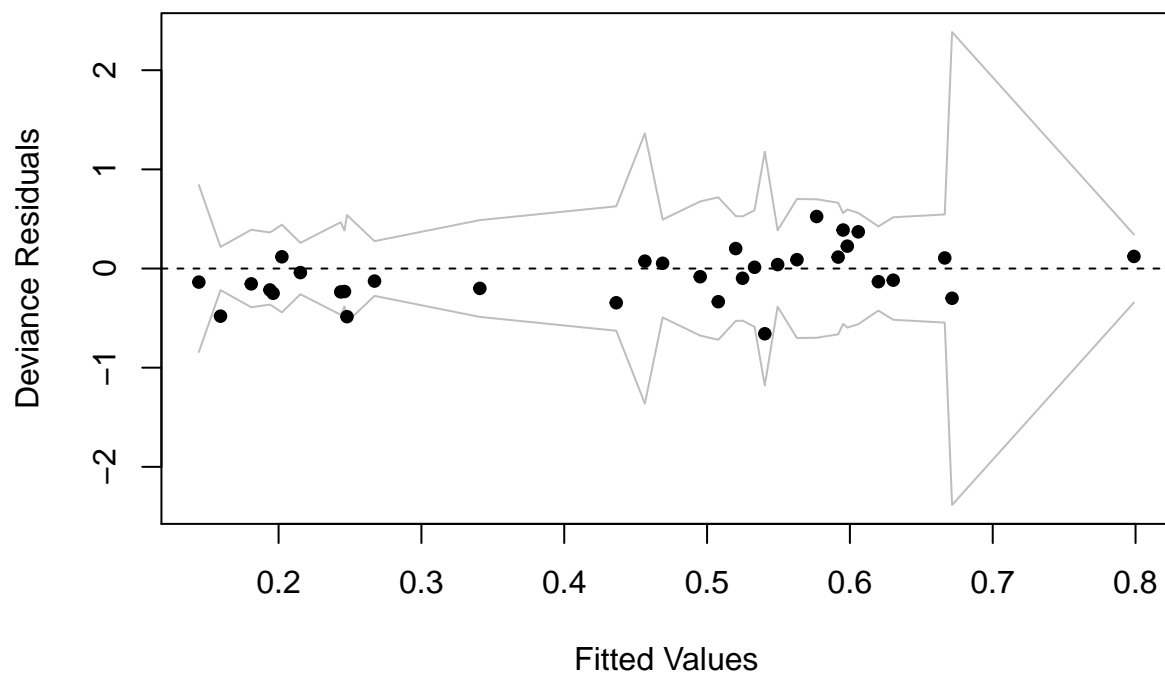


Untitled

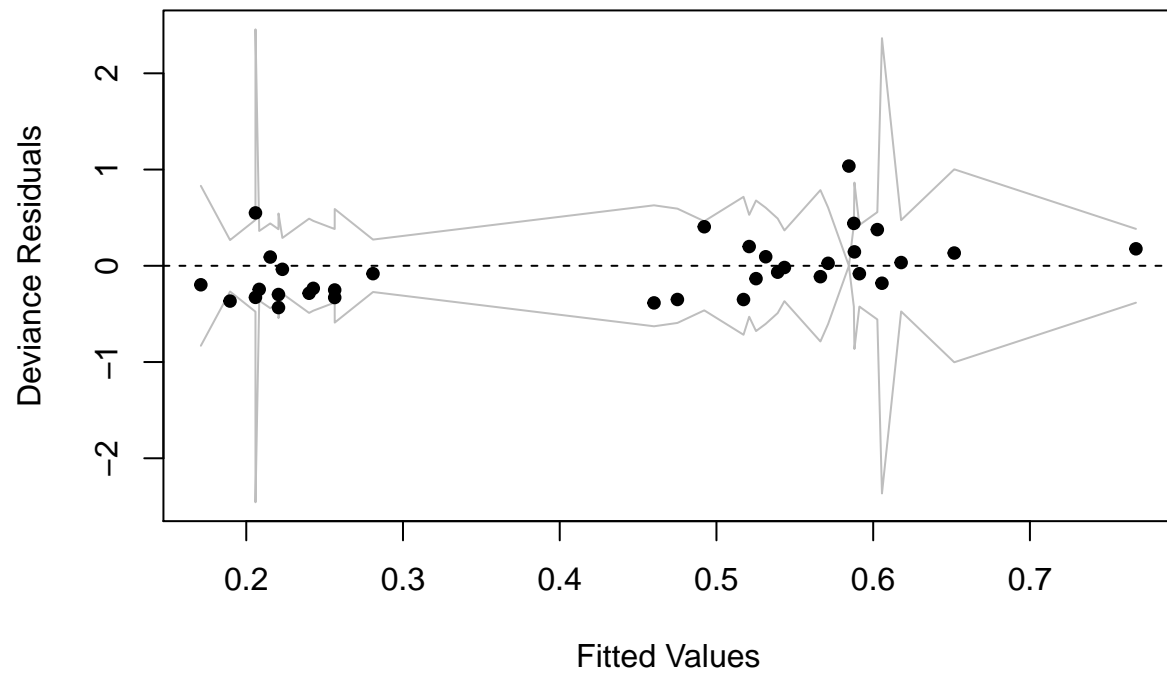
Nathaniel Brown, In Hee Ho, Sarah Zimmermann

October 19, 2017

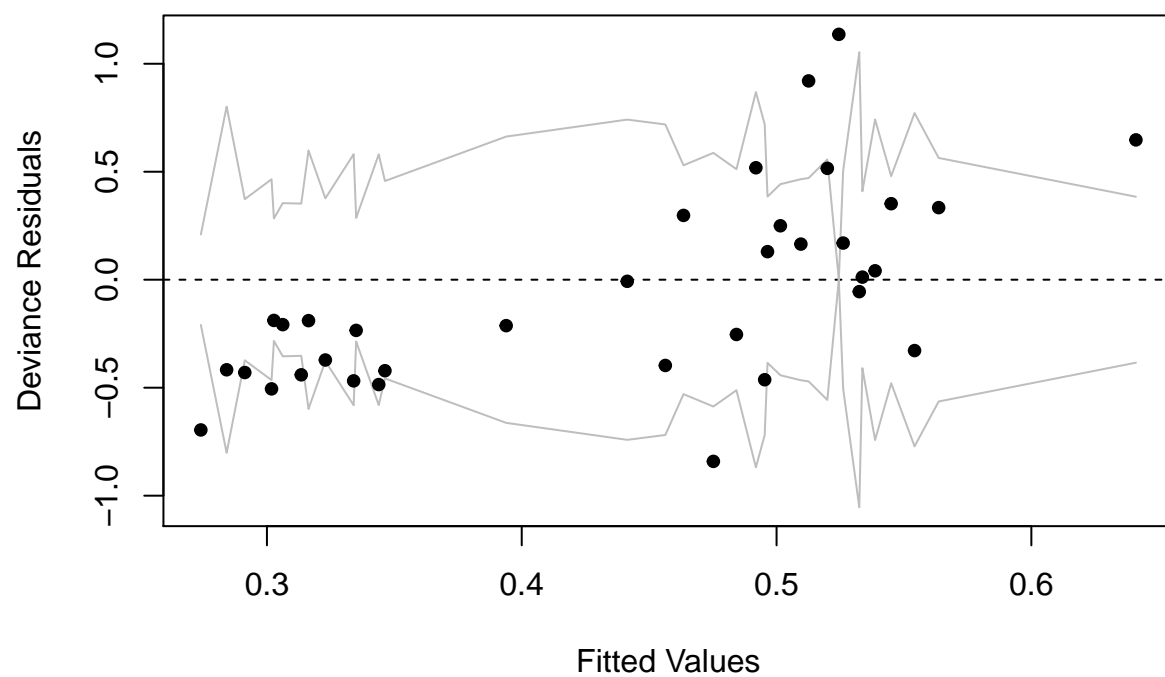
OLS Logistic Regression Binned Residuals



