### Homework1\_620

#### Zihan Wang

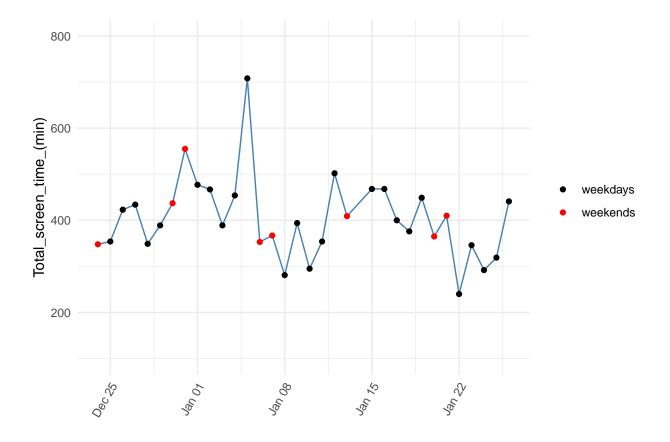
#### 2024-02-04

```
# 1(a)
# Hypothesis: High amount of screen time is associated with high levels of
# depressive symptoms
# Tang, S., Werner-Seidler, A., Torok, M., Mackinnon, A. J., & Christensen, H.
# (2021). The relationship between screen time and mental health in young
# people: A systematic review of longitudinal studies. Clinical psychology
# review, 86, 102021. https://doi.org/10.1016/j.cpr.2021.102021
# 1(b)
# Informed consent form ensures that participants are fully aware of the
# studies they are invoveld in regard of their purposes, procedures, risks and
# benefits and also the right to withdraw from the study and the
# confidentiality of the raw research data.
# 1(c)
# Screen activity is recorded in real-time by the mobile device. We collected
# daily entries of total screen time (Total.ST: total screen time in HH-MM
# format, and Total.ST.min: total screen time in MM format), social app screen
# time (Social.ST: social app screen time in HH-MM format, and Social.ST.min:
# social app screen time in MM format), total number of times the user picked up
# the phone (Pickups), and the time of the first pick-up (Pickup.1st). We are
# using the screen time data from 12/24/2023 to 01/26/2024 stored in the mobile
# device.
Dataframe <- read_excel("Screen_Time.xlsx")</pre>
Dataframe Pickup.1st <- sapply(Dataframe Pickup.1st, function(x) format(x, "%H:%M"))
Dataframe <- na.omit(Dataframe)</pre>
# 1(d)
Dataframe$Proportion_Social = Dataframe$Social.ST.min/Dataframe$Total.ST.min
Dataframe$Duration = Dataframe$Total.ST.min/Dataframe$Pickups
Dataframe$Date <- as.Date(Dataframe$Date, format = "%m/%d/%Y")
Dataframe$if weekend <- ifelse(weekdays(Dataframe$Date) %in% c("Saturday", "Sunday"),
    "Weekend", "Weekday")
```

