

STAT UN2104 – Take Home Quiz

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```
job_sat_data <- read.table("jobsat-data.txt", header = T)

#Changing numerical variables region and age(groups) into factor variables
job_sat_data$region <- factor(job_sat_data$region)
job_sat_data$age <- factor(job_sat_data$age)

job_sat_data.logit <- glm(cbind(sat,unsat) ~ region + race + age + gender +
                          race*gender,
                          data = job_sat_data, family = binomial)

summary(job_sat_data.logit)

##
## Call:
## glm(formula = cbind(sat, unsat) ~ region + race + age + gender +
##      race * gender, family = binomial, data = job_sat_data)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.4630  -0.5681   0.1558   0.7102   2.4220
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -0.005388   0.091138  -0.059  0.95286
## region2       0.436366   0.103868   4.201 2.66e-05 ***
## region3       0.175139   0.073084   2.396  0.01656 *
## region4       0.086916   0.066481   1.307  0.19108
## region5       0.123197   0.067601   1.822  0.06839 .
## region6       0.288263   0.073567   3.918 8.91e-05 ***
## region7       0.411842   0.096729   4.258 2.07e-05 ***
## racew         0.213454   0.090890   2.348  0.01885 *
## age2          0.127823   0.051895   2.463  0.01377 *
## age3          0.363392   0.051602   7.042 1.89e-12 ***
## genderm       0.490254   0.111201   4.409 1.04e-05 ***
## racew:genderm -0.380160   0.123149  -3.087  0.00202 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 204.141  on 83  degrees of freedom
## Residual deviance:  87.941  on 72  degrees of freedom
## AIC: 476.95
##
## Number of Fisher Scoring iterations: 4

anova(job_sat_data.logit)
```

```
## Analysis of Deviance Table
##
## Model: binomial, link: logit
##
## Response: cbind(sat, unsat)
##
## Terms added sequentially (first to last)
##
##
##           Df Deviance Resid. Df Resid. Dev
## NULL                83    204.141
## region           6    34.558      77    169.583
## race             1     3.547      76    166.036
## age              2    54.202      74    111.834
## gender           1    14.317      73     97.516
## race:gender      1     9.575      72     87.941
```

To test the model's goodness of fit

```
pchisq(q=87.941, df=72, lower.tail = F)
```

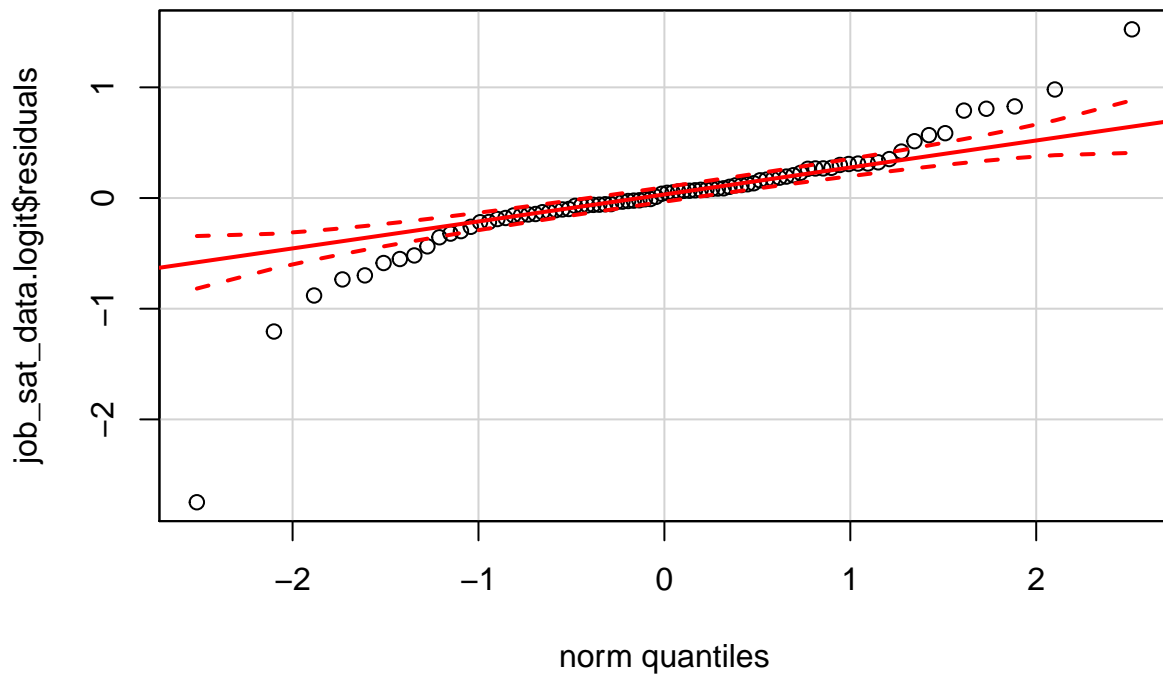
```
## [1] 0.09748707
```