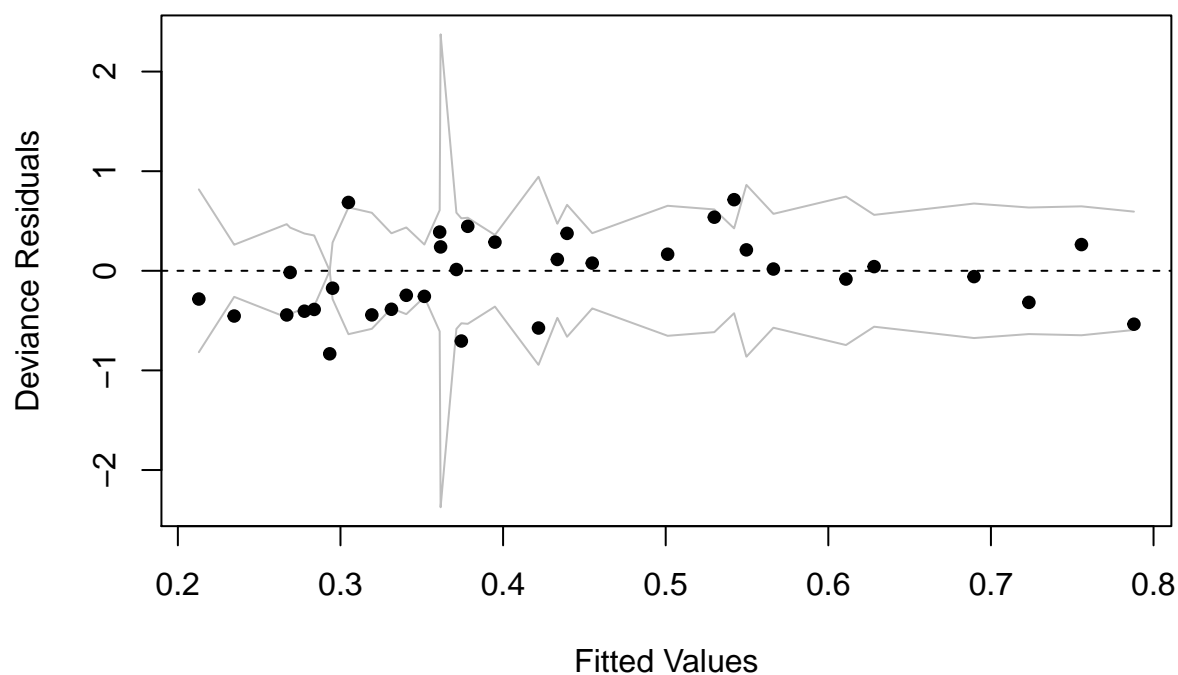
LASSO Logistic Regression Binned Residuals



Ridge Logistic Regression Binned Residuals



Kernel Logistic Regression Binned Residuals



	Deviance p-value
OLS	2e-04
LASSO Penalty	3e-04
Ridge Penalty	0e+00
Kernels	0e+00

	Lower	Upper
symptom0	-1.2348	0.1283
symptom1	-0.8128	0.4192
symptom2	-0.9683	0.3673
raceother	-0.2452	0.4814
male	-0.6261	0.0439
X1	-1.6083	-0.2653
X2	-0.1101	1.2464
X3	0.1159	1.6606
X4	-0.6474	1.2667
X5	-1.0553	1.3369
X6	-926.4905	958.4814

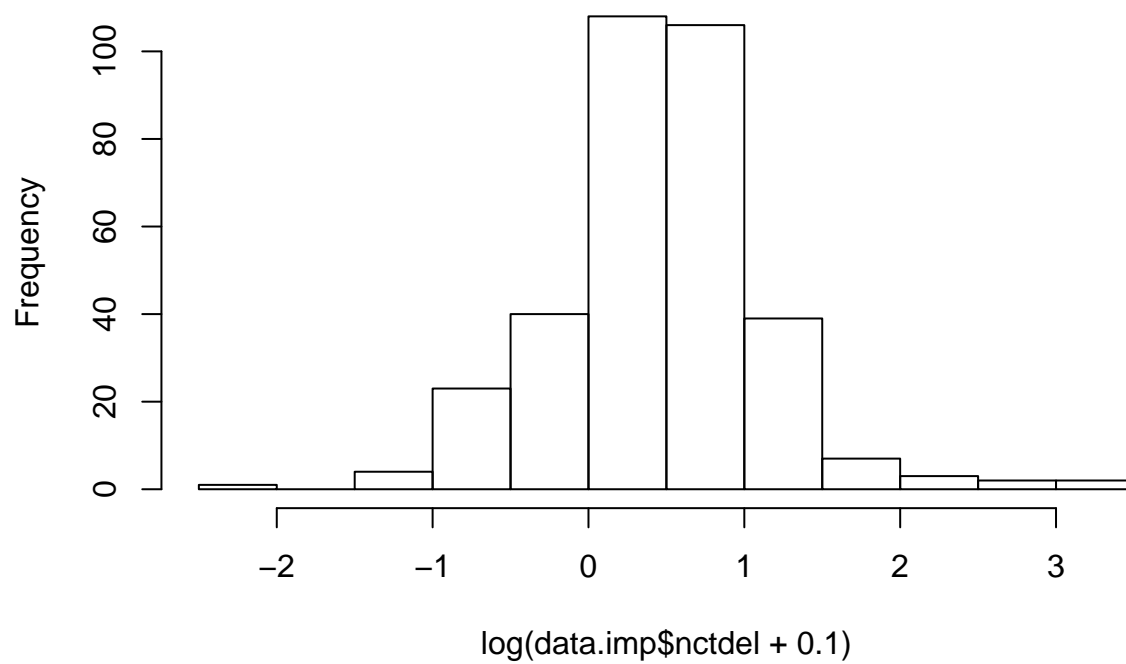
	LASSO Estimate
(Intercept)	0.0000
symptom0	0.0000

LASSO Estimate	
symptom1	0.0000
symptom2	0.0000
raceother	0.0000
male	0.0000
X1	-1.0788
X2	0.0347
X3	0.1736
X4	0.0000
X5	0.0000
X6	0.9557

Ridge Estimate	
(Intercept)	0.0000
symptom0	-0.1646
symptom1	-0.0401
symptom2	-0.0893
raceother	-0.0588
male	-0.1393
X1	-0.5499
X2	0.2039
X3	0.3068
X4	0.0513
X5	-0.0139
X6	0.9175

	Lower	Upper
symptom0	-1.3827	-0.0953
symptom1	-0.9360	0.2222
symptom2	-1.0734	0.1903
raceother	-0.2915	0.4013
male	-0.5736	0.0674
k1	-5.6259	-0.1256
k2	5.8663	13.2598

