

Booklet of Code and Output
for
STAD29/STA 1007 Midterm Exam

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```

library(MASS)
library(tidyverse)

## -- Attaching packages -----
tidyverse 1.3.1 --
## v ggplot2 3.3.6      v purrr  0.3.4
## v tibble  3.1.7      v dplyr  1.0.9
## v tidyr   1.2.0      v stringr 1.4.0
## v readr   2.1.2      v forcats 0.5.1
## -- Conflicts -----
tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## x dplyr::select() masks MASS::select()

library(broom)
library(survival)
library(survminer)

## Loading required package: ggpubr
##
## Attaching package: 'survminer'
## The following object is masked from 'package:survival':
##
##   myeloma

```

Figure 1: Packages

minutes	machines
97.0	7.0
86.0	6.0
78.0	5.0
10.0	1.0
75.0	5.0
62.0	4.0
101.0	7.0
39.0	3.0
53.0	4.0
33.0	2.0
118.0	8.0
65.0	5.0
25.0	2.0
71.0	5.0
105.0	7.0
17.0	1.0
49.0	4.0
68.0	5.0

Figure 2: Copiers data

```
ggplot(copiers, aes(x=machines, y=minutes))+geom_point()
```

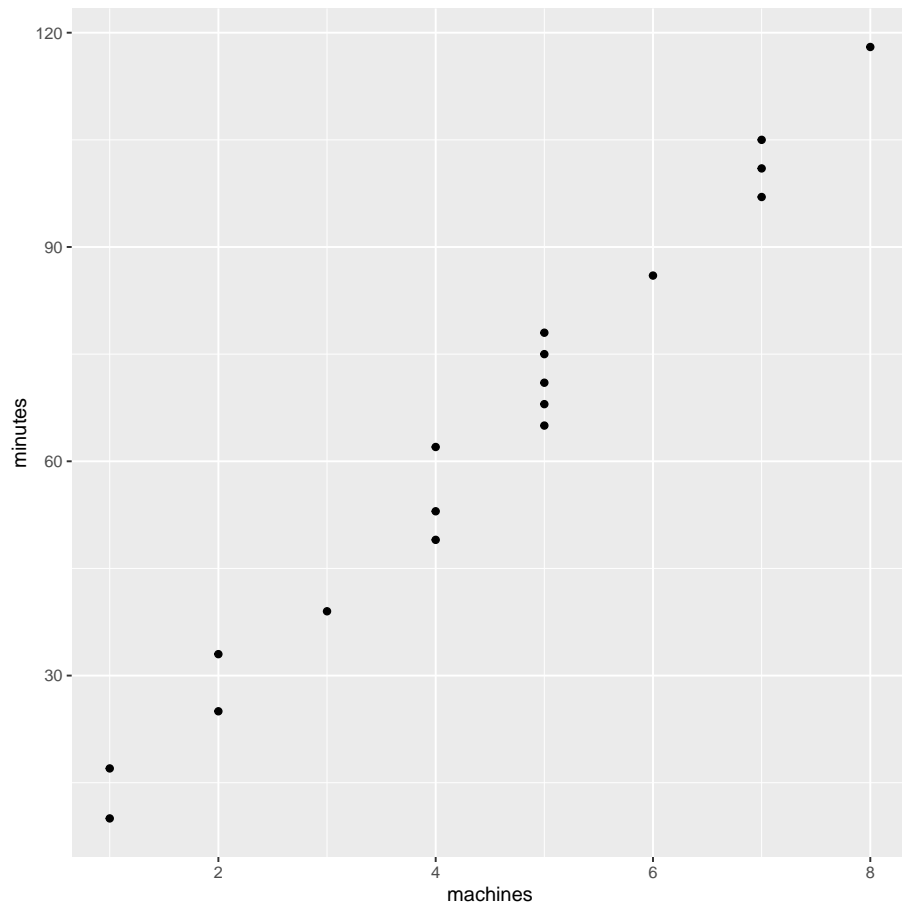


Figure 3: Copiers scatterplot

```

copiers.1=lm(minutes~machines, data=copiers)
summary(copiers.1)