```
#Proving that the residuals follow a relatively follow a standard normal
#distribution
```

```
library('car')
```

```
## Warning: package 'car' was built under R version 3.4.1
```

```
qqPlot(job_sat_data.logit$residuals)
```



Adding observed satisfaction probability percentage, fitted satisfaction probability percentage and fitted satisfaction odds percentage

```
job_sat_data.obs.prop <- job_sat_data$sat/(job_sat_data$sat + job_sat_data$unsat)
cbind(job_sat_data, job_sat_data.obs.prop, fitted(job_sat_data.logit))
```

##	region	race	age	gender	sat	unsat	job_sat_data.obs.prop
## 1	1	w	1	m	288	177	0.6193548
## 2	1	w	1	f	60	57	0.5128205
## 3	1	w	2	m	224	166	0.5743590
## 4	1	w	2	f	35	19	0.6481481
## 5	1	w	3	m	337	172	0.6620825
## 6	1	w	3	f	70	30	0.7000000
## 7	1	o	1	m	38	33	0.5352113
## 8	1	o	1	f	19	35	0.3518519
## 9	1	o	2	m	32	11	0.7441860
## 10	1	o	2	f	22	20	0.5238095
## 11	1	o	3	m	21	8	0.7241379
## 12	1	o	3	f	15	10	0.6000000
## 13	2	w	1	m	90	45	0.6666667
## 14	2	w	1	f	19	12	0.6129032
## 15	2	w	2	m	96	42	0.6956522
## 16	2	w	2	f	12	5	0.7058824

## 17	2	w	3	m	124	39	0.7607362
## 18	2	w	3	f	17	2	0.8947368
## 19	2	o	1	m	18	6	0.7500000
## 20	2	o	1	f	13	7	0.6500000
## 21	2	o	2	m	7	2	0.7777778
## 22	2	o	2	f	0	3	0.0000000
## 23	2	o	3	m	9	2	0.8181818
## 24	2	o	3	f	1	1	0.5000000
## 25	3	w	1	m	226	128	0.6384181
## 26	3	w	1	f	88	57	0.6068966
## 27	3	w	2	m	189	117	0.6176471
## 28	3	w	2	f	44	34	0.5641026
## 29	3	w	3	m	156	73	0.6812227
## 30	3	w	3	f	70	25	0.7368421
## 31	3	o	1	m	45	31	0.5921053
## 32	3	o	1	f	47	35	0.5731707
## 33	3	o	2	m	18	3	0.8571429
## 34	3	o	2	f	13	7	0.6500000
## 35	3	o	3	m	11	2	0.8461538
## 36	3	o	3	f	9	2	0.8181818
## 37	4	w	1	m	285	179	0.6142241
## 38	4	w	1	f	110	93	0.5418719
## 39	4	w	2	m	225	141	0.6147541
## 40	4	w	2	f	53	24	0.6883117
## 41	4	w	3	m	324	140	0.6982759
## 42	4	w	3	f	60	47	0.5607477
## 43	4	o	1	m	40	25	0.6153846
## 44	4	o	1	f	66	56	0.5409836
## 45	4	o	2	m	19	11	0.6333333
## 46	4	o	2	f	25	19	0.5681818
## 47	4	o	3	m	22	2	0.9166667
## 48	4	o	3	f	11	12	0.4782609
## 49	5	w	1	m	270	180	0.6000000
## 50	5	w	1	f	176	151	0.5382263
## 51	5	w	2	m	215	108	0.6656347
## 52	5	w	2	f	80	40	0.6666667
## 53	5	w	3	m	269	136	0.6641975
## 54	5	w	3	f	110	40	0.7333333
## 55	5	o	1	m	36	20	0.6428571
## 56	5	o	1	f	25	16	0.6097561
## 57	5	o	2	m	9	7	0.5625000
## 58	5	o	2	f	11	5	0.6875000
## 59	5	o	3	m	16	3	0.8421053
## 60	5	o	3	f	4	5	0.4444444
## 61	6	w	1	m	252	126	0.6666667
## 62	6	w	1	f	97	61	0.6139241
## 63	6	w	2	m	162	72	0.6923077
## 64	6	w	2	f	47	27	0.6351351
## 65	6	w	3	m	199	93	0.6815068
## 66	6	w	3	f	62	24	0.7209302
## 67	6	o	1	m	69	27	0.7187500
## 68	6	o	1	f	45	36	0.5555556
## 69	6	o	2	m	14	7	0.6666667
## 70	6	o	2	f	8	4	0.6666667

## 71	6	o	3	m	14	5	0.7368421
## 72	6	o	3	f	2	0	1.0000000
## 73	7	w	1	m	119	58	0.6723164
## 74	7	w	1	f	62	33	0.6526316
## 75	7	w	2	m	66	20	0.7674419
## 76	7	w	2	f	20	10	0.6666667
## 77	7	w	3	m	67	21	0.7613636
## 78	7	w	3	f	25	10	0.7142857
## 79	7	o	1	m	45	16	0.7377049
## 80	7	o	1	f	22	15	0.5945946
## 81	7	o	2	m	15	10	0.6000000
## 82	7	o	2	f	10	8	0.5555556
## 83	7	o	3	m	8	6	0.5714286
## 84	7	o	3	f	6	2	0.7500000
##	fitted(job_sat_data.logit)						
## 1							0.5788758
## 2							0.5518297
## 3							0.6096837
## 4							0.5831915
## 5							0.6640850
## 6							0.6390995
## 7							0.6188964
## 8							0.4986531
## 9							0.6485540
## 10							0.5305705
## 11							0.7002016
## 12							0.5885571
## 13							0.6801643
## 14							0.6557548
## 15							0.7073088
## 16							0.6840085
## 17							0.7536024
## 18							0.7325941
## 19							0.7152933
## 20							0.6061073
## 21							0.7405935
## 22							0.6361751
## 23							0.7832360
## 24							0.6887689
## 25							0.6208833
## 26							0.5946459
## 27							0.6504736
## 28							0.6250473
## 29							0.7019688
## 30							0.6784366
## 31							0.6592616
## 32							0.5423361
## 33							0.6873648
## 34							0.5738493
## 35							0.7356338
## 36							0.6302158
## 37							0.5999066
## 38							0.5732154
## 39							0.6301590