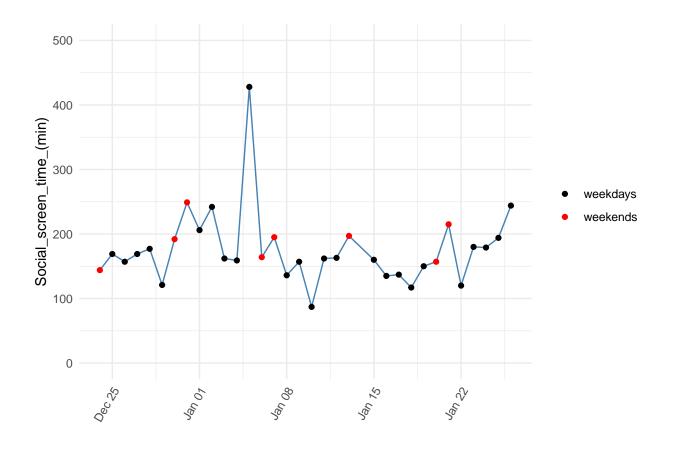
```
# Z(a)

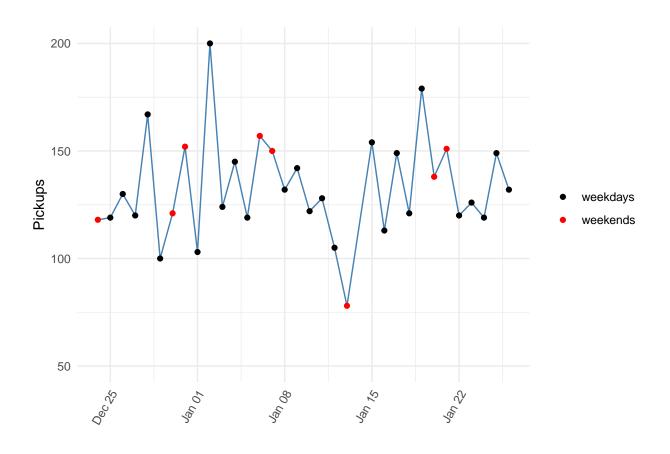
# The data collected during weekend days are colored in red in order to
# visualize any differences of mobile device use between weekdays and weekends.
# It seems that there is no obvious difference of the temporal pattern between
# weekdays and weekends in regard of five variables.

total = ggplot(Dataframe, aes(x = Date, y = Total.ST.min, color = if_weekend)) +
    geom_line(color = "steelblue") + geom_point() + xlab("") + ylab("Total_screen_time_(min)") +
    ylim(100, 800) + scale_color_manual(labels = c("weekdays", "weekends"), values = c("black",
    "red")) + theme_minimal() + theme(axis.text.x = element_text(angle = 60, hjust = 1),
    legend.title = element_blank())
print(total)
```

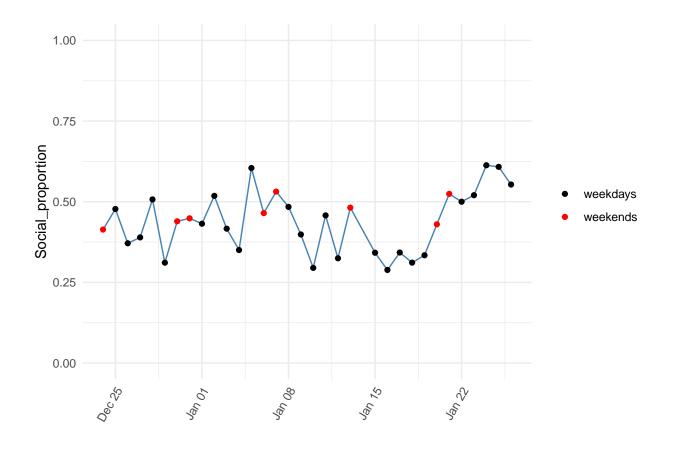


```
social = ggplot(Dataframe, aes(x = Date, y = Social.ST.min, color = if_weekend)) +
    geom_line(color = "steelblue") + geom_point() + xlab("") + ylab("Social_screen_time_(min)") +
    ylim(0, 500) + scale_color_manual(labels = c("weekdays", "weekends"), values = c("black",
    "red")) + theme_minimal() + theme(axis.text.x = element_text(angle = 60, hjust = 1),
    legend.title = element_blank())
print(social)
```