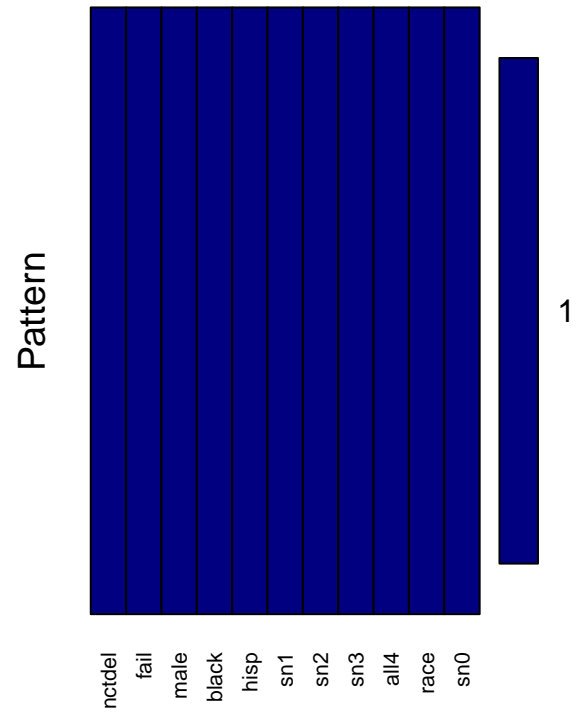
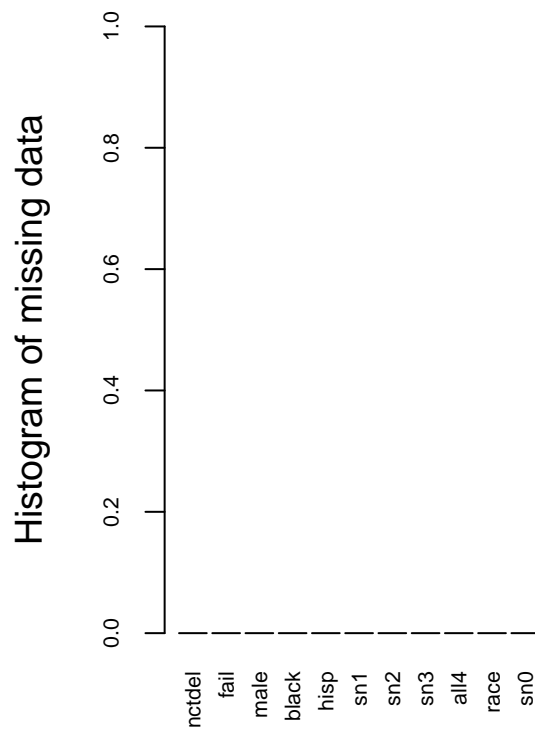
Histogram of $\log(\text{data.imp\$nctdel} + 0.1)$

