immune Related Adverse Events Indicator",
       x="agg_final_adjusted Fitted Values")+
  guides(color = guide_legend(title = "Blood Type"))
```



```
# Adverse.Events
ggplot(agg_data,aes(x=jitter(agg_final_adjusted$fitted.values,amount=0),
                        y=agg_data$Adverse.Events)) +
  geom_point(aes(color=Blood.type.cat))+
  labs(y="Immune Related Adverse Events Indicator",
       x="agg_final_adjusted Fitted Values")+
  guides(color = guide_legend(title = "Blood Type"))
```



```
autoplot(agg_final_adjusted)
```



Interpretation

```
# comparing proportions
```

```
# Blood Type
```

```
bt_p <- ggplot(dt, aes(fill=Adverse.Events, x=Blood.type.cat)) +  
  geom_bar(position="fill")+scale_y_continuous(labels = scales::percent)+labs(x="Blood Type",y="Percent")
```

```
# Histology
```

```
hist_p <- ggplot(dt, aes(fill=Adverse.Events, x=Histology3)) +  
  geom_bar(position="fill")+scale_y_continuous(labels = scales::percent)+labs(x="Histology",y="Percent")
```

```
ggarrange(bt_p,hist_p,ncol=2,nrow=1,labels=c("A","B"))
```

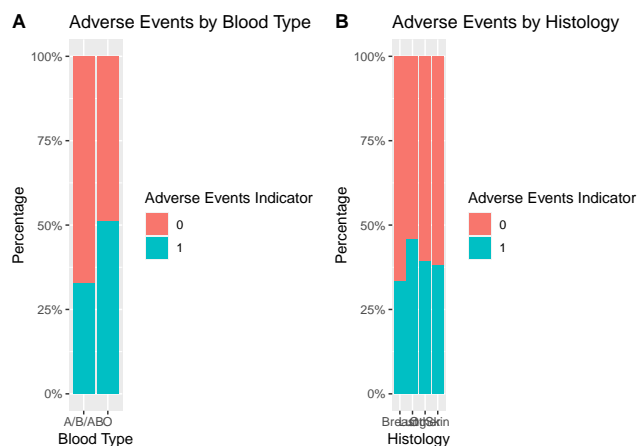


Table Of Coefficients, P-Values, Standard Errors, And Confidence Intervals

```
final_table = cbind(exp(agg_final_adjusted$coefficients),
                     summary(agg_final_adjusted)$coefficients[,4],
                     confint(agg_final_adjusted, level = 0.9),
                     sqrt(diag(vcov(agg_final_adjusted))))
colnames(final_table) = c("Estimate", "Pr(>|z|)", "5%", "95%", "Std. Error")
print(final_table)
```

##	Estimate	Pr(> z)	5%	95%	Std. Error
## (Intercept)	5.008199e-01	0.10900103	-1.42108944	6.658973e-03	0.4314650
## Ethnicity.2White	9.455747e-01	0.89620225	-0.75834494	6.586172e-01	0.4289609
## Blood.type.cat0	2.081179e+00	0.06638868	0.07935668	1.396369e+00	0.3992474
## Histology3Breast	3.275452e-08	0.99289061	NA	1.529957e+02	1934.1669754
## Histology3Lung	1.376436e+00	0.51168543	-0.48775231	1.122850e+00	0.4868786
## Histology3Skin	9.201482e-01	0.87321496	-0.96358075	7.646407e-01	0.5215098

Interactions

```
int_model = glm(Adverse.Events ~ Ethnicity.2 * Age.cat * GENDER *
                 Blood.type.cat * Histology3,
                 family = binomial(link = "logit"),
                 data = dt)
summary(int_model)
```