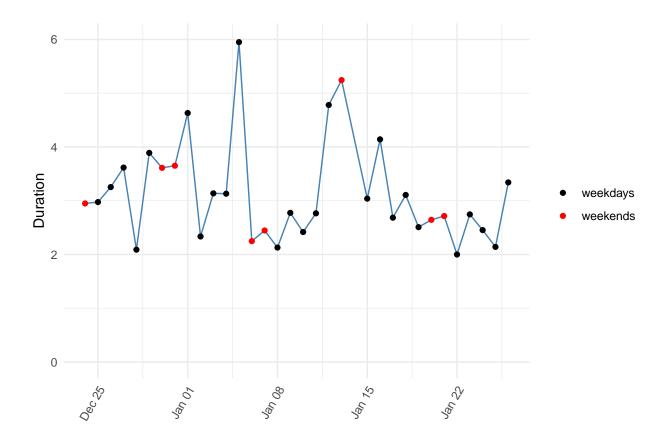


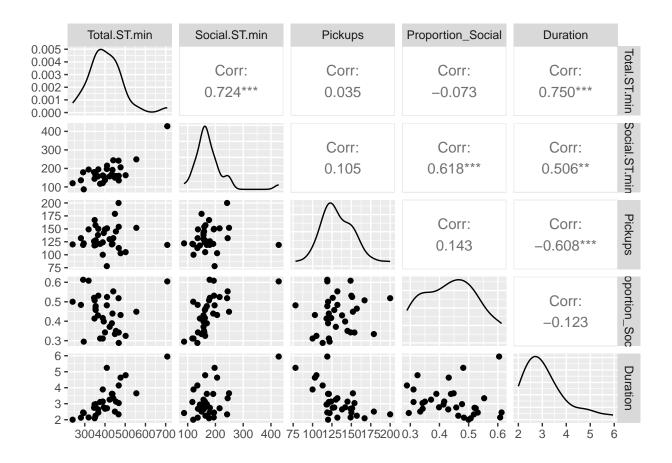


```
social_proportion = ggplot(Dataframe, aes(x = Date, y = Proportion_Social, color = if_weekend)) +
    geom_line(color = "steelblue") + geom_point() + xlab("") + ylab("Social_proportion") +
    ylim(0, 1) + scale_color_manual(labels = c("weekdays", "weekends"), values = c("black",
    "red")) + theme_minimal() + theme(axis.text.x = element_text(angle = 60, hjust = 1),
    legend.title = element_blank())
print(social_proportion)
```



```
duration = ggplot(Dataframe, aes(x = Date, y = Duration, color = if_weekend)) + geom_line(color = "stee
    geom_point() + xlab("") + ylab("Duration") + ylim(0, 6) + scale_color_manual(labels = c("weekdays",
    "weekends"), values = c("black", "red")) + theme_minimal() + theme(axis.text.x = element_text(angle
    hjust = 1), legend.title = element_blank())
print(duration)
```



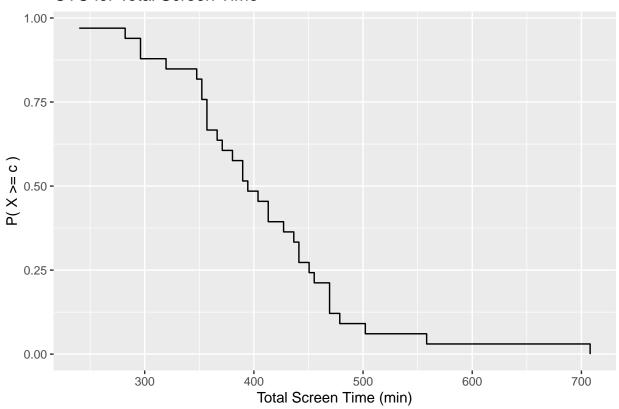


```
# 2(c)

# For total screen time, social screen time and pickups after the patterns are
# quite similar, that is after a certain point the curves drop dramatically,
# indicating that high screen time, high social screen time and high pickups
# are relatively less common.While for the daily proportion of the social
# screen time and daily duration per use the curves are relatively smooth.

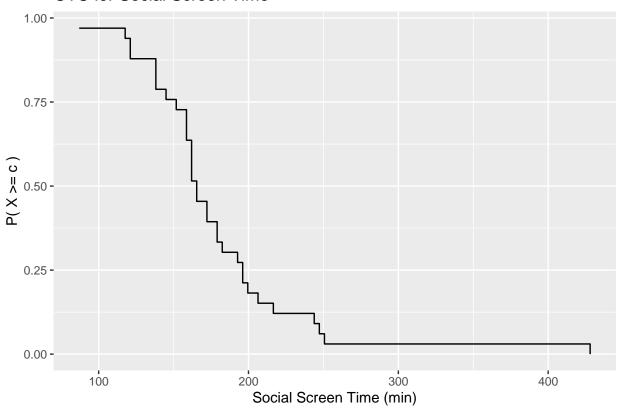
ggplot(Dataframe, aes(x = Total.ST.min)) + stat_function(fun = function(x) 1 - ecdf(Dataframe$Total.ST.min)) + stat_function(fun = function(x) 1 - ecdf(Dataframe$Total.ST.min)) + graph with the patterns are
# 2(c)
# For total screen time and pickups after the patterns are
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# 2(c)
#
```