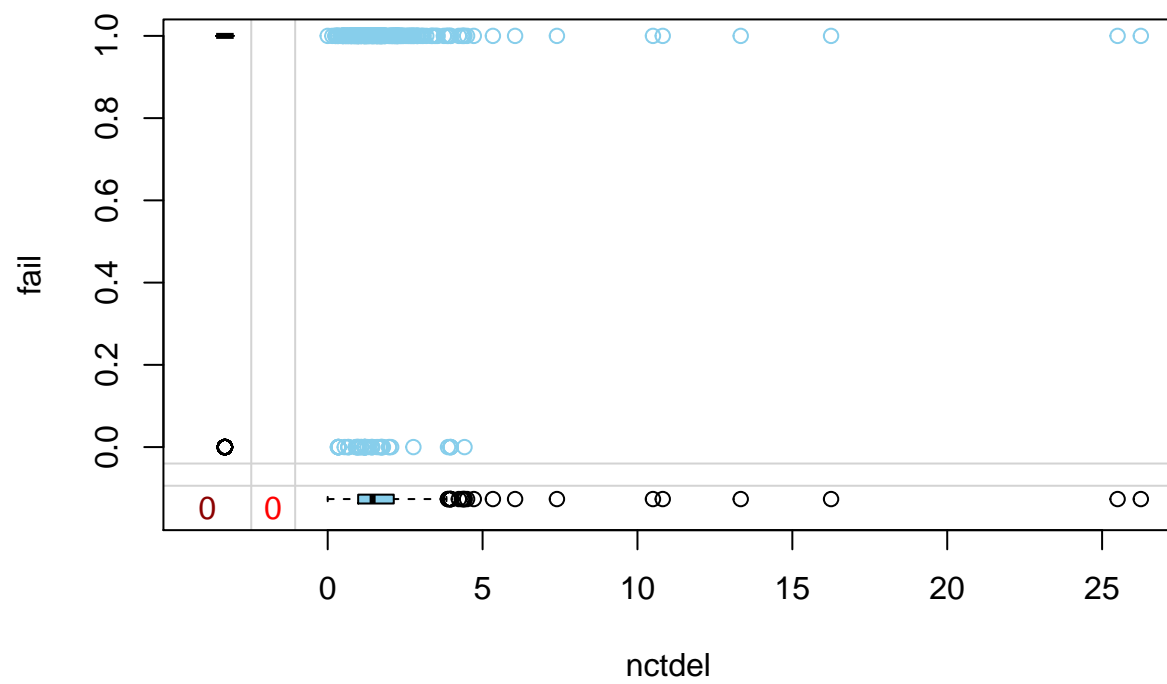


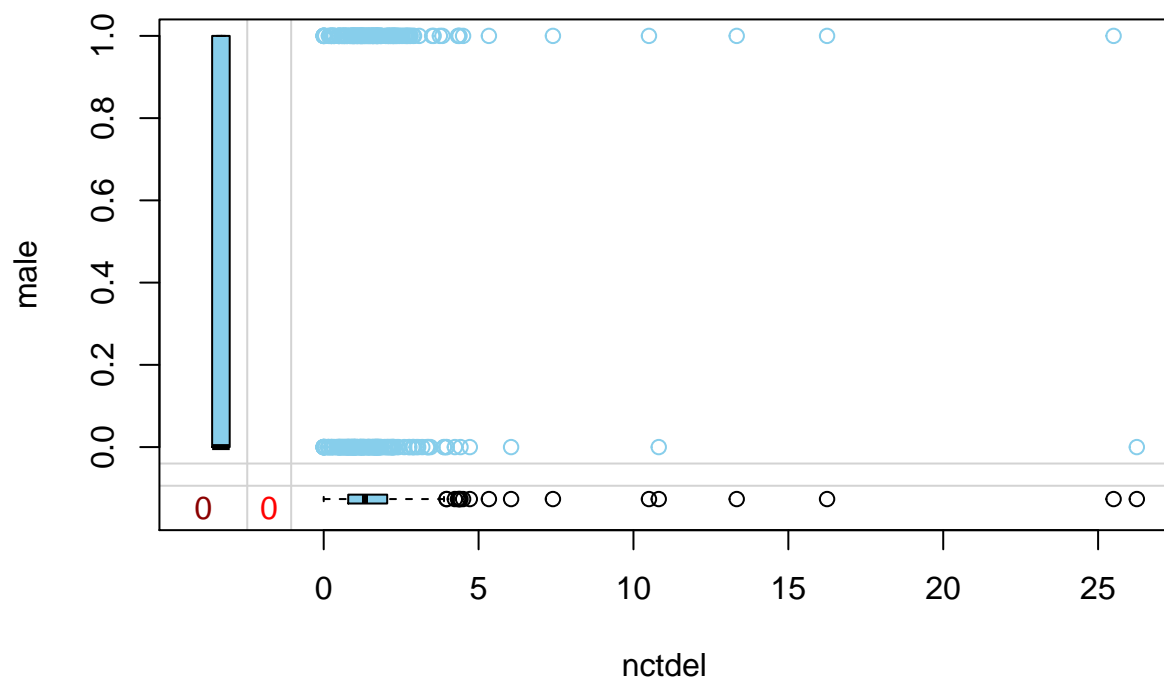
Assumptions

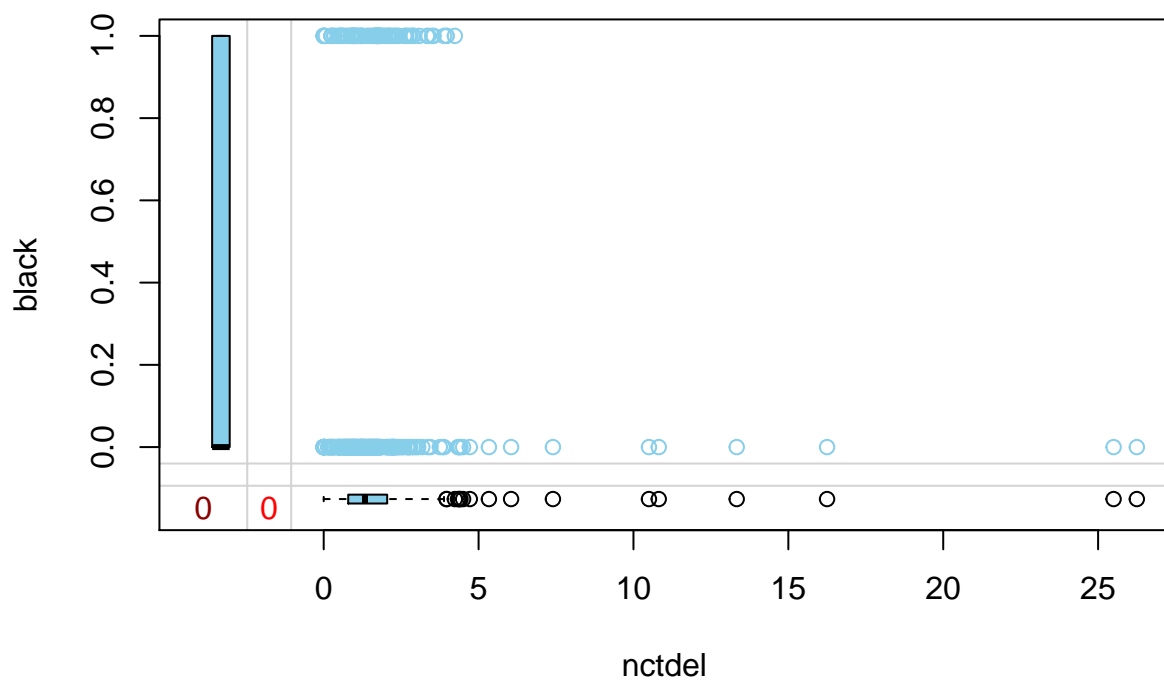
```
##      nctdel fail male black hisp sn1 sn2 sn3 all4 race sn0
## [1,]    1    1    1    1    1    1    1    1    1    1    1  0
## [2,]    0    0    0    0    0    0    0    0    0    0    0  0
```

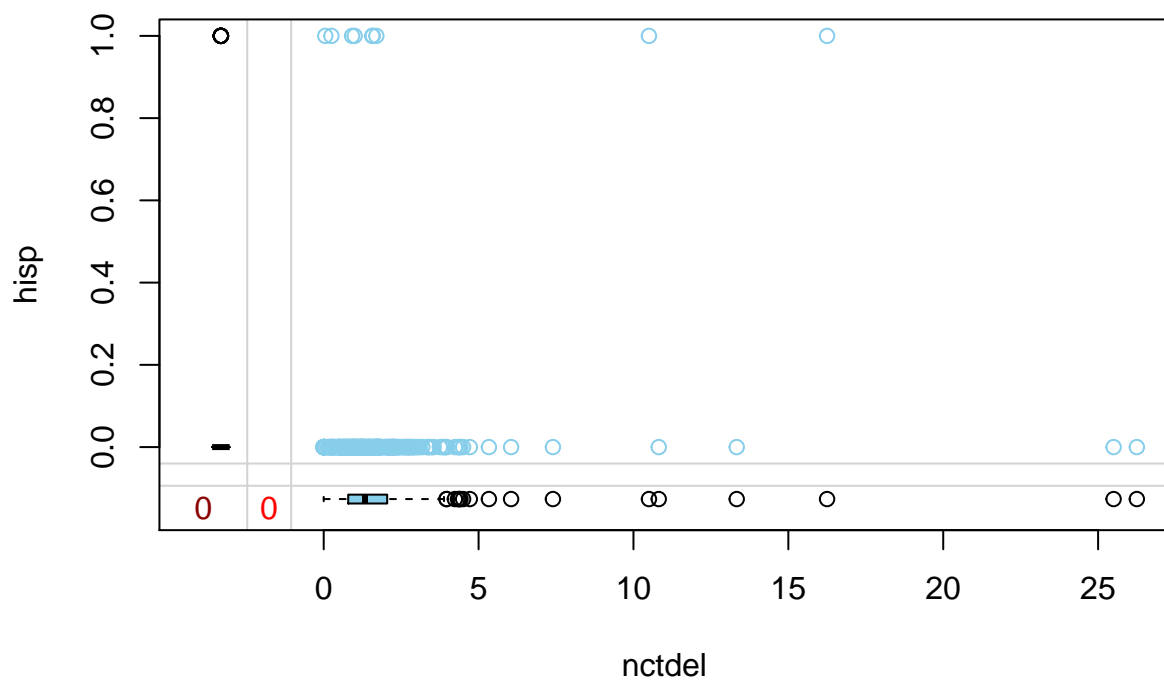
```
##
## Variables sorted by number of missings:
## Variable Count
## nctdel      0
## fail        0
## male        0
## black       0
## hisp        0
## sn1         0
## sn2         0
## sn3         0
## all4        0
## race        0
## sn0         0
```