```
##
## Call:
## glm(formula = Adverse.Events ~ Ethnicity.2 * Age.cat * GENDER *
##      Blood.type.cat * Histology3, family = binomial(link = "logit"),
##      data = dt)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.79412  -0.75853  -0.00013   0.66805   2.14597
##
## Coefficients: (24 not defined because of singularities)
##
##              Estimate
## (Intercept)    -1.857e+01
## Ethnicity.2White -4.943e-08
```

## Age.cat65-90	-1.617e+02
## GENDERMale	1.996e-07
## Blood.type.cat0	-6.316e-11
## Histology3Lung	3.713e+01
## Histology30ther	1.718e+01
## Histology3Skin	-1.864e-07
## Ethnicity.2White:Age.cat65-90	1.069e+02
## Ethnicity.2White:GENDERMale	-1.966e+01
## Age.cat65-90:GENDERMale	7.244e+01
## Ethnicity.2White:Blood.type.cat0	3.713e+01
## Age.cat65-90:Blood.type.cat0	8.838e+01
## GENDERMale:Blood.type.cat0	3.713e+01
## Ethnicity.2White:Histology3Lung	-2.018e+01
## Ethnicity.2White:Histology30ther	2.079e+00
## Ethnicity.2White:Histology3Skin	3.713e+01
## Age.cat65-90:Histology3Lung	9.248e+01
## Age.cat65-90:Histology30ther	5.541e+01
## Age.cat65-90:Histology3Skin	7.259e+01
## GENDERMale:Histology3Lung	-3.713e+01
## GENDERMale:Histology30ther	6.931e-01
## GENDERMale:Histology3Skin	NA
## Blood.type.cat0:Histology3Lung	-1.072e-07
## Blood.type.cat0:Histology30ther	6.931e-01
## Blood.type.cat0:Histology3Skin	1.514e-07
## Ethnicity.2White:Age.cat65-90:GENDERMale	-7.149e+01
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0	-7.080e+01
## Ethnicity.2White:GENDERMale:Blood.type.cat0	8.109e-01
## Age.cat65-90:GENDERMale:Blood.type.cat0	-3.640e+01
## Ethnicity.2White:Age.cat65-90:Histology3Lung	-5.472e+01
## Ethnicity.2White:Age.cat65-90:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:Histology3Skin	NA
## Ethnicity.2White:GENDERMale:Histology3Lung	7.697e+01
## Ethnicity.2White:GENDERMale:Histology30ther	1.608e+01
## Ethnicity.2White:GENDERMale:Histology3Skin	NA
## Age.cat65-90:GENDERMale:Histology3Lung	-3.247e+00
## Age.cat65-90:GENDERMale:Histology30ther	NA
## Age.cat65-90:GENDERMale:Histology3Skin	NA
## Ethnicity.2White:Blood.type.cat0:Histology3Lung	-1.696e+01
## Ethnicity.2White:Blood.type.cat0:Histology30ther	-3.823e+01
## Ethnicity.2White:Blood.type.cat0:Histology3Skin	-7.398e+01
## Age.cat65-90:Blood.type.cat0:Histology3Lung	-1.919e+01
## Age.cat65-90:Blood.type.cat0:Histology30ther	NA
## Age.cat65-90:Blood.type.cat0:Histology3Skin	NA
## GENDERMale:Blood.type.cat0:Histology3Lung	-9.525e+01
## GENDERMale:Blood.type.cat0:Histology30ther	-3.673e+01
## GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Lung	NA

## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung	9.440e+01
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin	NA
##	Std. Error
## (Intercept)	6.523e+03
## Ethnicity.2White	9.224e+03
## Age.cat65-90	3.048e+04
## GENDERMale	7.989e+03
## Blood.type.cat0	9.224e+03
## Histology3Lung	7.989e+03
## Histology30ther	6.523e+03
## Histology3Skin	9.224e+03
## Ethnicity.2White:Age.cat65-90	2.554e+04
## Ethnicity.2White:GENDERMale	1.031e+04
## Age.cat65-90:GENDERMale	1.191e+04
## Ethnicity.2White:Blood.type.cat0	1.220e+04
## Age.cat65-90:Blood.type.cat0	2.291e+04
## GENDERMale:Blood.type.cat0	1.130e+04
## Ethnicity.2White:Histology3Lung	1.031e+04
## Ethnicity.2White:Histology30ther	9.224e+03
## Ethnicity.2White:Histology3Skin	1.305e+04
## Age.cat65-90:Histology3Lung	2.063e+04
## Age.cat65-90:Histology30ther	1.031e+04
## Age.cat65-90:Histology3Skin	1.098e+04
## GENDERMale:Histology3Lung	1.130e+04
## GENDERMale:Histology30ther	7.989e+03
## GENDERMale:Histology3Skin	NA
## Blood.type.cat0:Histology3Lung	1.220e+04
## Blood.type.cat0:Histology30ther	9.224e+03
## Blood.type.cat0:Histology3Skin	1.305e+04
## Ethnicity.2White:Age.cat65-90:GENDERMale	1.191e+04
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0	1.993e+04