##
## Call:
## lm(formula = minutes ~ machines, data = copiers)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.6309 -3.2500 -0.2383  4.0235  6.6309
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -2.3221     2.5644  -0.906   0.379
## machines     14.7383     0.5193  28.383 4.1e-15 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.482 on 16 degrees of freedom
## Multiple R-squared:  0.9805, Adjusted R-squared:  0.9793
## F-statistic: 805.6 on 1 and 16 DF,  p-value: 4.097e-15

```

Figure 4: Regression for copiers data

```

new=tibble(machines=6)
p1=predict(copiers.1, new, interval="c")
cbind(new,p1)

##   machines      fit      lwr      upr
## 1         6 86.10738 83.32504 88.88973

p2=predict(copiers.1, new, interval="p")
cbind(new,p2)

##   machines      fit      lwr      upr
## 1         6 86.10738 76.20721 96.00756

```

Figure 5: Predictions for copiers data

```

bottles
## # A tibble: 6 x 3
##   deposit sold returned
##   <dbl> <dbl> <dbl>
## 1     2   500     72
## 2     5   500    103
## 3    10   500    170
## 4    20   500    296
## 5    25   500    406
## 6    30   500    449

```

Figure 6: Soft drink bottle return data

```

bottles.1=glm(y~deposit, data=bottles, family=binomial)
summary(bottles.1)

##
## Call:
## glm(formula = y ~ deposit, family = binomial, data = bottles)
##
## Deviance Residuals:
##      1      2      3      4      5      6
## 0.1754  0.4330  0.5784 -2.9193  1.2710  1.2209
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -2.076565  0.084839  -24.48  <2e-16 ***
## deposit      0.135851  0.004772   28.47  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 1108.171  on 5  degrees of freedom
## Residual deviance:   12.181  on 4  degrees of freedom
## AIC: 53.419
##
## Number of Fisher Scoring iterations: 3

```

Figure 7: Bottles logistic regression

```
probability=predict(bottles.1, bottles, type="response")
preds=cbind(bottles, probability)
preds

##   deposit sold returned probability
## 1      2  500      72  0.1412601
## 2      5  500     103  0.1982432
## 3     10  500     170  0.3278210
## 4     20  500     296  0.6548554
## 5     25  500     406  0.7891326
## 6     30  500     449  0.8806877
```

Figure 8: Predictions for bottles data

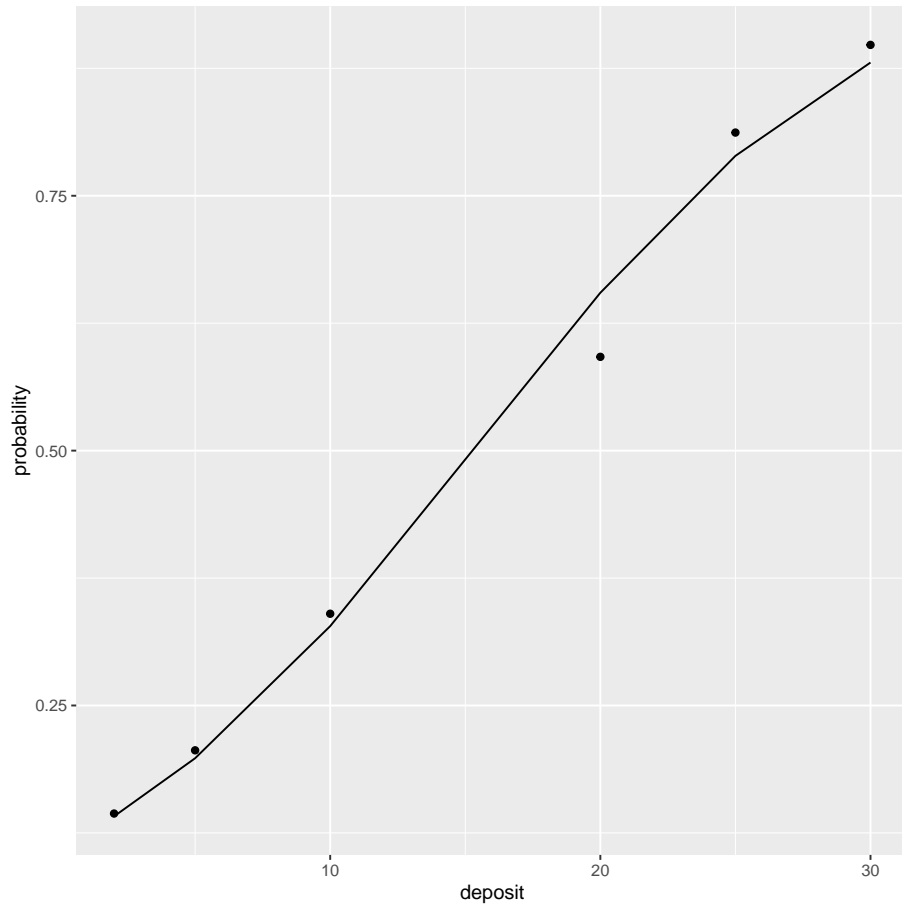


Figure 9: Plot of predictions and proportions