### OTC for Total Screen Time

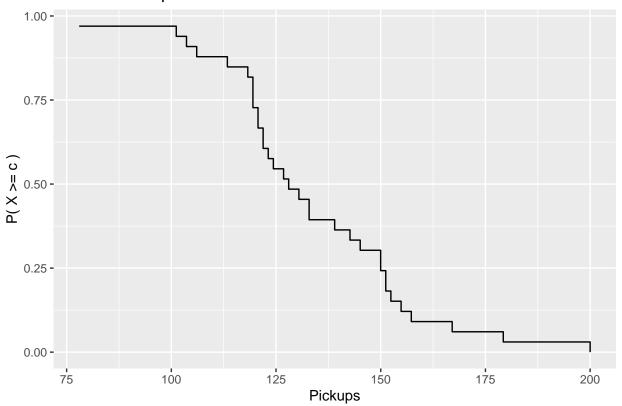


```
ggplot(Dataframe, aes(x = Social.ST.min)) + stat_function(fun = function(x) 1 - ecdf(Dataframe$Social.S'
geom = "step") + labs(title = "OTC for Social Screen Time", x = "Social Screen Time (min)",
y = "P( X >= c )")
```

### OTC for Social Screen Time

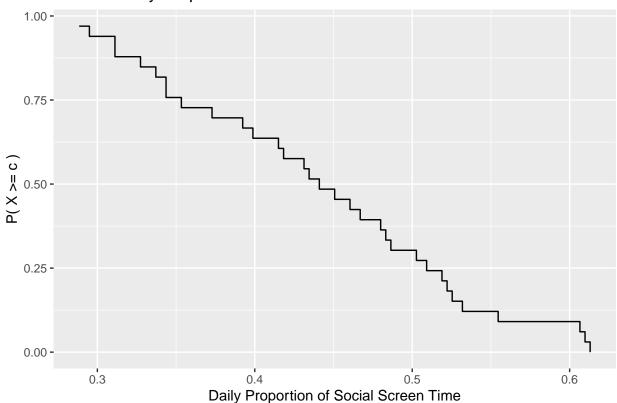


## OTC for Pickups

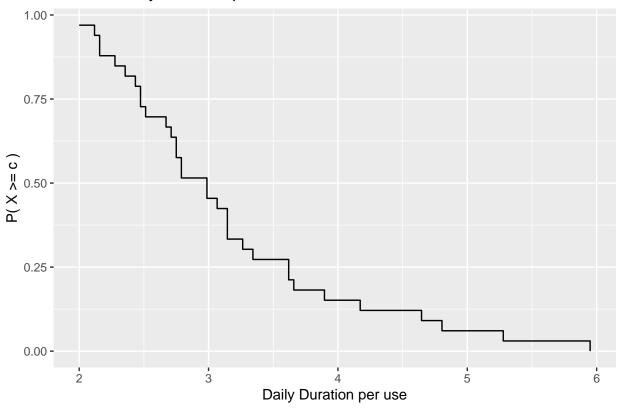


```
ggplot(Dataframe, aes(x = Proportion_Social)) + stat_function(fun = function(x) 1 -
    ecdf(Dataframe$Proportion_Social)(x), geom = "step") + labs(title = "OTC for Daily Proportion of So
    x = "Daily Proportion of Social Screen Time", y = "P( X >= c )")
```