## Ethnicity.2White:GENDERMale:Blood.type.cat0	2.833e+00
## Age.cat65-90:GENDERMale:Blood.type.cat0	7.532e+03
## Ethnicity.2White:Age.cat65-90:Histology3Lung	1.459e+04
## Ethnicity.2White:Age.cat65-90:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:Histology3Skin	NA
## Ethnicity.2White:GENDERMale:Histology3Lung	1.459e+04
## Ethnicity.2White:GENDERMale:Histology30ther	1.031e+04
## Ethnicity.2White:GENDERMale:Histology3Skin	NA
## Age.cat65-90:GENDERMale:Histology3Lung	7.989e+03
## Age.cat65-90:GENDERMale:Histology30ther	NA
## Age.cat65-90:GENDERMale:Histology3Skin	NA
## Ethnicity.2White:Blood.type.cat0:Histology3Lung	1.598e+04
## Ethnicity.2White:Blood.type.cat0:Histology30ther	1.220e+04
## Ethnicity.2White:Blood.type.cat0:Histology3Skin	1.384e+04
## Age.cat65-90:Blood.type.cat0:Histology3Lung	1.031e+04
## Age.cat65-90:Blood.type.cat0:Histology30ther	NA
## Age.cat65-90:Blood.type.cat0:Histology3Skin	NA

## GENDERMale:Blood.type.cat0:Histology3Lung	1.598e+04
## GENDERMale:Blood.type.cat0:Histology30ther	1.130e+04
## GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung	1.506e+04
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin	NA
##	z value
## (Intercept)	-0.003
## Ethnicity.2White	0.000
## Age.cat65-90	-0.005
## GENDERMale	0.000
## Blood.type.cat0	0.000
## Histology3Lung	0.005
## Histology30ther	0.003
## Histology3Skin	0.000
## Ethnicity.2White:Age.cat65-90	0.004
## Ethnicity.2White:GENDERMale	-0.002
## Age.cat65-90:GENDERMale	0.006
## Ethnicity.2White:Blood.type.cat0	0.003
## Age.cat65-90:Blood.type.cat0	0.004
## GENDERMale:Blood.type.cat0	0.003
## Ethnicity.2White:Histology3Lung	-0.002
## Ethnicity.2White:Histology30ther	0.000
## Ethnicity.2White:Histology3Skin	0.003
## Age.cat65-90:Histology3Lung	0.004
## Age.cat65-90:Histology30ther	0.005
## Age.cat65-90:Histology3Skin	0.007
## GENDERMale:Histology3Lung	-0.003
## GENDERMale:Histology30ther	0.000
## GENDERMale:Histology3Skin	NA
## Blood.type.cat0:Histology3Lung	0.000
## Blood.type.cat0:Histology30ther	0.000
## Blood.type.cat0:Histology3Skin	0.000
## Ethnicity.2White:Age.cat65-90:GENDERMale	-0.006
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0	-0.004
## Ethnicity.2White:GENDERMale:Blood.type.cat0	0.286
## Age.cat65-90:GENDERMale:Blood.type.cat0	-0.005
## Ethnicity.2White:Age.cat65-90:Histology3Lung	-0.004
## Ethnicity.2White:Age.cat65-90:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:Histology3Skin	NA
## Ethnicity.2White:GENDERMale:Histology3Lung	0.005

## Ethnicity.2White:GENDERMale:Histology30ther	0.002
## Ethnicity.2White:GENDERMale:Histology3Skin	NA
## Age.cat65-90:GENDERMale:Histology3Lung	0.000
## Age.cat65-90:GENDERMale:Histology30ther	NA
## Age.cat65-90:GENDERMale:Histology3Skin	NA
## Ethnicity.2White:Blood.type.cat0:Histology3Lung	-0.001
## Ethnicity.2White:Blood.type.cat0:Histology30ther	-0.003
## Ethnicity.2White:Blood.type.cat0:Histology3Skin	-0.005
## Age.cat65-90:Blood.type.cat0:Histology3Lung	-0.002
## Age.cat65-90:Blood.type.cat0:Histology30ther	NA
## Age.cat65-90:Blood.type.cat0:Histology3Skin	NA
## GENDERMale:Blood.type.cat0:Histology3Lung	-0.006
## GENDERMale:Blood.type.cat0:Histology30ther	-0.003
## GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung	0.006
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther	NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin	NA
##	Pr(> z)
## (Intercept)	0.998
## Ethnicity.2White	1.000
## Age.cat65-90	0.996
## GENDERMale	1.000
## Blood.type.cat0	1.000
## Histology3Lung	0.996
## Histology30ther	0.998
## Histology3Skin	1.000
## Ethnicity.2White:Age.cat65-90	0.997
## Ethnicity.2White:GENDERMale	0.998
## Age.cat65-90:GENDERMale	0.995
## Ethnicity.2White:Blood.type.cat0	0.998
## Age.cat65-90:Blood.type.cat0	0.997
## GENDERMale:Blood.type.cat0	0.997
## Ethnicity.2White:Histology3Lung	0.998
## Ethnicity.2White:Histology30ther	1.000
## Ethnicity.2White:Histology3Skin	0.998
## Age.cat65-90:Histology3Lung	0.996
## Age.cat65-90:Histology30ther	0.996
## Age.cat65-90:Histology3Skin	0.995
## GENDERMale:Histology3Lung	0.997
## GENDERMale:Histology30ther	1.000
## GENDERMale:Histology3Skin	NA