```

arthritis=read_table("arthritis.txt")

##
## -- Column specification -----
## cols(
##   id = col_double(),
##   treatment = col_character(),
##   sex = col_character(),
##   age = col_double(),
##   impr = col_double()
## )
## Warning: 43 parsing failures.
## row col expected actual file
## 1 -- 5 columns 6 columns 'arthritis.txt'
## 2 -- 5 columns 6 columns 'arthritis.txt'
## 3 -- 5 columns 6 columns 'arthritis.txt'
## 4 -- 5 columns 6 columns 'arthritis.txt'
## 5 -- 5 columns 6 columns 'arthritis.txt'
## ... ..
## See problems(...) for more details.

arthritis

## # A tibble: 85 x 5
##   id treatment sex age impr
##   <dbl> <chr> <chr> <dbl> <dbl>
## 1 57 Treated Male 27 1
## 2 46 Treated Male 29 0
## 3 77 Treated Male 30 0
## 4 17 Treated Male 32 2
## 5 36 Treated Male 46 2
## 6 23 Treated Male 58 2
## 7 75 Treated Male 59 0
## 8 39 Treated Male 59 2
## 9 33 Treated Male 63 0
## 10 55 Treated Male 63 0
## # ... with 75 more rows

```

Figure 10: Arthritis data (some)

```

arthritis.1=polr(factor(impr)~treatment+sex+age, data=arthritis)
drop1(arthritis.1, test="Chisq")

## Single term deletions
##
## Model:
## factor(impr) ~ treatment + sex + age
##           Df    AIC      LRT  Pr(>Chi)
## <none>           155.46
## treatment  1 168.17 14.7095 0.0001254 ***
## sex        1 159.15  5.6880 0.0170812 *
## age        1 158.03  4.5715 0.0325081 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Figure 11: Arthritis model

```

##   treatment  sex age      0      1      2
## 1  Placebo Female 46 0.6849211 0.15738536 0.15769352
## 2  Placebo Female 63 0.5318884 0.20439438 0.26371724
## 3  Placebo  Male 46 0.8837222 0.06545116 0.05082667
## 4  Placebo  Male 63 0.7988957 0.10817798 0.09292635
## 5  Treated Female 46 0.2751180 0.20743935 0.51744266
## 6  Treated Female 63 0.1655413 0.16217047 0.67228824
## 7  Treated  Male 46 0.5702499 0.19503589 0.23471420
## 8  Treated  Male 63 0.4095362 0.22067590 0.36978787

```

Figure 12: Arthritis predictions

```

## Rows: 15 Columns: 5
## -- Column specification -----
## Delimiter: "\t"
## chr (1): status
## dbl (4): id, survtime, treatment, age
##
## i Use 'spec()' to retrieve the full column specification for
this data.
## i Specify the column types or set 'show_col_types = FALSE' to
quiet this message.

## # A tibble: 15 x 5
##       id survtime status  treatment  age
##   <dbl> <dbl> <chr>    <chr>    <dbl>
## 1     1     1     1 Died      B       75
## 2     2     2     1 Died      B       79
## 3     3     3     4 Died      B       85
## 4     4     4     5 Died      B       76
## 5     5     5     6 Unknown  B       66
## 6     6     6     8 Died      A       75
## 7     7     7     9 Survived B       72
## 8     8     8     9 Died      B       70
## 9     9     9    12 Died      A       71
## 10    10    15 Unknown  A       73
## 11    11    22 Died      B       66
## 12    12    25 Survived A       73
## 13    13    37 Died      A       68
## 14    14    55 Died      A       59
## 15    15    72 Survived A       61