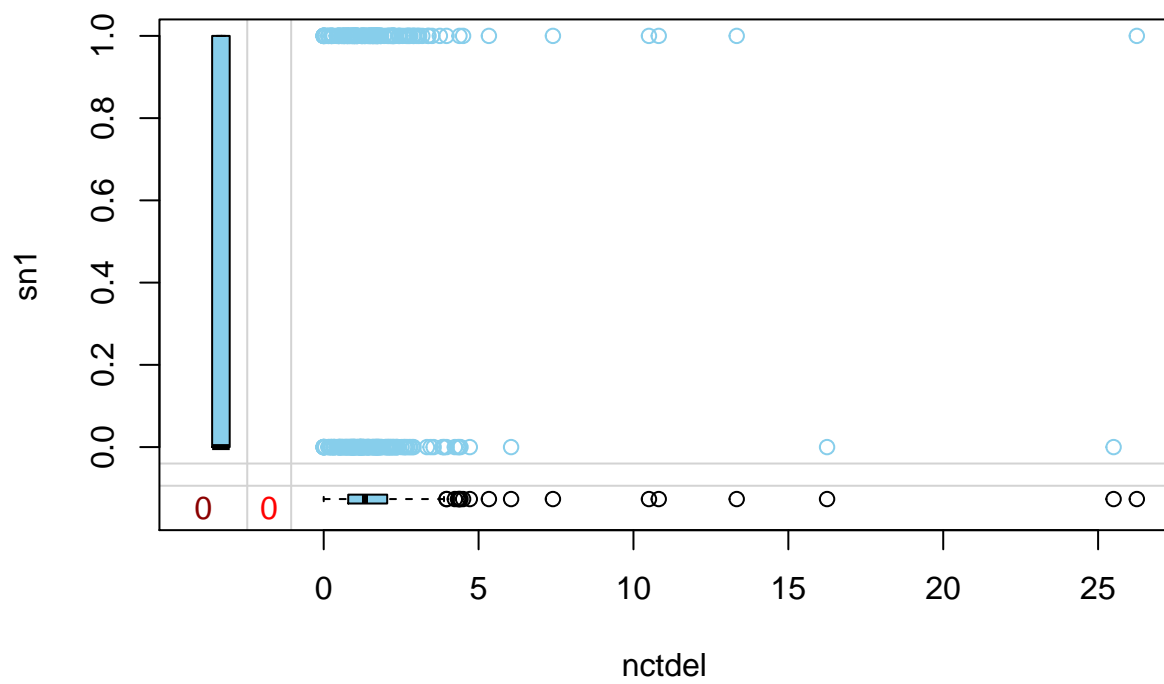


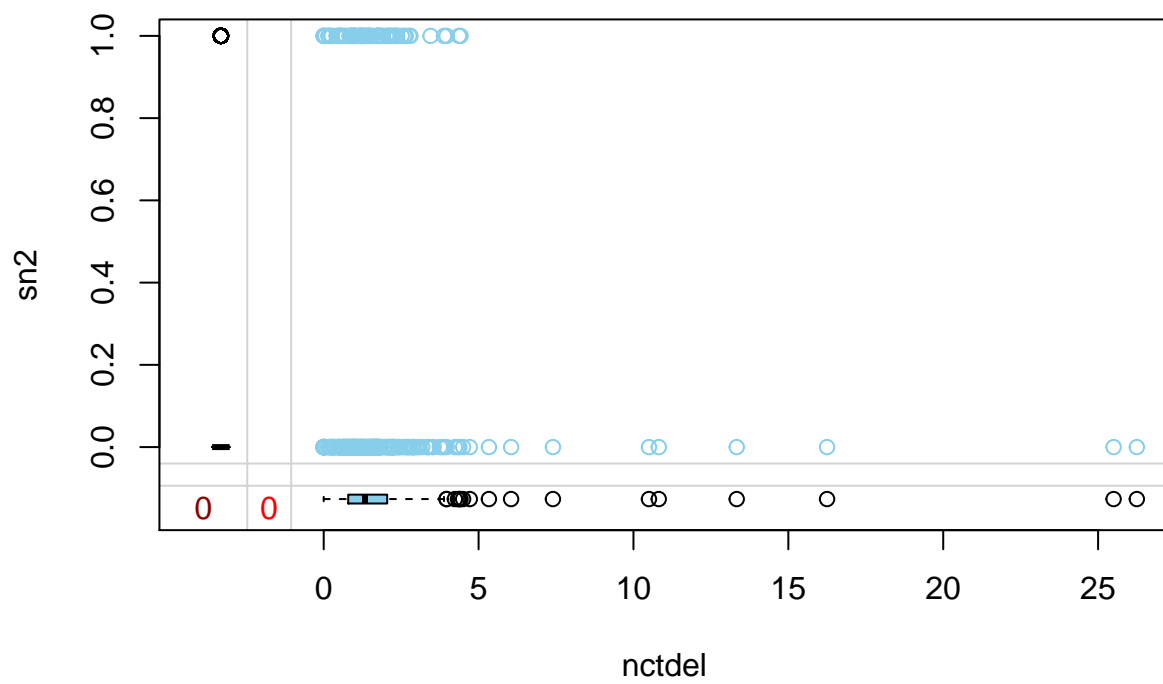


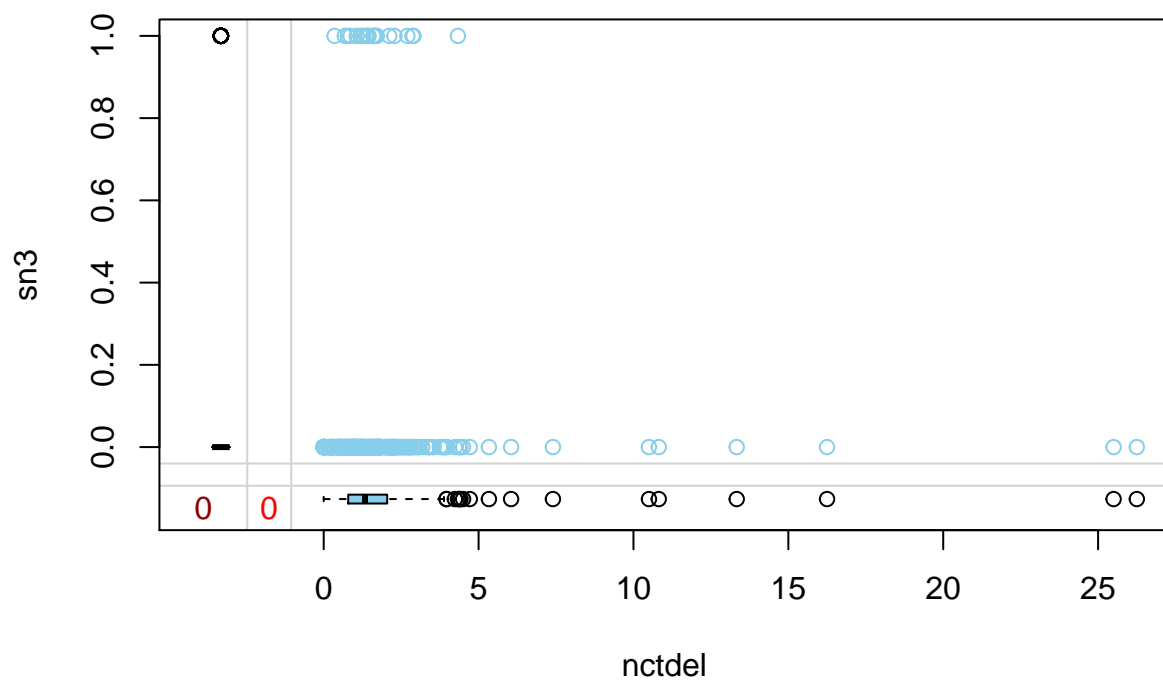


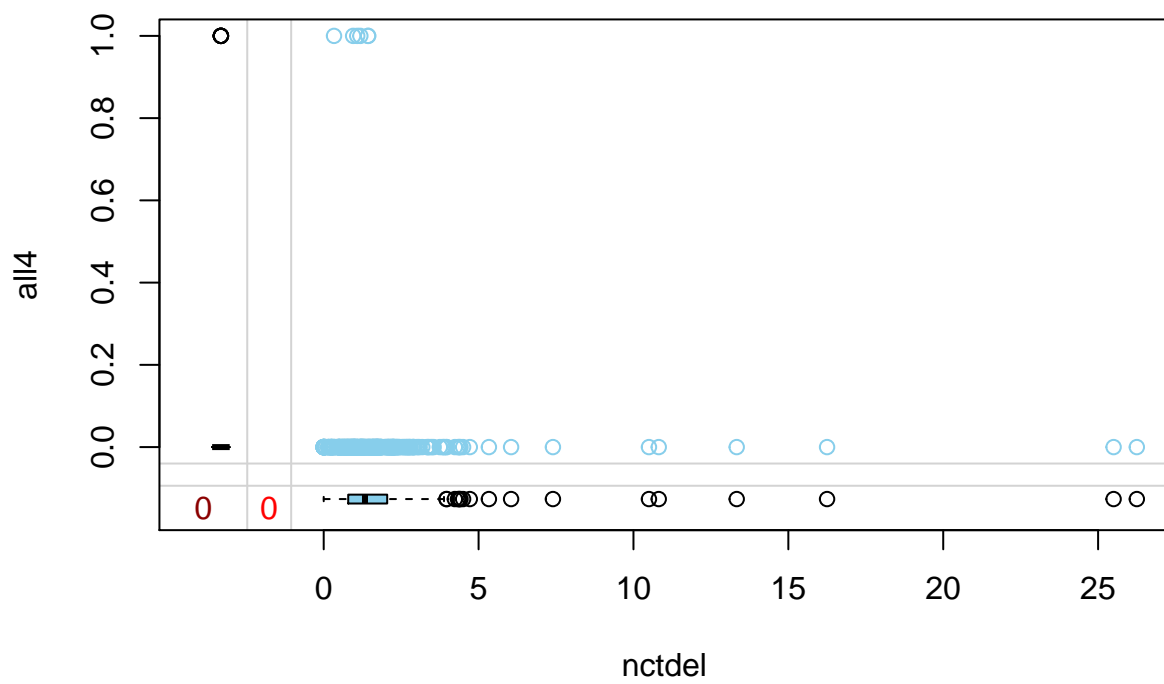