```

## Blood.type.cat0:Histology3Lung 1.000
## Blood.type.cat0:Histology30ther 1.000
## Blood.type.cat0:Histology3Skin 1.000
## Ethnicity.2White:Age.cat65-90:GENDERMale 0.995
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0 0.997
## Ethnicity.2White:GENDERMale:Blood.type.cat0 0.775
## Age.cat65-90:GENDERMale:Blood.type.cat0 0.996
## Ethnicity.2White:Age.cat65-90:Histology3Lung 0.997
## Ethnicity.2White:Age.cat65-90:Histology30ther NA
## Ethnicity.2White:Age.cat65-90:Histology3Skin NA
## Ethnicity.2White:GENDERMale:Histology3Lung 0.996
## Ethnicity.2White:GENDERMale:Histology30ther 0.999
## Ethnicity.2White:GENDERMale:Histology3Skin NA
## Age.cat65-90:GENDERMale:Histology3Lung 1.000
## Age.cat65-90:GENDERMale:Histology30ther NA
## Age.cat65-90:GENDERMale:Histology3Skin NA
## Ethnicity.2White:Blood.type.cat0:Histology3Lung 0.999
## Ethnicity.2White:Blood.type.cat0:Histology30ther 0.998
## Ethnicity.2White:Blood.type.cat0:Histology3Skin 0.996
## Age.cat65-90:Blood.type.cat0:Histology3Lung 0.999
## Age.cat65-90:Blood.type.cat0:Histology30ther NA
## Age.cat65-90:Blood.type.cat0:Histology3Skin NA
## GENDERMale:Blood.type.cat0:Histology3Lung 0.995
## GENDERMale:Blood.type.cat0:Histology30ther 0.997
## GENDERMale:Blood.type.cat0:Histology3Skin NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0 NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Lung NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology30ther NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Histology3Skin NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Lung NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology30ther NA
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0:Histology3Skin NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Lung NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology30ther NA
## Ethnicity.2White:GENDERMale:Blood.type.cat0:Histology3Skin NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung 0.995
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther NA
## Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Lung NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology30ther NA
## Ethnicity.2White:Age.cat65-90:GENDERMale:Blood.type.cat0:Histology3Skin NA
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 161.52 on 119 degrees of freedom
## Residual deviance: 100.63 on 80 degrees of freedom
## AIC: 180.63
##
## Number of Fisher Scoring iterations: 17

```

Identifying Non-Significant Interaction Values