```

Figure 13: Patient survival data

```

patients.1=coxph(y~treatment+age, data=patients)
summary(patients.1)

## Call:
## coxph(formula = y ~ treatment + age, data = patients)
##
##   n= 15, number of events= 10
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## treatmentB 1.88484   6.58531  0.96833  1.946  0.05160 .
## age         0.21739   1.24283  0.08429  2.579  0.00991 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## treatmentB    6.585      0.1519    0.987    43.936
## age           1.243      0.8046    1.054    1.466
##
## Concordance= 0.873 (se = 0.034 )
## Likelihood ratio test= 14.41 on 2 df,  p=7e-04
## Wald test              = 9.03 on 2 df,  p=0.01
## Score (logrank) test = 12.61 on 2 df,  p=0.002

```

Figure 14: Patient survival Cox model

Treatment	Male	Female
A	22	21
	25	19
	26	18
	27	24
	24	25
B	14	21
	17	20
	19	23
	20	27
	17	25
C	15	37
	17	34
	19	36
	14	26
	12	29

Figure 15: Headache pain relief times, original layout of data

```

## Rows: 15 Columns: 3
## -- Column specification -----
##
## chr (1): Treatment
## dbl (2): Male, Female
##
## i Use 'spec()' to retrieve the full column specification for
this data.
## i Specify the column types or set 'show_col_types = FALSE' to
quiet this message.

## # A tibble: 30 x 3
##   Treatment Gender   Time
##   <chr>      <chr> <dbl>
## 1 A          Male    22
## 2 A          Female  21
## 3 A          Male    25
## 4 A          Female  19
## 5 A          Male    26
## 6 A          Female  18
## 7 A          Male    27
## 8 A          Female  24
## 9 A          Male    24
## 10 A         Female  25
## # ... with 20 more rows

painrelief.1=aov(Time~Treatment*Gender, data=painrelief)
summary(painrelief.1)

##           Df Sum Sq Mean Sq F value    Pr(>F)
## Treatment    2   71.5   35.73   3.822  0.0362 *
## Gender        1  313.6  313.63  33.544 5.70e-06 ***
## Treatment:Gender  2  521.9  260.93  27.907 5.46e-07 ***
## Residuals    24  224.4    9.35
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Figure 16: ANOVA with interaction for headache pain relief times

```

painrelief.2=update(painrelief.1, .~.-Treatment:Gender)
summary(painrelief.2)

##           Df Sum Sq Mean Sq F value Pr(>F)
## Treatment  2   71.5   35.73   1.245 0.30456
## Gender     1  313.6  313.63  10.927 0.00277 **
## Residuals 26  746.3   28.70
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

TukeyHSD(painrelief.2)

## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = Time ~ Treatment + Gender, data = painrelief)
##
## $Treatment
##      diff      lwr      upr      p adj
## B-A -2.8 -8.753648 3.153648 0.4819497
## C-A  0.8 -5.153648 6.753648 0.9405350
## C-B  3.6 -2.353648 9.553648 0.3062075
##
## $Gender
##           diff      lwr      upr      p adj
## Male-Female -6.466667 -10.48785 -2.445488 0.0027695

```

Figure 17: Analysis of main effects for headache pain relief times

```

painrelief %>% filter(Gender=="Male") -> painrelief_male
painrelief.3a=aov(Time~Treatment, data=painrelief_male)
summary(painrelief.3a)

##              Df Sum Sq Mean Sq F value    Pr(>F)
## Treatment      2  245.2   122.60   22.56 8.59e-05 ***
## Residuals     12   65.2    5.43
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

TukeyHSD(painrelief.3a)

##      Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = Time ~ Treatment, data = painrelief_male)
##
## $Treatment
##      diff          lwr          upr      p adj
## B-A -7.4 -11.333026 -3.466974 0.0008076
## C-A -9.4 -13.333026 -5.466974 0.0000968
## C-B -2.0  -5.933026  1.933026 0.3928918

```

Figure 18: Simple effects of treatment for males


```

painrelief %>% filter(Gender=="Female") -> painrelief_female
painrelief.3b=aov(Time~Treatment, data=painrelief_female)
summary(painrelief.3b)