## OTC for Daily Proportion of Social Screen Time

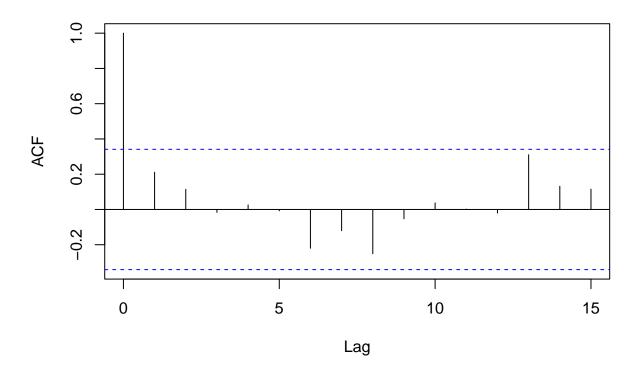


## OTC for Daily Duration per use



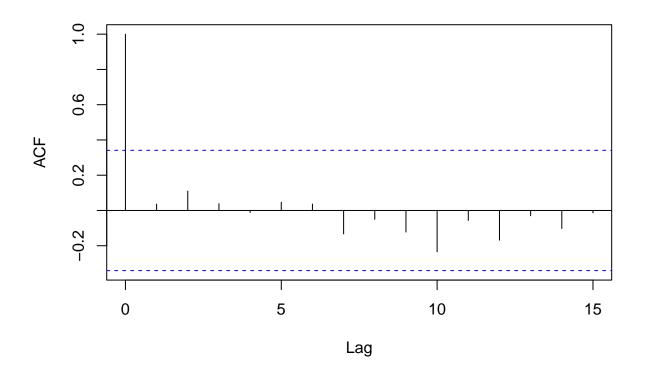
```
# 2(d)
# From the acf plot we can infer the only possible autocorrelation is in regard
# of daily proportion of the social screen time.
acf(Dataframe$Total.ST.min)
```

# Series Dataframe\$Total.ST.min



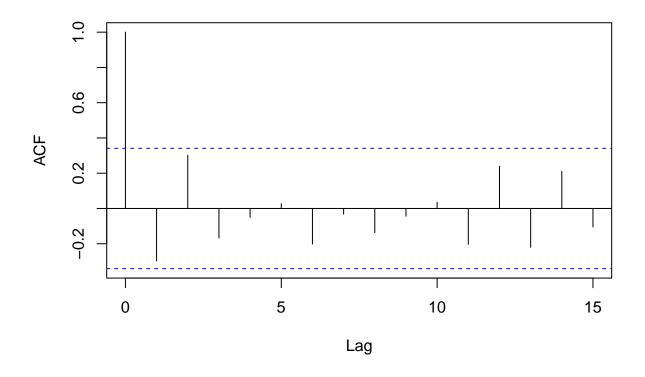
acf(Dataframe\$Social.ST.min)

## Series Dataframe\$Social.ST.min



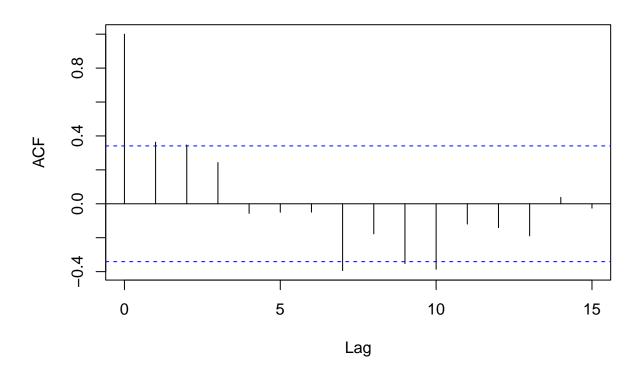
acf(Dataframe\$Pickups)

