HW4 yq2378

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1 summarize th data

```
df1 = tibble(
  satisfaction = rep(c("low", "medium", "high"), each = 6),
  contact = rep(rep(c("low", "high"), each = 3),3),
 housing = rep(c("Tower Block", "Apartment", "House"),6),
  cnt = c(65,130,67,34,141,130,54,76,48,47,116,105,100,111,62,100,191,104))
  mutate(satisfaction = factor(satisfaction, levels = c("low", "medium", "high"), ordered = TRUE),
         contact = factor(contact, levels = c("high","low")) )
# Create a table for satisfaction vs contact
contact_table <- xtabs(cnt ~ satisfaction + contact, data = df1)</pre>
# Convert counts to percentages
round(prop.table(contact_table,2) * 100,2)
               contact
## satisfaction high
##
         low
               31.51 36.75
##
         medium 27.69 24.96
              40.81 38.29
        high
round(prop.table(contact_table,1) * 100,2)
##
               contact
## satisfaction high
                        low
##
         low
               53.79 46.21
         medium 60.09 39.91
##
              59.13 40.87
         high
# Create a table for satisfaction vs housing
housing_table <- xtabs(cnt ~ satisfaction + housing, data = df1)
# Convert counts to percentages
round(prop.table(housing table,2) * 100,2)
##
               housing
## satisfaction Apartment House Tower Block
##
         low
                    35.42 38.18
                                      24.75
##
         medium
                    25.10 29.65
                                      25.25
                                      50.00
##
         high
                    39.48 32.17
```

1) contact with other residents and levels of satisfaction

High contact with other residents has a larger proportion of high satisfaction of residents (40.81%) compared to low contact with other residents (38.29%). However, medium satisfaction residents have the least proportion in both high and low contact with other.

2) type of housing and levels of satisfaction

Those living in the block has 50% of high satisfaction and nearly equal amount of low and medium satisfaction. While for those living in the house, the majority (38.18%) has low satisfaction.

2

```
df2 = tibble(
  contact = rep(c("low", "high"), each = 3),
  housing = rep(c("Tower Block", "Apartment", "House"), 2),
  low_satisfaction = c(65,130,67,34,141,130),
  medium_satisfaction = c(54,76,48,47,116,105),
  high_satisfaction = c(100, 111, 62, 100, 191, 104))
library(nnet)
mult <- multinom(cbind(low_satisfaction, medium_satisfaction, high_satisfaction)~contact+housing, data=
## # weights: 15 (8 variable)
## initial value 1846.767257
## iter 10 value 1803.046285
## final value 1802.740161
## converged
summary_mult <- summary(mult)</pre>
summary_mult
## Call:
## multinom(formula = cbind(low_satisfaction, medium_satisfaction,
##
       high_satisfaction) ~ contact + housing, data = df2)
##
## Coefficients:
##
                       (Intercept) contactlow housingHouse housingTower Block
## medium_satisfaction -0.2180364 -0.2959832
                                                 0.06967922
                                                                     0.4067631
                         0.2474047 -0.3282264 -0.30402275
## high satisfaction
                                                                     0.6415948
##
## Std. Errors:
##
                       (Intercept) contactlow housingHouse housingTower Block
## medium_satisfaction 0.10930968 0.1301046
                                                  0.1437749
                                                                     0.1713009
                        0.09783068 0.1181870
                                                  0.1351693
                                                                     0.1500774
## high_satisfaction
## Residual Deviance: 3605.48
## AIC: 3621.48
```

For "medium satisfaction" relative to the baseline "low satisfaction"

Being low in contact with others has a negative coefficient (-0.2960), suggesting that lower contact is associated with a decrease in the log-odds of medium satisfaction. As for housing in House and Tower Block, both have positive coefficient (0.0697 and 0.4068 respectively), suggesting an increase in the log-odds of medium

satisfaction and clearly Tower Block has a stronger impact than House.

For "high satisfaction" relative to the baseline "low satisfaction"