```
## 40          0.6041542
## 41          0.6831894
## 42          0.6588949
## 43          0.6391744
## 44          0.5203707
## 45          0.6681002
## 46          0.5521474
## 47          0.7181238
## 48          0.6094307
## 49          0.6085823
## 50          0.5820666
## 51          0.6385738
## 52          0.6127972
## 53          0.6909893
## 54          0.6670015
## 55          0.6474989
## 56          0.5294182
## 57          0.6760956
## 58          0.5611010
## 59          0.7254094
## 60          0.6180313
## 61          0.6471245
## 62          0.6215963
## 63          0.6757363
## 64          0.6511622
## 65          0.7250826
## 66          0.7026023
## 67          0.6841975
## 68          0.5702510
## 69          0.7111452
## 70          0.6012552
## 71          0.7570404
## 72          0.6561687
## 73          0.6748057
## 74          0.6501976
## 75          0.7022060
## 76          0.6786840
## 77          0.7490202
## 78          0.7277624
## 79          0.7102726
## 80          0.6002373
## 81          0.7358543
## 82          0.6304800
## 83          0.7790433
## 84          0.6834875
```

```
logit.fit.prob <- fitted(job_sat_data.logit)
logit.fit.ods <- logit.fit.prob/(1-logit.fit.prob)

#Creation of a new data frame with added columns
sat_data <- cbind(job_sat_data, logit.fit.prob, logit.fit.ods)

sat_data_region1 <- subset(sat_data, region=="1")
sat_data_region2 <- subset(sat_data, region=="2")
```

```
sat_data_region3 <- subset(sat_data, region=="3")
sat_data_region4 <- subset(sat_data, region=="4")
sat_data_region5 <- subset(sat_data, region=="5")
sat_data_region6 <- subset(sat_data, region=="6")
sat_data_region7 <- subset(sat_data, region=="7")
```

Contingency Tables by Race and Region

```
sat_data_race_w <- subset(sat_data, race=="w")
sat_data_race_o <- subset(sat_data, race=="o")
```

Contingency Tables by Age Group, and Region

```
sat_data_age1 <- subset(sat_data, age=="1")
sat_data_age2 <- subset(sat_data, age=="2")
sat_data_age3 <- subset(sat_data, age=="3")
```

Contingency Table by Gender and Region

```
sat_data_f <- subset(sat_data, gender=="f")
sat_data_m <- subset(sat_data, gender=="m")

sat_data_w_m <- subset(sat_data, race=="w" & gender == "m")
sat_data_w_f <- subset(sat_data, race=="w" & gender == "f")
sat_data_o_m <- subset(sat_data, race=="o" & gender == "m")
sat_data_o_f <- subset(sat_data, race=="o" & gender == "f")
```