# Series Dataframe\$Pickups



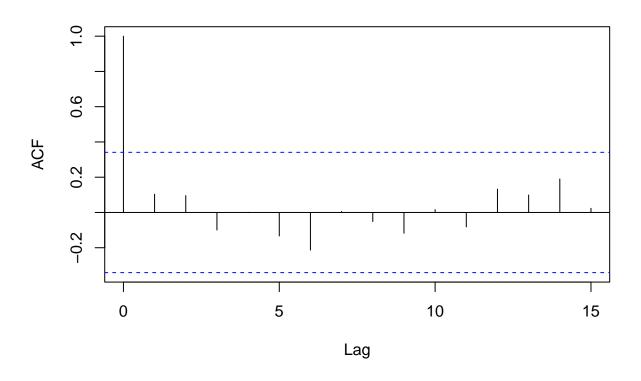
acf(Dataframe\$Proportion\_Social)

# Series Dataframe\$Proportion\_Social



acf(Dataframe\$Duration)

### Series Dataframe\$Duration



```
acf(Dataframe$Total.ST.min, plot = FALSE)
##
## Autocorrelations of series 'Dataframe$Total.ST.min', by lag
##
##
                   2
                         3
                                     5
                                           6
                                                7
               1.000 0.211
##
##
      11
            12
                  13
                        14
                              15
   0.002 -0.020 0.311 0.132 0.115
acf(Dataframe$Social.ST.min, plot = FALSE)
##
## Autocorrelations of series 'Dataframe$Social.ST.min', by lag
##
                                                7
             1
                   2
                         3
                                     5
  1.000 0.036 0.110 0.038 -0.011 0.046 0.037 -0.133 -0.050 -0.122 -0.235
##
            12
                  13
                        14
## -0.056 -0.168 -0.030 -0.102 -0.014
acf(Dataframe$Pickups, plot = FALSE)
##
```