```
Anova(int_model)
```


	LR Chisq	Df	Pr(>Chisq)
Ethnicity.2	0.0192662	1	0.8896061
Age.cat	0.1652290	1	0.6843872
GENDER	2.7627573	1	0.0964818
Blood.type.cat	3.7700610	1	0.0521777
Histology3	1.9680128	3	0.5790722
Ethnicity.2:Age.cat	0.0138203	1	0.9064164
Ethnicity.2:GENDER	4.0771763	1	0.0434662
Age.cat:GENDER	0.8584553	1	0.3541713
Ethnicity.2:Blood.type.cat	1.2093809	1	0.2714548
Age.cat:Blood.type.cat	0.1280819	1	0.7204292
GENDER:Blood.type.cat	0.4153868	1	0.5192479
Ethnicity.2:Histology3	7.1748644	3	0.0665282
Age.cat:Histology3	4.3847334	3	0.2228051
GENDER:Histology3	3.4010340	2	0.1825891
Blood.type.cat:Histology3	7.8814467	3	0.0485265
Ethnicity.2:Age.cat:GENDER	3.5726027	1	0.0587402
Ethnicity.2:Age.cat:Blood.type.cat	NA	0	NA
Ethnicity.2:GENDER:Blood.type.cat	0.0818530	1	0.7748019
Age.cat:GENDER:Blood.type.cat	2.2716863	1	0.1317565
Ethnicity.2:Age.cat:Histology3	-0.0000003	1	1.0000000
Ethnicity.2:GENDER:Histology3	10.1188280	2	0.0063493
Age.cat:GENDER:Histology3	6.4601477	1	0.0110320
Ethnicity.2:Blood.type.cat:Histology3	2.0456171	3	0.5629929
Age.cat:Blood.type.cat:Histology3	7.7283353	1	0.0054361
GENDER:Blood.type.cat:Histology3	5.7841129	2	0.0554620
Ethnicity.2:Age.cat:GENDER:Blood.type.cat	NA	0	NA
Ethnicity.2:Age.cat:GENDER:Histology3	NA	0	NA
Ethnicity.2:Age.cat:Blood.type.cat:Histology3	NA	0	NA
Ethnicity.2:GENDER:Blood.type.cat:Histology3	NA	0	NA
Age.cat:GENDER:Blood.type.cat:Histology3	-0.0000001	1	1.0000000
Ethnicity.2:Age.cat:GENDER:Blood.type.cat:Histology3	NA	0	NA

Final Interaction Model

```
int_final = glm(Adverse.Events ~ Ethnicity.2 * Age.cat * GENDER *
  Blood.type.cat * Histology3 -
  Ethnicity.2:Age.cat:GENDER:Blood.type.cat:Histology3 -
  Age.cat:GENDER:Blood.type.cat:Histology3 -
  Ethnicity.2:GENDER:Blood.type.cat:Histology3 -
  Ethnicity.2:Age.cat:Blood.type.cat:Histology3 -
  Ethnicity.2:Age.cat:GENDER:Histology3 -
  Ethnicity.2:Age.cat:GENDER:Blood.type.cat,
  family = binomial(link = "logit"),
  data = dt)
summary(int_final)

##
## Call:
## glm(formula = Adverse.Events ~ Ethnicity.2 * Age.cat * GENDER *
##   Blood.type.cat * Histology3 - Ethnicity.2:Age.cat:GENDER:Blood.type.cat:Histology3 -
##   Age.cat:GENDER:Blood.type.cat:Histology3 - Ethnicity.2:GENDER:Blood.type.cat:Histology3 -
##   Ethnicity.2:Age.cat:Blood.type.cat:Histology3 - Ethnicity.2:Age.cat:GENDER:Histology3 -
```