##              Df Sum Sq Mean Sq F value    Pr(>F)
## Treatment      2  348.1  174.07   13.12 0.000955 ***
## Residuals     12  159.2   13.27
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

TukeyHSD(painrelief.3b)

##      Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = Time ~ Treatment, data = painrelief_female)
##
## $Treatment
##      diff          lwr          upr      p adj
## B-A    1.8 -4.345745  7.945745  0.7210392
## C-A   11.0  4.854255 17.145745  0.0012146
## C-B    9.2  3.054255 15.345745  0.0046858

```

Figure 19: Simple effects of treatment for females

```

## Rows: 30 Columns: 3
## -- Column specification -----
## Delimiter: ","
## dbf (3): Eth, Phil, Other
##
## i Use 'spec()' to retrieve the full column specification for
this data.
## i Specify the column types or set 'show_col_types = FALSE' to
quiet this message.

## # A tibble: 20 x 2
##   discipline score
##   <fct>         <dbl>
## 1 Eth           3
## 2 Other         5
## 3 Other         5
## 4 Phil          1
## 5 Other         4
## 6 Other         5
## 7 Phil          5
## 8 Phil          2
## 9 Phil          5
## 10 Eth          4
## 11 Other        5
## 12 Eth          5
## 13 Eth          5
## 14 Phil         5
## 15 Phil         4
## 16 Eth          2
## 17 Other        4
## 18 Eth          3
## 19 Other        5
## 20 Eth          5

```

Figure 20: Morality of eating meat data (some randomly chosen rows)

```
meat %>% group_by(discipline) %>%
  summarize(mean_score=mean(score))

## # A tibble: 3 x 2
##   discipline mean_score
##   <fct>         <dbl>
## 1 Eth           4.17
## 2 Other         4.63
## 3 Phil         3.73
```

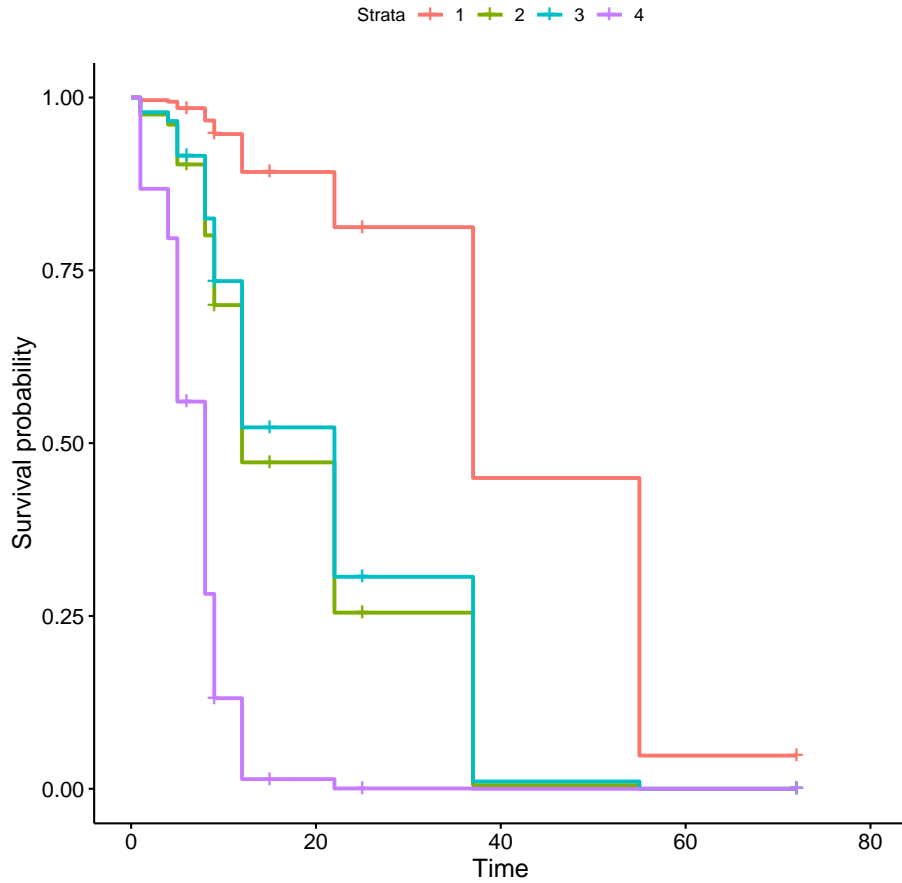
Figure 21: Mean scores by group for meat data