## Autocorrelations of series 'Dataframe\$Pickups', by lag

```
##
##
                                                  6
                                                         7
                                                                             10
               1
                      2
                             3
                                   4
                                           5
                                                                8
   1.000 -0.297 0.302 -0.168 -0.050 0.026 -0.202 -0.032 -0.137 -0.044 0.035
##
              12
                     13
                            14
                                  15
## -0.204 0.240 -0.220 0.210 -0.105
acf(Dataframe$Proportion Social, plot = FALSE)
##
## Autocorrelations of series 'Dataframe$Proportion_Social', by lag
##
                      2
                             3
                                    4
                                           5
                                                  6
                                                         7
               1
                                                                8
   1.000 0.363 0.347 0.243 -0.057 -0.050 -0.049 -0.394 -0.177 -0.353 -0.386
##
##
              12
                     13
                            14
## -0.120 -0.141 -0.189 0.038 -0.026
acf(Dataframe$Duration, plot = FALSE)
##
## Autocorrelations of series 'Dataframe$Duration', by lag
                      2
                             3
                                    4
                                           5
                                                  6
                                                         7
                                                                             10
##
               1
                                                                8
## 1.000 0.104 0.096 -0.099 0.001 -0.133 -0.213 0.005 -0.051 -0.118 0.016
       11
              12
                     13
                            14
                                   15
## -0.082 0.132 0.099 0.190 0.024
# 3(a)
Dataframe$Pickup.1st <- as.POSIXct(Dataframe$Pickup.1st, format = "%H:%M")
Dataframe = Dataframe %>%
    mutate(Pickup.1st.angular = (hour(Pickup.1st) * 60 + minute(Pickup.1st))/(24 *
        60) * 360)
# 3(b)
# The first pickup time is mainly concentrated between 12am and 1am.
first.pickup.cir = circular(Dataframe$Pickup.1st.angular, units = "degrees", template = "clock24")
png("HD1 C3(b).png")
plot(first.pickup.cir, col = "blue")
dev.off()
## pdf
##
# 3(c)
# I choose bin size to be 48, which corresponds to an interval of half an hour.
png("HD1 C3(c).png")
plot(first.pickup.cir, stack = TRUE, bins = 48, col = "red")
dev.off()
```

```
## pdf
##
# 4(a)
# The Yt is the daily number of pickups, but the rate lambda is the expected
# hourly rate, there exists an inconsistency between the unit.
# 4(b)
Dataframe$Total.ST.Hour = Dataframe$Total.ST.min/60
model = glm(Pickups ~ offset(log(Total.ST.Hour)), family = "poisson", data = Dataframe)
lambda = exp(coef(model))
print(lambda)
## (Intercept)
     19.75362
# 4(c)
# (c.1) the p-value for Xt is 0.9139, so there is no evidence of significantly
# different behavior of daily pickups between weekdays and weekends using the
# significance level alpha=0.05
# (c.2)the p-value for Zt is 0.0506, so there is no evidence of significantly
# different behavior of daily pickups between whether or not being days on
# winter break.
Dataframe\frac{1}{1} break = c(rep(0, 17), rep(1, 16))
Dataframe$if_weekend_dummy = ifelse(Dataframe$if_weekend == "Weekday", 1, 0)
model_ <- glm(Pickups ~ if_break + if_weekend_dummy, family = poisson(link = "log"),</pre>
    data = Dataframe, offset = log(Total.ST.Hour))
summary(model_)
##
## Call:
## glm(formula = Pickups ~ if_break + if_weekend_dummy, family = poisson(link = "log"),
       data = Dataframe, offset = log(Total.ST.Hour))
## Coefficients:
##
                    Estimate Std. Error z value Pr(>|z|)
                     2.95839
                               0.03274 90.361
                                                 <2e-16 ***
## (Intercept)
## if_break
                     0.05949
                                0.03043
                                         1.955
                                                  0.0506 .
## if_weekend_dummy -0.00383
                                0.03543 -0.108
                                                  0.9139
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for poisson family taken to be 1)
##
       Null deviance: 306.43 on 32 degrees of freedom
##
```

```
## Residual deviance: 302.60 on 30 degrees of freedom
## AIC: 530.11
## Number of Fisher Scoring iterations: 4
# 5(a)
Dataframe$Pickup.1st.pi = (Dataframe$Pickup.1st.angular/180) * pi
mle_result <- mle.vonmises(Dataframe$Pickup.1st.pi)</pre>
## Warning in as.circular(x): an object is coerced to the class 'circular' using default value for the
    type: 'angles'
    units: 'radians'
##
##
    template: 'none'
##
    modulo: 'asis'
##
    zero: 0
##
    rotation: 'counter'
## conversion.circularxradians0counter2pi
print(mle_result)
##
## Call:
## mle.vonmises(x = Dataframe$Pickup.1st.pi)
## mu: 0.09784 ( 0.04751 )
## kappa: 13.94 ( 3.363 )
# 5(b)
pvonmises_result <- pvonmises((127.5/180) * pi, mu = 0.09784, kappa = 13.94)</pre>
## Warning in as.circular(x): an object is coerced to the class 'circular' using default value for the
     type: 'angles'
     units: 'radians'
##
##
    template: 'none'
    modulo: 'asis'
##
##
    zero: 0
    rotation: 'counter'
## conversion.circularqradians0counter
## Warning in as.circular(x): an object is coerced to the class 'circular' using default value for the
##
    type: 'angles'
    units: 'radians'
##
##
    template: 'none'
    modulo: 'asis'
##
##
    zero: 0
    rotation: 'counter'
##
## conversion.circularmuradiansOcounter
```

# The return value of the pronmises is nearly one, which means the probability # of my first pickup being 8:30 AM or later is nearly 0