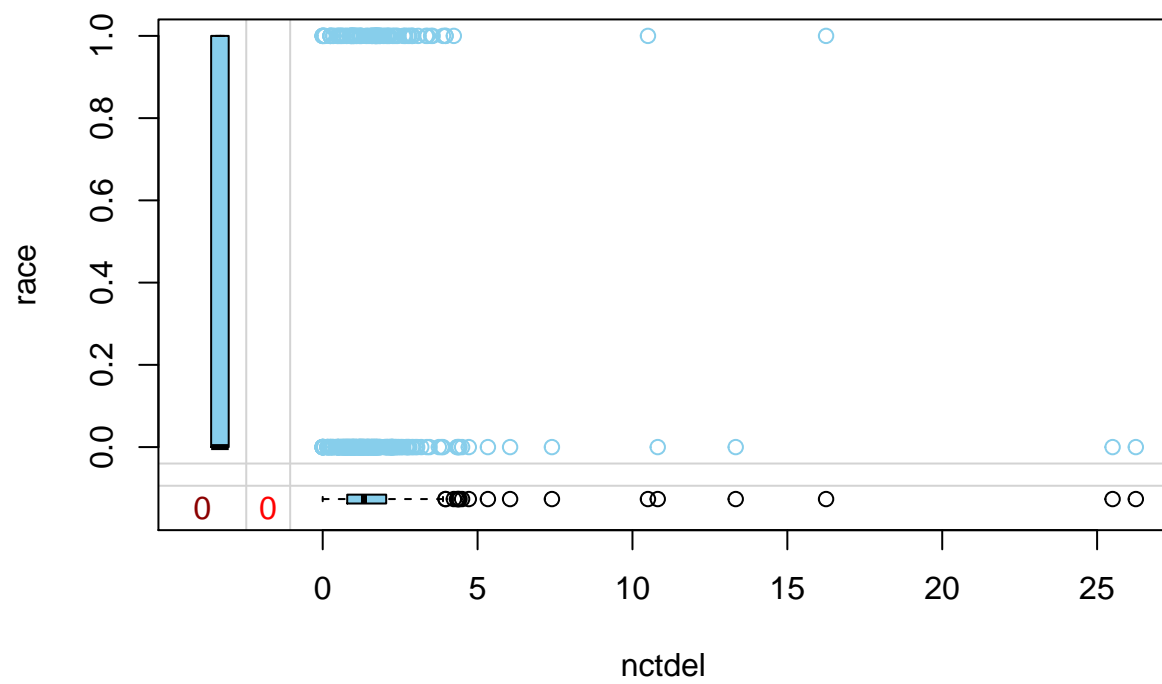






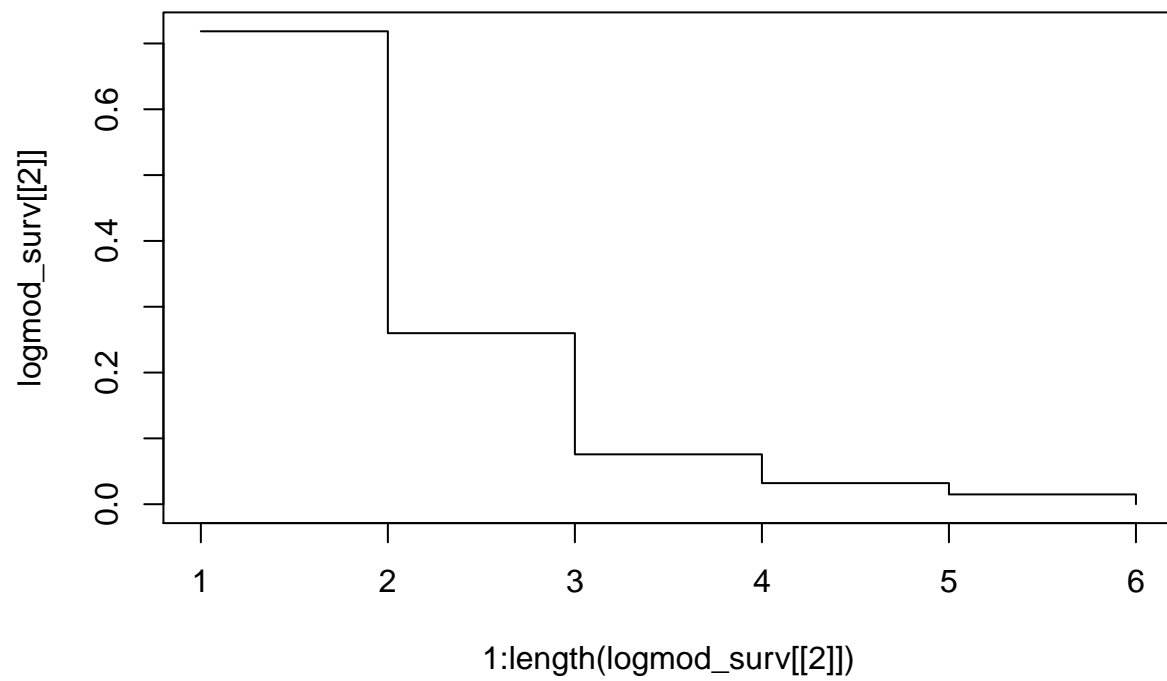


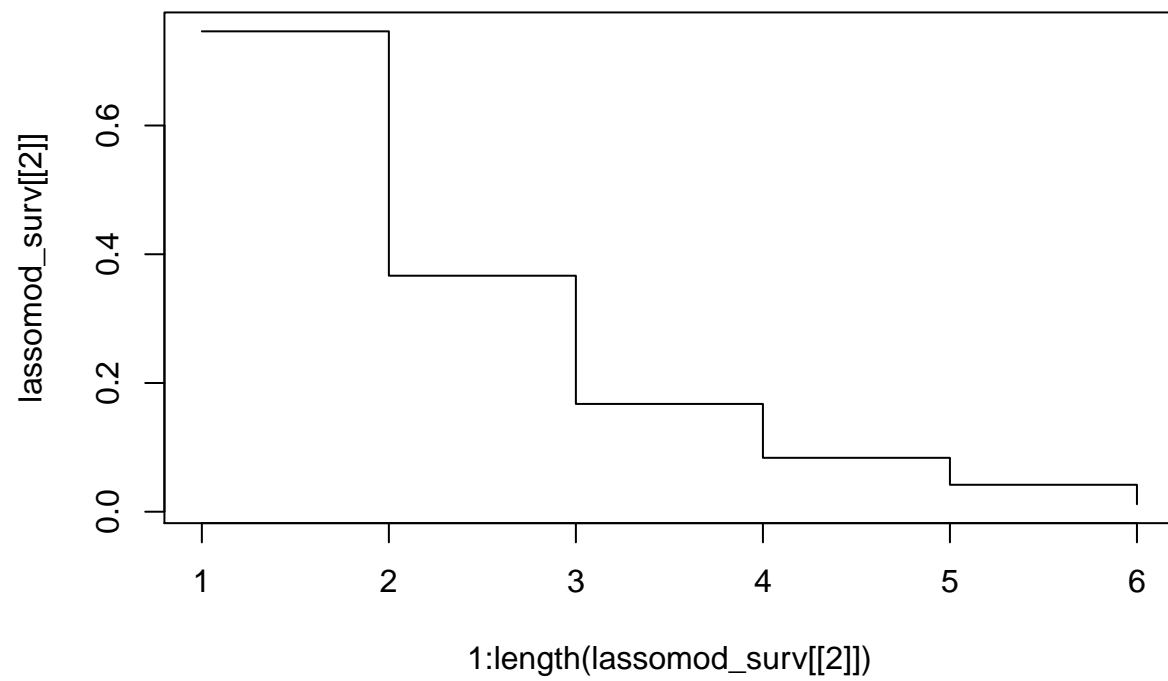