```

##      Ethnicity.2:Age.cat:GENDER:Blood.type.cat, family = binomial(link = "logit"),
##      data = dt)
##
## Deviance Residuals:
##      Min        1Q      Median        3Q        Max
## -1.79412  -0.75853  -0.00008   0.66805   2.14597
##
## Coefficients: (9 not defined because of singularities)
##
##              Estimate Std. Error z value
## (Intercept)    -1.957e+01  1.075e+04  -0.002
## Ethnicity.2White      7.030e-07  1.521e+04   0.000
## Age.cat65-90    -1.635e+02  3.737e+04  -0.004
## GENDERMale      -1.292e-07  1.317e+04   0.000
## Blood.type.cat0     7.607e-07  1.521e+04   0.000
## Histology3Lung      3.913e+01  1.317e+04   0.003
## Histology30ther     1.818e+01  1.075e+04   0.002
## Histology3Skin      7.531e-07  1.521e+04   0.000
## Ethnicity.2White:Age.cat65-90    1.078e+02  3.385e+04   0.003
## Ethnicity.2White:GENDERMale    -2.066e+01  1.700e+04  -0.001
## Age.cat65-90:GENDERMale      7.331e+01  1.611e+04   0.005
## Ethnicity.2White:Blood.type.cat0  3.913e+01  2.012e+04   0.002
## Age.cat65-90:Blood.type.cat0    8.824e+01  2.827e+04   0.003
## GENDERMale:Blood.type.cat0     3.913e+01  1.863e+04   0.002
## Ethnicity.2White:Histology3Lung  -2.118e+01  1.700e+04  -0.001
## Ethnicity.2White:Histology30ther  2.079e+00  1.521e+04   0.000
## Ethnicity.2White:Histology3Skin   3.913e+01  2.151e+04   0.002
## Age.cat65-90:Histology3Lung      2.638e+00  2.435e+04   0.000
## Age.cat65-90:Histology30ther     5.640e+01  1.374e+04   0.004
## Age.cat65-90:Histology3Skin      7.346e+01  1.420e+04   0.005
## GENDERMale:Histology3Lung    -3.913e+01  1.863e+04  -0.002
## GENDERMale:Histology30ther     6.931e-01  1.317e+04   0.000
## GENDERMale:Histology3Skin              NA              NA              NA
## Blood.type.cat0:Histology3Lung    -1.279e-06  2.012e+04   0.000
## Blood.type.cat0:Histology30ther     6.931e-01  1.521e+04   0.000
## Blood.type.cat0:Histology3Skin    -8.788e-07  2.151e+04   0.000
## Ethnicity.2White:Age.cat65-90:GENDERMale  -7.237e+01  1.611e+04  -0.004
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0  -7.167e+01  2.744e+04  -0.003
## Ethnicity.2White:GENDERMale:Blood.type.cat0    8.109e-01  2.833e+00   0.286
## Age.cat65-90:GENDERMale:Blood.type.cat0   -3.528e+01  5.302e+03  -0.007
## Ethnicity.2White:Age.cat65-90:Histology3Lung  -5.369e+01  1.944e+04  -0.003
## Ethnicity.2White:Age.cat65-90:Histology30ther              NA              NA              NA
## Ethnicity.2White:Age.cat65-90:Histology3Skin              NA              NA              NA
## Ethnicity.2White:GENDERMale:Histology3Lung      7.896e+01  2.186e+04   0.004
## Ethnicity.2White:GENDERMale:Histology30ther     1.708e+01  1.700e+04   0.001
## Ethnicity.2White:GENDERMale:Histology3Skin              NA              NA              NA
## Age.cat65-90:GENDERMale:Histology3Lung      8.756e+01  8.527e+03   0.010
## Age.cat65-90:GENDERMale:Histology30ther              NA              NA              NA
## Age.cat65-90:GENDERMale:Histology3Skin              NA              NA              NA
## Ethnicity.2White:Blood.type.cat0:Histology3Lung  -1.997e+01  2.437e+04  -0.001
## Ethnicity.2White:Blood.type.cat0:Histology30ther  -4.023e+01  2.012e+04  -0.002
## Ethnicity.2White:Blood.type.cat0:Histology3Skin  -7.798e+01  2.281e+04  -0.003
## Age.cat65-90:Blood.type.cat0:Histology3Lung      7.263e+01  7.568e+03   0.010
## Age.cat65-90:Blood.type.cat0:Histology30ther              NA              NA              NA
## Age.cat65-90:Blood.type.cat0:Histology3Skin              NA              NA              NA

```

## GENDERMale:Blood.type.cat0:Histology3Lung	-9.421e+01	1.983e+04	-0.005
## GENDERMale:Blood.type.cat0:Histology30ther	-3.873e+01	1.863e+04	-0.002
## GENDERMale:Blood.type.cat0:Histology3Skin	NA	NA	NA
##	Pr(> z)		
## (Intercept)	0.999		
## Ethnicity.2White	1.000		
## Age.cat65-90	0.997		
## GENDERMale	1.000		
## Blood.type.cat0	1.000		
## Histology3Lung	0.998		
## Histology30ther	0.999		
## Histology3Skin	1.000		
## Ethnicity.2White:Age.cat65-90	0.997		
## Ethnicity.2White:GENDERMale	0.999		
## Age.cat65-90:GENDERMale	0.996		
## Ethnicity.2White:Blood.type.cat0	0.998		
## Age.cat65-90:Blood.type.cat0	0.998		
## GENDERMale:Blood.type.cat0	0.998		
## Ethnicity.2White:Histology3Lung	0.999		
## Ethnicity.2White:Histology30ther	1.000		
## Ethnicity.2White:Histology3Skin	0.999		
## Age.cat65-90:Histology3Lung	1.000		
## Age.cat65-90:Histology30ther	0.997		
## Age.cat65-90:Histology3Skin	0.996		
## GENDERMale:Histology3Lung	0.998		
## GENDERMale:Histology30ther	1.000		
## GENDERMale:Histology3Skin	NA		
## Blood.type.cat0:Histology3Lung	1.000		
## Blood.type.cat0:Histology30ther	1.000		
## Blood.type.cat0:Histology3Skin	1.000		
## Ethnicity.2White:Age.cat65-90:GENDERMale	0.996		
## Ethnicity.2White:Age.cat65-90:Blood.type.cat0	0.998		
## Ethnicity.2White:GENDERMale:Blood.type.cat0	0.775		
## Age.cat65-90:GENDERMale:Blood.type.cat0	0.995		
## Ethnicity.2White:Age.cat65-90:Histology3Lung	0.998		
## Ethnicity.2White:Age.cat65-90:Histology30ther	NA		
## Ethnicity.2White:Age.cat65-90:Histology3Skin	NA		
## Ethnicity.2White:GENDERMale:Histology3Lung	0.997		
## Ethnicity.2White:GENDERMale:Histology30ther	0.999		
## Ethnicity.2White:GENDERMale:Histology3Skin	NA		
## Age.cat65-90:GENDERMale:Histology3Lung	0.992		
## Age.cat65-90:GENDERMale:Histology30ther	NA		
## Age.cat65-90:GENDERMale:Histology3Skin	NA		
## Ethnicity.2White:Blood.type.cat0:Histology3Lung	0.999		
## Ethnicity.2White:Blood.type.cat0:Histology30ther	0.998		
## Ethnicity.2White:Blood.type.cat0:Histology3Skin	0.997		
## Age.cat65-90:Blood.type.cat0:Histology3Lung	0.992		
## Age.cat65-90:Blood.type.cat0:Histology30ther	NA		
## Age.cat65-90:Blood.type.cat0:Histology3Skin	NA		
## GENDERMale:Blood.type.cat0:Histology3Lung	0.996		
## GENDERMale:Blood.type.cat0:Histology30ther	0.998		
## GENDERMale:Blood.type.cat0:Histology3Skin	NA		
##			
## (Dispersion parameter for binomial family taken to be 1)			

```
##  
##      Null deviance: 161.52  on 119  degrees of freedom  
## Residual deviance: 100.63  on   81  degrees of freedom  
## AIC: 178.63  
##  
## Number of Fisher Scoring iterations: 18
```