```
meat.1=lm(score~discipline, data=meat)
summary(meat.1)

##
## Call:
## lm(formula = score ~ discipline, data = meat)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.6333 -0.7333  0.3667  0.8333  1.2667
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   4.1778     0.1180  35.409 < 2e-16 ***
## disciplinec1  0.2167     0.1445   1.499  0.13739
## disciplinec2  0.4556     0.1669   2.730  0.00766 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.119 on 87 degrees of freedom
## Multiple R-squared:  0.1003, Adjusted R-squared:  0.07965
## F-statistic: 4.851 on 2 and 87 DF, p-value: 0.01006
```

Figure 22: Testing of contrasts for meat data

```
## Warning: 'gather()' was deprecated in tidyr 1.2.0.  
## Please use 'gather()' instead.  
## This warning is displayed once every 8 hours.  
## Call 'lifecycle::last_lifecycle_warnings()' to see where this  
warning was generated.
```



```
## # A tibble: 4 x 2  
##   age treatment  
##   <dbl> <chr>  
## 1    67 A  
## 2    67 B  
## 3    75 A  
## 4    75 B
```

Figure 23: Predicted survival curves for ages and treatments given below plot

```
## 'summarise()' has grouped output by 'Treatment'. You can  
override using the  
## '.groups' argument.
```

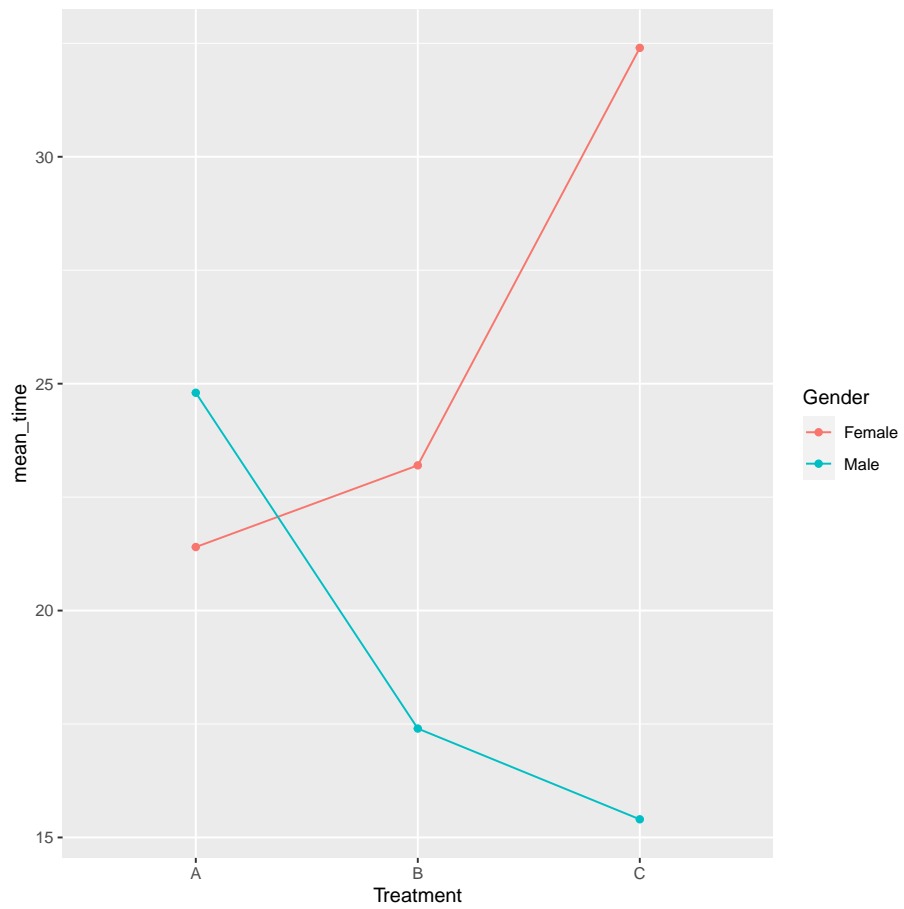


Figure 24: Headache pain relief times, interaction plot