coefficients <- summary_mult\$coefficients</pre>

[1] 6.893028

iter 10 value 1800.128659

Being low in contact with others has a negative coefficient (-0.3282), which means larger impact, but still indicates consistency in the direction with medium satisfaction. Housing in House has a negative coefficient (-0.3040), suggesting living in a house is associated with a decrease in the log-odds of high satisfaction. Housing in Tower Block has a positive coefficient, significantly increasing the log-odds of high satisfaction for tower block residents

```
std_errors <- summary_mult$standard.errors</pre>
# Calculate 95% Confidence Intervals
conf_int_lower <- coefficients - 1.96 * std_errors</pre>
conf_int_upper <- coefficients + 1.96 * std_errors</pre>
CI = paste("(",round(conf_int_lower,2),", ", round(conf_int_upper,2), ")",sep = "")
CI = rbind(CI[seq(1,8,by =2)],CI[seq(2,8,by =2)])
colnames(CI) = colnames(conf_int_lower)
rownames(CI) = rownames(conf_int_lower)
print("95% Confidence Interval")
## [1] "95% Confidence Interval"
CI
##
                        (Intercept)
                                        contactlow
                                                          housingHouse
                                        "(-0.55, -0.04)" "(-0.21, 0.35)"
## medium_satisfaction "(-0.43, 0)"
## high_satisfaction
                        "(0.06, 0.44)" "(-0.56, -0.1)" "(-0.57, -0.04)"
                        housingTower Block
## medium_satisfaction "(0.07, 0.74)"
## high_satisfaction
                        "(0.35, 0.94)"
From the CI, we find only housing being House and medium satisfaction has a CI contains 0. It indicates
except for this combination, all other coefficients are significant under 0.05 significance value.
# goodness of fit
pihat1=predict(mult,type='probs')
m = rowSums(df2[,3:5])
res.pearson1=(df2[,3:5]-pihat1*m)/sqrt(pihat1*m) # pearson residuals
G.stat1=sum(res.pearson1^2) # Generalized Pearson Chisq Stat
G.stat1 # Generalized Pearson chisq statistic
## [1] 6.932341
D.stat1=sum(2*df2[,3:5]*log(df2[,3:5]/(m*pihat1)))
D.stat1 # Deviance statistics
```

As for teh goodness of fit, the Generalized Pearson chisq statistic is 6.9323412 and the Deviance statistics is 6.8930277.

mult2 <- multinom(cbind(low_satisfaction, medium_satisfaction, high_satisfaction)~contact*housing, data
weights: 21 (12 variable)
initial value 1846.767257</pre>

```
## final value 1799.293647
## converged
#summary(mult2)
anova(mult, mult2, test="Chisq")
##
                 Model Resid. df Resid. Dev
                                                Test
                                                        Df LR stat.
                                                                       Pr(Chi)
## 1 contact + housing
                                4
                                     3605.480
                                                        NA
                                                                  NA
                                                                            NA
                                                          4 6.893028 0.1416504
## 2 contact * housing
                                0
                                     3598.587 1 vs 2
```

Given interaction leads to a nested model, we test if the null model stands, which suggests that there is no interaction term. From the result we know the p-value is 0.1417. Under significance value of 0.05, we fail to reject the null hypothesis, believing there is no interaction of contact level by house type.

3

```
library (MASS)
##
## Attaching package: 'MASS'
## The following object is masked from 'package:dplyr':
##
##
       select
# fit proportional odds model
polr= polr(satisfaction~contact+housing,data=df1,weights=cnt)
summary(polr)
##
## Re-fitting to get Hessian
## Call:
## polr(formula = satisfaction ~ contact + housing, data = df1,
##
       weights = cnt)
##
##
  Coefficients:
##
                         Value Std. Error t value
## contactlow
                                  0.09306
                                           -2.713
                       -0.2524
## housingHouse
                       -0.2353
                                           -2.236
                                  0.10521
## housingTower Block 0.5010
                                  0.11675
                                            4.291
##
##
  Intercepts:
##
               Value
                        Std. Error t value
## low|medium -0.7488
                        0.0818
                                   -9.1570
  medium|high 0.3637
                                    4.5393
##
                        0.0801
## Residual Deviance: 3610.286
## AIC: 3620.286
```

- contact low: This coefficient (0.2524) represents the effect of the "contact" predictor on the log odds of being in a higher satisfaction category, holding other variables constant. The positive sign indicates that being contact low is associated with lower odds of being in a higher satisfaction category compared to being contact high. This trend is consistent with nominal regression. The t-value of -2.713 and the corresponding standard error suggest that this effect is statistically significant.
- housing House: The coefficient (0.2353) for "housing House" indicates that living in a house, compared

to the baseline housing category, Apartment and among those with same level of contact, is associated with lower odds of being in a higher satisfaction category. The negative sign denotes a decrease in the odds of higher satisfaction for residents in houses, and the t-value of -2.236 suggests this effect is statistically significant. This is obviously not consistent with the nominal regression, as whose result shows non-monotone trend.

• housing Tower Block: The coefficient (-0.5010) for "housing Tower Block" suggests that living in a tower block, relative to the baseline, and among those with same level of contact, significantly increases the odds of being in a higher satisfaction category. This is consistent with the nominal regression. The positive coefficient and the t-value of 4.291 strongly indicate this is a significant predictor of higher satisfaction levels.

4 the largest discrepancies are between the observed frequencies and the expted frequencies.

```
# residuals
pihat=predict(polr,df1,type='p')
m=rowSums(cbind(df2$low_satisfaction,df2$medium_satisfaction,df2$high_satisfaction))
res.pearson=(df2[,3:5]-pihat*m)/sqrt(pihat*m)
G=sum(res.pearson^2)
which(abs(res.pearson) == max(abs(res.pearson)), arr.ind = TRUE)
##
        row col
## [1,]
          4
df2[4,]
## # A tibble: 1 x 5
     contact housing
                          low satisfaction medium satisfaction high satisfaction
##
##
     <chr>>
             <chr>>
                                     <dbl>
                                                          <dbl>
                                                                             <dbl>
## 1 high
             Tower Block
                                                                               100
```

The Pearson chisq-square statistic is 181.6268587 and the high satisfaction with high contacts and housing in Tower Block has the largest discrepancies.