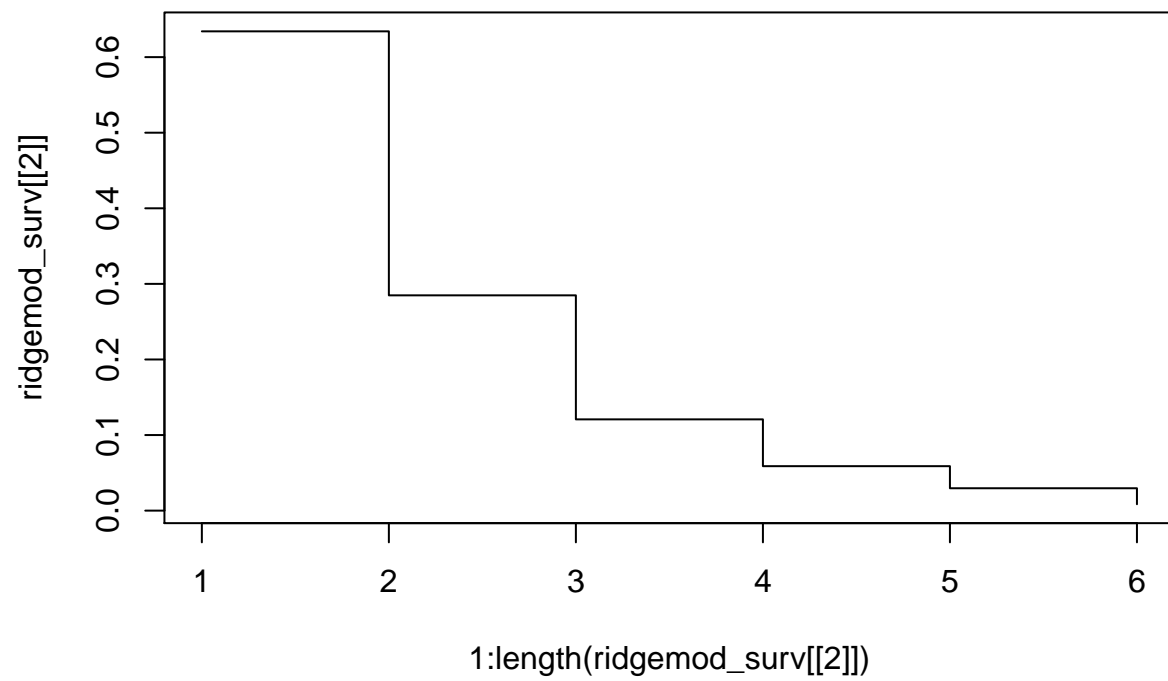


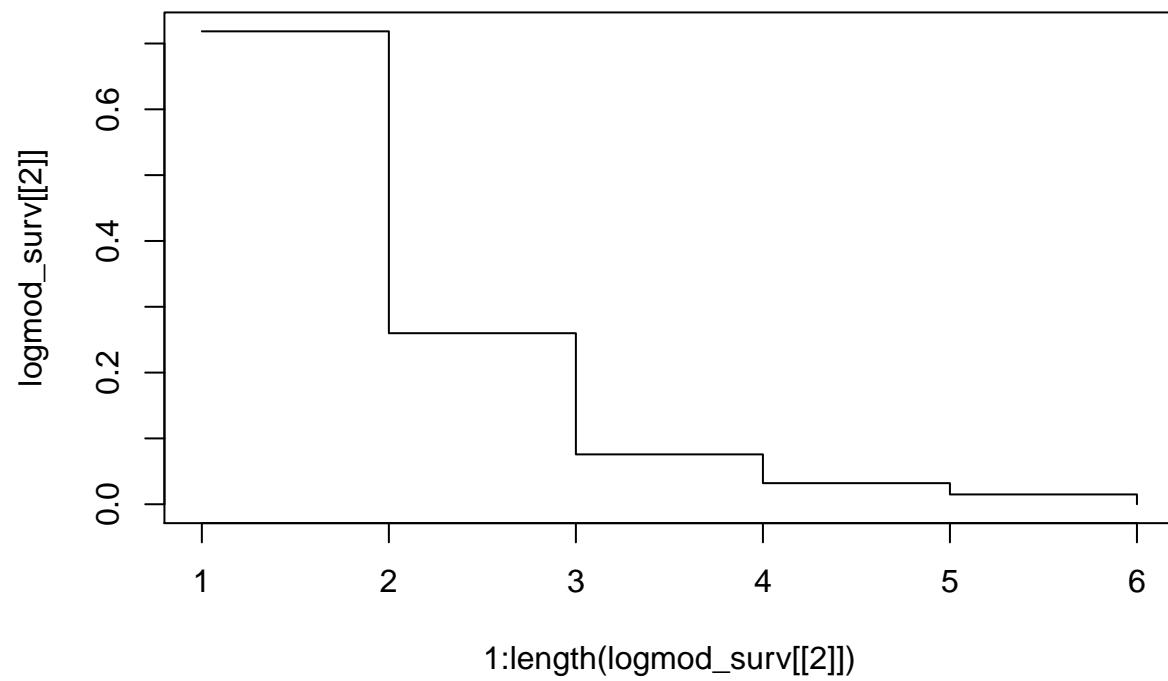
Survival Curves:

Question for Jonathan: how do we plot a survival curve from glm???

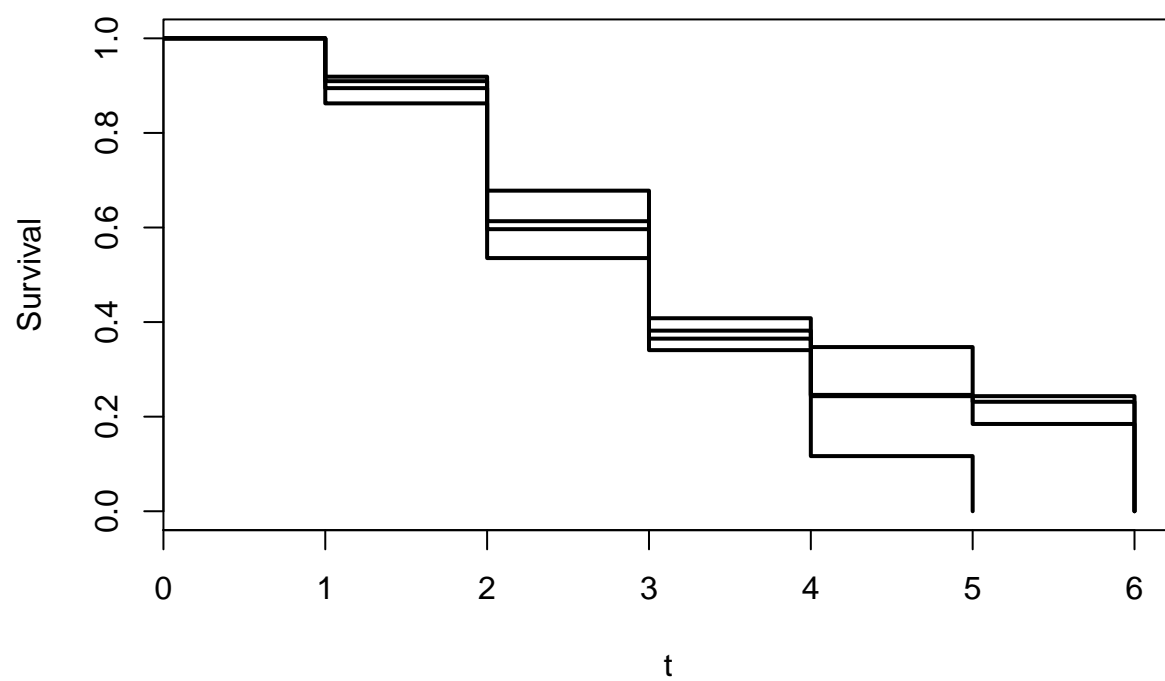




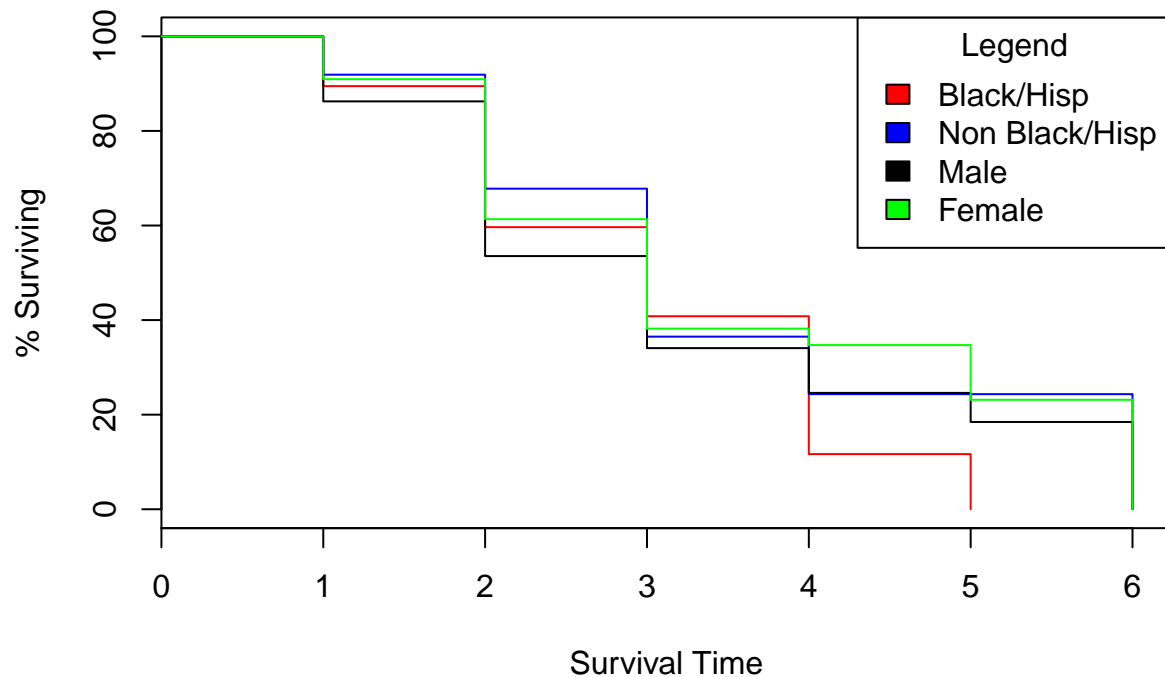




Kaplan–Meier Estimate $\hat{S}(t)$ with CI



Survival Distributions



```
## Call:
## survdiff(formula = Surv(timecat, fail) ~ raceother + male, data = datcat_X)
##
##               N Observed Expected (O-E)^2/E (O-E)^2/V
## raceother=0, male=0 95      38    35.6    0.164    0.26
## raceother=0, male=1 111     42    47.7    0.675    1.17
## raceother=1, male=0 240    103    91.2    1.517    3.20
## raceother=1, male=1 243     94   102.5    0.706    1.60
##
## Chisq= 4.4  on 3 degrees of freedom, p= 0.225
```

Discussion

why nothing is significant:

```
## # A tibble: 4 x 6
##   symptom    mean      n      sd    lower    upper
##   <chr>    <dbl> <int>   <dbl>   <dbl>   <dbl>
## 1      0 1.560370    45 0.8675425 1.306892 1.813849
## 2      1 1.547995   133 0.7804779 1.415350 1.680640
## 3      2 1.618750    56 0.7784150 1.414871 1.822629
## 4     3+ 1.493333    25 0.6746227 1.228881 1.757785

## # A tibble: 2 x 3
##   gender    mean  median
##   <chr>    <dbl>   <dbl>
```

```
## 1 female 1.516541 1.433333
## 2   male 1.606217 1.566667

## # A tibble: 2 x 3
##           race      mean  median
##           <chr>    <dbl>   <dbl>
## 1 Black or Hispanic 1.727556 1.716667
## 2           Other 1.491938 1.383333
```

References

<https://www.r-bloggers.com/imputing-missing-data-with-r-mice-package/>